

# “A genomic and bioinformatic-based approach to identify genetic variants for liver cancer across multiple continents”

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1	Submit Artikel	11 Agustus 2023	
2	Artikel mendapatkan hasil review dari 2 reviewer	15 November 2023	

## History Artikel

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Status	<b>Accepted</b>				
** Title	A Genomic and Bioinformatic-based Approach to Identify Genetic Variants for Liver Cancer across Multiple Continents				
Running Title	Identify genomic variants for Liver Cancer				
** Abstract	<p>Liver cancer is the fourth leading cause of death worldwide. Well-known risk factors include hepatitis B virus (HBV) and hepatitis C virus (HCV), along with exposure to aflatoxins, excessive alcohol consumption, obesity, and type 2 diabetes. Genomic variants play a crucial role in mediating liver cancer among these factors. This study utilizes a bioinformatics approach to identify genetic variants associated with liver cancer from across various continents. The single nucleotide polymorphisms (SNPs) associated with liver cancer were retrieved from the Genome-Wide Association Studies (GWAS) catalog. The subsequent prioritization was performed using functional annotation with HaploReg v4.1 and Ensembl database. The prevalence and allele frequencies of each variants were evaluated by using pearson correlation. Our results indicate that two variants, rs2294915 and rs2896019, encoded by the PNPLA3 gene, were found to be highly expressed in the liver tissue, as well as in the skin, cell-cultured fibroblasts and adipose-subcutaneous tissue, all of which contribute to the risk of liver cancer. We further obtained that these two SNPs (rs2294915 and rs2896019) were positively correlated with the prevalence rate. Positive association of prevalence rates were underlined more frequent in East Asian and African population. We highlight the utility of this population-specific PNPLA3 genetic variant for genetic association studies and for early prognosis and treatment of liver cancer. This study highlights the potential of integrating genomic databases with bioinformatic analysis to identify genetic variations involved in the pathogenesis of liver cancer. We recommend that future research prioritize the validation of these variations in clinical settings.</p>				
** Keywords	Genomic Variants, Liver Cancer, Bioinformatics				
English Proof-reading	Yes				

Pada tanggal 15 November 2023 artikel mendapatkan revisian yang pertama oleh 2 reviewers.

::: 1st Review :::	
Editorial Comment	
Invited date	Sep 14, 2023
Complete Review Date	Nov 15, 2023
Recommendation	<b>Review after major revision</b>
Comments to the Author	<p>• <b>Reviewer A :</b></p> <p>This paper found new association results about liver cancer using various bioinformatics tools such as GWAS catalog, HaploReg, eQTL analysis, etc. I think this paper needs a lot of improvements to be accepted for publication. Please address following issues as much as you can, and I will make a decision based on the revised manuscript.</p> <ol style="list-style-type: none"> <li>The english writing is not good. There are lots of awkward or imperfect expressions. Please take a professional english editing to improve the overall quality I listed some of the examples below. - page2, in the abstract, "Liver cancer remains a global burden, ranking fourth in mortality rates globally" : being ranked fourth may be more appropriate. - page4, "One of the websites through a bioinformatics approach that discusses gene variation is Genome-Wide Association Studies (GWAS)" : GWAS should be changed to GWAS catalog, and a period is needed after the sentence.</li> <li>Detailed explanations are not provided. - Only HaploReg results are provided without any interim process. - Aren't the rest of the significant SNPs from the GWAS catalog associated to any genes?</li> <li>Some presentations are not correct. - In Fig 4, Africa pie chart are not matched with the proportions of the SNPs</li> <li>Interpretation is not enough - In the last step, they provide allele frequencies for each of continents, but the association between allele frequencies and some statistics for the liver cancer is not provided.</li> </ol> <p>• <b>Reviewer B :</b></p> <p>This submitted article suggested a pipeline that can draw more meaningful results using several bioinformatics databases and analysis tools. However, the writing of the manuscript is not clear to understand and should be improved more. I listed some of points to be improved to make it better manuscript. - In the part of providing statistics of SNPs from multiple continents, information in Table 4 is not matched with the information in Figure 4 - No interpretation is provided for continent specific allele information to the disease. - In the eQTL analysis, I could not understand how the two tissues in the Table 3 are associated to the disease.</p>

Pada tanggal 28 November 2023 artikel mendapatkan revisian yang kedua oleh reviewers.

::: 2nd Review :::	
Editorial Comment	
Invited date	Nov 28, 2023
Complete Review Date	Nov 28, 2023
Recommendation	<b>Accept after minor revision</b>
Comments to the Author	<ul style="list-style-type: none"><li>- Although the reviewers pointed out that Figure 4 was not correct, they did not fully correct the figure, for example, rs2294915 in Africa, the pie chart seems still not correct. Please correct it.</li><li>- The interpretation using the allele information from the continents are still not clear. There are not enough information about association between allele frequency and the prevalence of the liver cancer of each continent. Please add some more contents based on the actual prevalence information.</li></ul>

Pada tanggal 29 November 2023 hasil revisian yang kedua telah diterima oleh editor jurnal. Dan pada tanggal 30 November 2023 artikel telah diterima / accepted oleh editor.

Editorial Comment	
Invited date	Nov 29, 2023
Complete Review Date	Nov 30, 2023
Recommendation	<b>Accept as it is</b>
Comments to the Author	<b>None (or N/A)</b>
Files	<b>None (or N/A)</b>

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Berikut akan dilampirkan beberapa dokumen perubahan dari artikel yang diterbitkan sejak proses submit, hasil review dari reviewer, artikel revisi, dan artikel versi terbit.

**Artikel pertama kali saat disubmit  
pada tanggal 11 Agustus 2023**

# **An Insight of Genomic Variants and Bioinformatic-based Approach Mediating Liver Cancer across Multiple Continents**

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## Abstract

Liver cancer remains a global burden, ranking fourth in mortality rates globally. Risk factors such as hepatitis B virus (HBV) and hepatitis C virus (HCV) have been widely reported. Additionally, exposure to contamination with aflatoxin, alcohol consumption, obesity, type 2 diabetes, and genomic variations have been investigated as potential risk factors. Genomic variants play a crucial role in mediating liver cancer among these factors. However, specific variants involved in this process are still limitedly studied. This study uses a bioinformatics approach to identify genetic variants associated with liver cancer from across various continents. The single nucleotide polymorphisms (SNPs) associated with liver cancer were retrieved from the Genome-Wide Association Studies (GWAS) catalog. The subsequent prioritization was performed using functional annotation with HaploReg v4.1 and the Ensembl database. Our results indicate that two variants, rs2294915 and rs2896019, encoded by the *PNPLA3* gene, were found to be highly expressed in the liver tissue, as well as in the skin, cell-cultured fibroblasts and adipose-subcutaneous tissue, all of which contribute to the risk of liver cancer. We highlight the utility of this population-specific *PNPLA3* genetic variant for genetic association studies and for early prognosis and treatment of liver cancer. This study emphasizes that integrating genomic databases and bioinformatic analysis is a promising approach to identify genetic variations that play a role in the pathogenesis of liver cancer. We suggest that future researchers focus on these gene variations to be validated in clinical studies.

**Keywords:** *Genomic Variants, Liver Cancer, Bioinformatics*

## Introduction

Liver cancer is a type of carcinoma that has the highest mortality rate in the world every year (McGlynn et al., 2021). There were 841,000 cases of liver cancer in 2018, of which the death rate caused by liver cancer reached 782,000 (Bray et al., 2018). Cases of liver cancer are an average number of cases and deaths that can increase 2 to 3 times in men in parts of the world. According to the Global Cancer Statistics (GLOBOCAN), in 2020, liver cancer occupies the third position (8.3%) as a deadly disease due to cancer. Liver cancer had an incidence of 905 thousand cases in 2020 and a mortality rate of 830 thousand (Sung et al., 2021). In Indonesia, liver cancer is the second most common in men, amounting to 12.4 per 100,000 of the Indonesian population, with an average death rate of 7.6 per 100,000 (Kemenkes RI, 2019).

Factors that cause liver cancer include chronic infection with hepatitis B virus (HBV), hepatitis C virus (HCV), the result of contamination with aflatoxin, alcohol consumption, history of obesity, history of type 2 diabetes, and smoking addiction (Bray et al., 2018). According to Villanueva (2019), other risk factors are thought to exacerbate the occurrence of liver cancer, such as an unhealthy lifestyle, geographic conditions, gender, age, family history of the disease, and the severity of damage to the liver. Liver cancer is also found in areas that have cases of hepatitis B. In these areas, liver cancer is prevalent at a young age. This is because some infected with hepatitis B are obtained vertically through the delivery process (Mittal & El-Serag, 2013).

Patients often felt complaints in the form of fatigue, pain, diarrhea, skin abnormalities, and decreased appetite, all of which have affected their quality of life (Waller et al., 2015). Therefore, detecting the presence of disease symptoms in liver cancer can be done by examining deoxyribonucleic acid (DNA). Gene variation can be associated with disease progression and



pathogenesis, which includes liver cancer. One of the websites through a bioinformatics approach that discusses gene variation is Genome-Wide Association Studies (GWAS) GWAS is a database with single nucleotide polymorphism (SNP) search results that has identified several variants associated with liver fat content, circulating liver enzymes, and the development of NAFLD as well as genetic markers used in predicting a disease disorder (Wang et al., 2021).

Genetic identification in humans aims to identify inherited genetic risk factors for liver cancer. This study uses the GWAS database to map genes from genetic variations across several populations that play an essential role in the pathogenesis of liver cancer. The most significant gene variations based on their function in protein changes will be further verified.

## Methods



**Figure 1.** Analysis methodology for integrated bioinformatic, database and genomic analysis of genetic variation that affect liver cancer. The figure was created with BioRender.com under agreement number “FM2500073C”

In this study, we adopted the method used by Ma’ruf et al (2023) to identify genomic variants associated with Stevens-Johnson syndrome (SJS). The implementation of the methodology is illustrated in Figure 1. Liver cancer-associated SNPs were obtained from the GWAS Catalog of the National Human Genome Research Institute (NHGRI) GWAS Catalog database (<http://www.ebi.ac.uk/gwas>) (accessed 15-02-2023). Subsequently, we performed further analysis using HaploReg (version 4.1). The  $p$ -value  $< 10^{-8}$  was applied to account for multiple tests in the GWAS Catalog, as this threshold is commonly used to identify associations between common genetic variants and traits with adjacent gene expression (Chen et al., 2021). Furthermore, to evaluate the relationships between various genetic variants and gene expression profiles, we utilized e-QTL analysis with data from the GTEx Portal database (<http://www.gtexportal.org/home/>) (accessed on 16-02-2023), considering gene expression across



various tissues in humans. Additionally, we confirmed the identified variants using the Ensembl Genome Browser (<https://www.ensembl.org/index.html>) (accessed on 17-02-2023). For this study, we considered allele frequencies in populations from Europe, Africa, America, East Asia, and Southeast Asia. Then, to understand the functions of the various gene variants, we performed evaluations using the SNP nexus database (<https://www.snp-nexus.org>) (accessed on 20-02-2023).

## Results and Discussion

### 1. Identification of Genomic Variants of Liver Cancer

This study identified SNPs associated with liver cancer from the GWAS database. Among them, 29 SNPs were further confirmed through SNP duplication, as shown in Table 1. Subsequently, HaploReg version 4.1 was utilized, and a p-value  $<10^{-8}$  was applied based on the number of SNPs obtained. Based on the findings presented in Table 2, we found the risk of two genes for “Liver Cancer” disease. This study analyzed tissue expression affecting liver cancer with the missense variant *PNPLA3*.

Through our integrative bioinformatics approach, two variants with a missense mutation (rs rs2294915, rs2896019) that encoded the *PNPLA3* genes were prioritized as the biological risk SNPs for Liver Cancer. Primary liver cancer, also known as hepatocellular carcinoma, is a pathological condition characterised by the development of malignant cells within the hepatic tissues. The development of cancer in extraneous anatomical sites that then metastasizes to the liver does not constitute primary liver cancer. Primary liver cancer encompasses many kinds, including hepatocellular carcinoma (HCC), intrahepatic cholangiocarcinoma (iCCA), and less frequent varieties such as mixed hepatocellular cholangiocarcinoma (HCC-CCA), fibrolamellar HCC (FLC), and the paediatric neoplasm hepatoblastoma (Wage et al., 2021).

**Table 1.** SNPs from the GWAS catalog with p-value  $<10^{-8}$

No.	Variation and risk allele	p-value
1	rs2856723	$3 \times 10^{-43}$
2	rs34675408	$1 \times 10^{-32}$
3	rs9272105	$5 \times 10^{-22}$
4	rs913493	$5 \times 10^{-20}$
5	rs2294915	$2 \times 10^{-19}$
6	rs17401966	$2 \times 10^{-18}$
7	rs3096380	$1 \times 10^{-17}$
8	rs9275319	$3 \times 10^{-17}$
9	rs584368	$2 \times 10^{-14}$
10	rs2596542	$4 \times 10^{-13}$
11	rs1110446	$9 \times 10^{-13}$
12	rs58489806	$3 \times 10^{-12}$
13	rs6078460	$2 \times 10^{-11}$
14	rs2523961	$6 \times 10^{-11}$
15	rs7574865	$2 \times 10^{-10}$
16	rs1110446	$3 \times 10^{-10}$
17	rs455804	$5 \times 10^{-10}$
18	rs58542926	$6 \times 10^{-10}$
19	rs2523961	$6 \times 10^{-10}$
20	rs8107030	$8 \times 10^{-10}$
21	rs10272859	$9 \times 10^{-10}$
22	rs190121281	$4 \times 10^{-9}$
23	rs9275572	$6 \times 10^{-9}$
24	rs2242652	$6 \times 10^{-9}$

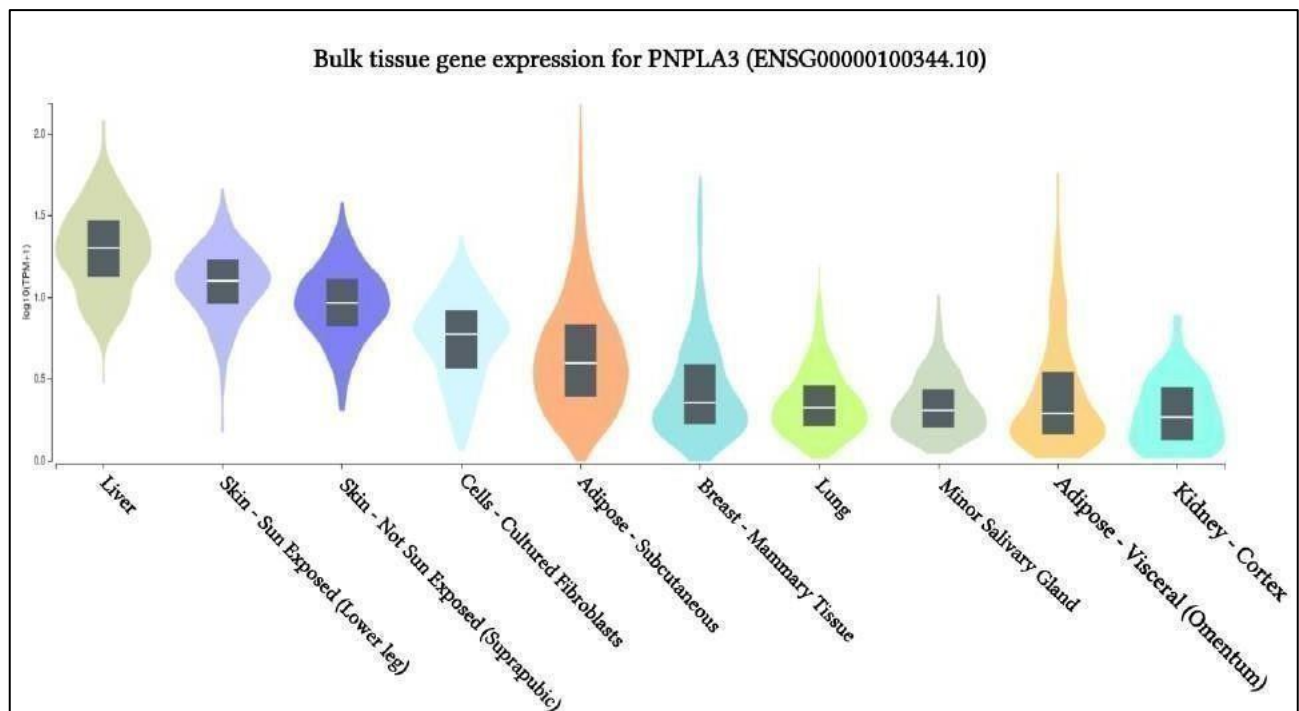
25	rs188273166	$1 \times 10^{-8}$
26	rs708113	$1 \times 10^{-8}$
27	rs2896019	$2 \times 10^{-8}$
28	rs17047200	$3 \times 10^{-8}$
29	rs541860626	$5 \times 10^{-8}$

**Table 2.** Variants and risk alleles of liver cancer encoding prioritized SNPs

Variation and risk alleles	Variants near risk allele ( $r^2 > 0.8$ )	<i>p</i> -value	Gencode	Type of allele
rs2294915	rs738409	$2 \times 10^{-19}$	<i>PNPLA3</i>	missense
rs2896019	rs3761472	$2 \times 10^{-8}$	<i>PNPLA3</i>	missense

## 2. Gene expression of *PNPLA3* in 10 human tissues

The results of *PNPLA3* gene expression in 10 human tissues comprise the most apparent functional consequences of genetic variation. Liver, sun-exposed skin (lower legs), non-sun-exposed skin (suprapubic), and adipose subcutaneous fibroblasts and cell cultures showed the highest *PNPLA3* gene expression in the 10 human tissues analyzed from GTEx (Figure 2). In addition, we have found that the SNP IDs rs2294915 and rs2896019 have similar gene expression variations in Sun-Exposed skin (lower legs).



**Figure 2.** *PNPLA3* gene expression associated with liver cancer in several human tissues based on GTEx Portal analysis

## 3. Correlation between Gene Expression of *PNPLA3* and eQTL

The result in a correlation between the Gene Expression of *PNPLA3* and eQTL, we identified an allele of rs2294915 and rs2896019 in *PNPLA3* directly related to liver cancer. As shown in Table 3 and Figure 3, the CC genotypes rs2294915, and rs2896019 were associated with higher

expression of *PNPLA3* in sun-exposed (lower leg) and non-sun-exposed (suprapubic) skin tissues compared to the genotypes TT.

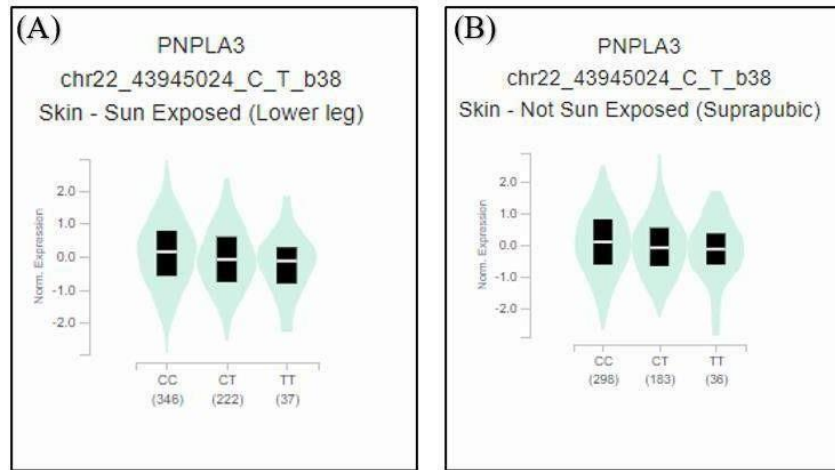
**Table 3.** Results of e-QTL in liver cancer from the GTEx portal database

SNP	Gencode ID (ENSG00000-)	Gene Symbol	p-value	Effect Size	Tissue	Expression Level
rs2294915	100344.10	<i>PNPLA3</i>	$2.8 \times 10^{-8}$	-0.15	Skin - Sun Exposed (Lower leg)	CC>CT>TT
	100344.10	<i>PNPLA3</i>	$5 \times 10^{-8}$	-0.50	Skin - Not Sun Exposed (Suprapubic)	CC>CT>TT
rs2896019	100344.10	<i>PNPLA3</i>	$6.7 \times 10^{-11}$	-0.19	Skin - Sun Exposed (Lower leg)	CC>CT>TT
	100344.10	<i>PNPLA3</i>	$2 \times 10^{-9}$	-0.22	Skin - Not Sun Exposed (Suprapubic)	CC>CT>TT

Source: Expression Quantitative Trait Loci (eQTL) obtained from the GTEx Portal.

Note : *PNPLA3*, patatin like phospholipase 3; SNP, single nucleotide polymorphism; eQTL, expression quantitative trait loci; SNP, single nucleotide polymorphism.

This study found that gene expression in *PNPLA3* at rs2294915 and rs2896019 has a link with “liver cancer.” Furthermore, rs2294915 and rs2896019 encoded missense mutations and the CC genotype had the highest expression of *PNPLA3* in sun-exposed (lower leg) and non-sun-exposed (suprapubic) skin tissue, and the TT genotype showed the most melancholy expression (Fig 3). According to Trepo et al (2022) at rs2294915, the *PNPLA3* locus on chromosome 22q13.31 has also been linked to alcohol-related liver cancer ( $p=3.71 \times 10^{-7}$ ), and at rs2896019, it has a single nucleotide polymorphism relationship with liver steatosis in obese pediatric patients in children and adolescents (Stasinou et al., 2022).



**Figure 3.** Patatin like phospholipase 3 (*PNPLA3*) gene expression for each genotype of the single nucleotide polymorphism (SNP): (A) rs2294915 and (B) rs2896019.

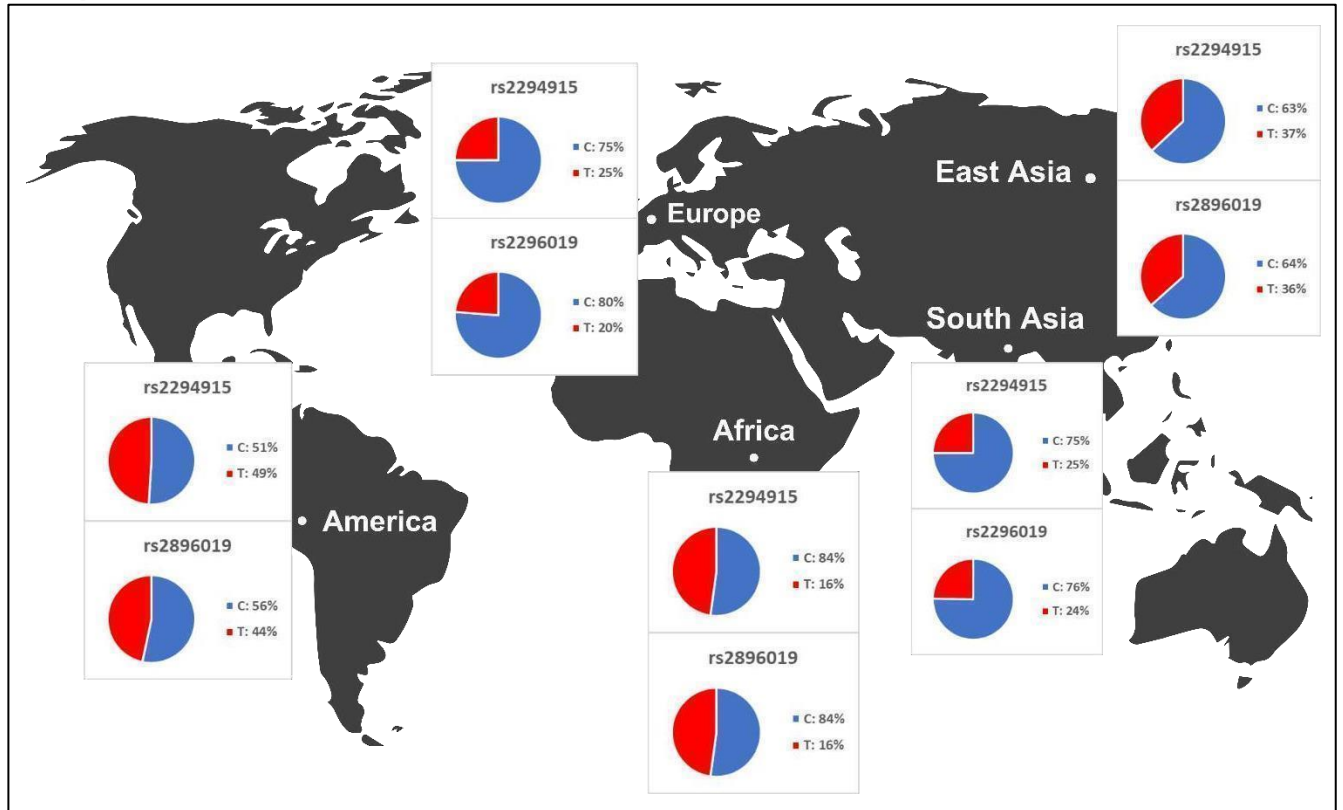
#### 4. Allele frequencies of candidate variants in populations in different continents

The results of the research we have done, we have identified variants associated with liver cancer gene expression and carried out allele frequency analysis in various populations. As shown in Table 4, allele variant frequencies were evaluated in multiple people from Europe, America, East Asia, South Asia, and Africa. Allele frequencies across populations varied for each SNP, as depicted in Figure 4.

**Table 4.** Analysis of allele frequencies for the *PNPLA3* gene from variant annotation (SNPnexus)

SNP	Gene	Location	Allele		Allele Frequency (N)				
			Ref	Eff*	AFR	AMR	EAS	EUR	SAS
rs2294915	<i>PNPLA3</i>	Missense	C	T	T: 0.163 (215)	T: 0.490 (340)	T: 0.365 (368)	T: 0.252 (254)	T: 0.246 (241)
rs2896019	<i>PNPLA3</i>	Missense	C	T	T: 0.842 (1113)	T: 0.562 (390)	T: 0.636 (641)	T: 0.801 (806)	T: 0.764 (747)

Note : *PNPLA3*, patatin like phospholipase 3; SNP, single nucleotide polymorphism; Ref, Reference; Alt, slternate; AFR, Africa; AMR, America; EAS, East Asia; EUR, Europe; SAS, Southeast Asia.



**Fig 4.** The results of the distribution of allele frequencies affecting *PNPLA3* across various populations. EUR, Europe; AFR, Africa; SAS, South Asia; EAS, East Asia.

Based on this finding, rs2294915 and rs2896019 are potentially related to the susceptibility to “liver cancer with effect size the highest score of -0.50 can be interpreted on the skin not exposed to sunlight (suprapubic).” According to Poggiali & Vercelli (2023), this condition is characterized by a disruption in the heme biosynthesis pathway due to reduced hepatic uroporphyrinogen decarboxylase (UROD) activity. The consequence of this phenomenon is the buildup of light-sensitive by-products, such as uroporphyrinogen, leading to the development of fragility and blistering of sun-exposed skin, as well as impairment of liver function.

Across human populations, the frequency of the T allele at rs2294915 was associated with a high expression of *PNPLA3* in liver cancer, which is much lower in African populations (16%) compared to South Asians (25%), Europeans (25%), East Asians (37%) and America (49%). In contrast, the frequency of the C allele at rs2296019 was considerably higher in African (84%), European (80%), South Asian (76%), East Asian (64%), and American (56%) populations.

In patients with liver cancer who have a history of alcohol addiction to an amount of  $\geq 3$  drinks per day can increase the risk of liver cancer by 16% in the general population; diabetics and people

with central obesity also increase the risk of liver cancer by 2 times (McGlynn et al., 2021). The diagnosis of liver cancer in patients often involves using serological testing in conjunction with imaging techniques, which is considered the established approach for identifying liver carcinoma. Nevertheless, the diagnostic sensitivity of the often-used serological test, specifically designed to detect alpha-fetoprotein (AFP), is at around 60%. Imaging modalities, including magnetic resonance imaging (MRI), computed tomography (CT), and ultrasonography (US), exhibit notable levels of sensitivity and specificity in the identification of liver cancer, particularly in individuals afflicted with liver cirrhosis (Huang et al., 2022).

Variant alleles (rs2294915 and rs2896019) are associated with liver cancer. Accordingly, populations from the continents of Africa, America, East Asia, Europe, and South Asia show associated *PNPLA3* expression, resulting in a higher susceptibility to liver cancer. Identifying unique and pathogenic gene variations for a disease is very interesting for research and clinical validation. Identification of these variants can not only provide clues to disease susceptibility or as a diagnostic and prognostic biomarker. (Irham et al., 2020) and but can also be used to find drug target candidates or known as drug repurposing (genomic-driven drug repurposing) (Afief et al., 2022). We hope that the discovery of candidate gene variations for *PNPLA3* can lead to successful clinical validation, which paves the way for this promising diagnostic and prognostic biomarker for liver cancer.

It is important to consider that this study's gene variations found to be pathogenic are still preliminary studies using genomic and bioinformatics databases. However, these results are also important information for future researchers who wish to validate these gene variations in liver cancer patients. Future research is strongly recommended to follow up with additional functional annotations to further prioritize pathogenic gene variations.

## Conclusion

This study identifies genetic variants influencing "liver cancer" reveals the significance of the *PNPLA3* gene in liver tissue, as well as in skin regions exposed to the sun (lower legs), skin regions not exposed to the sun (suprapubic), cultured fibroblasts, and adipose-subcutaneous tissue, all of which contribute to an increased risk of liver cancer development. The two variants, rs2294915, and rs2896019, displayed varying allele frequencies across populations from the continents of Africa, America, East Asia, Europe, and South Asia, affecting *PNPLA3* gene expression. Consequently, these populations are more susceptible to liver cancer due to the associated *PNPLA3* expression. These findings underscore the importance of understanding and considering genomic variations in precision medicine and screening strategies for liver cancer in diverse populations across continents.

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## Hasil review yang pertama dari 2 reviewers pada tanggal 15 November 2023

::: 1st Review :::	
Editorial Comment	
Invited date	Sep 14, 2023
Complete Review Date	Nov 15, 2023
Recommendation	<b>Review after major revision</b>
Comments to the Author	<p>• <b>Reviewer A :</b></p> <p>This paper found new association results about liver cancer using various bioinformatics tools such as GWAS catalog, HaploReg, eQTL analysis, etc. I think this paper needs a lot of improvements to be accepted for publication. Please address following issues as much as you can, and I will make a decision based on the revised manuscript.</p> <ol style="list-style-type: none"><li>1. The English writing is not good. There are lots of awkward or imperfect expressions. Please take a professional English editing to improve the overall quality. I listed some of the examples below.<ul style="list-style-type: none"><li>- page 2, in the abstract, "Liver cancer remains a global burden, ranking fourth in mortality rates globally" : being ranked fourth may be more appropriate.</li><li>- page 4, "One of the websites through a bioinformatics approach that discusses gene variation is Genome-Wide Association Studies (GWAS)" : GWAS should be changed to GWAS catalog, and a period is needed after the sentence.</li></ul></li><li>2. Detailed explanations are not provided.<ul style="list-style-type: none"><li>- Only HaploReg results are provided without any interim process.</li><li>- Aren't the rest of the significant SNPs from the GWAS catalog associated to any genes?</li></ul></li><li>3. Some presentations are not correct.<ul style="list-style-type: none"><li>- In Fig 4, Africa pie chart are not matched with the proportions of the SNPs</li></ul></li><li>4. Interpretation is not enough<ul style="list-style-type: none"><li>- In the last step, they provide allele frequencies for each of continents, but the association between allele frequencies and some statistics for the liver cancer is not provided.</li></ul></li></ol> <p>• <b>Reviewer B :</b></p> <p>This submitted article suggested a pipeline that can draw more meaningful results using several bioinformatics databases and analysis tools. However, the writing of the manuscript is not clear to understand and should be improved more. I listed some of the points to be improved to make it better manuscript.<ul style="list-style-type: none"><li>- In the part of providing statistics of SNPs from multiple continents, information in Table 4 is not matched with the information in Figure 4</li><li>- No interpretation is provided for continent specific allele information to the disease.</li><li>- In the eQTL analysis, I could not understand how the two tissues in the Table 3 are associated to the disease.</li></ul></p>

**Artikel hasil dari revisi yang pertama setelah mendapatkan masukan dan saran oleh 2 reviewers pada tanggal 15 November 2023**

**November 28, 2023**

**Dear Editors,**

We are pleased to submit our revised manuscript titled “A Genomic and Bioinformatic-based Approach to Identify Genetic Variants for Liver Cancer across Multiple Continents” for consideration as an original research article in Genomics & Informatics (GI23067). We are grateful for your encouraging feedback on our manuscript. Enclosed is the revised version, addressing the comments provided by the reviewers. The revised sections of the manuscript are highlighted in yellow. We would like to express our gratitude for the opportunity to refine our manuscript and hope these revisions meet your expectations. Your review and assistance are invaluable, and we look forward to your feedback.

**Sincerely yours,**

**Apt. Lalu Muhammad Irham M.Farm Ph.D.**

Faculty of Pharmacy,

Universitas Ahmad Dahlan, Yogyakarta, Indonesia

Jl. Prof. DR. Soepomo SH, Warungboto,

Kec. Umbulharjo, Kota Yogyakarta, Daerah Istimewa Yogyakarta

## Recommendation Reviewer 1:

This paper found new association results about liver cancer using various bioinformatics tools such as GWAS catalog, HaploReg, eQTL analysis, etc. I think this paper needs a lot of improvements to be accepted for publication. Please address following issues as much as you can, and I will make a decision based on the revised manuscript.

**Answer:** We are grateful for the detailed review and constructive feedback. Efforts have been made to address the highlighted concerns.

**Q1:** The english writing is not good. There are lots of awkward or imperfect expressions. Please take a professional english editing to improve the overall quality I listed some of the examples below.

- page2, in the abstract, "Liver cancer remains a global burden, ranking fourth in mortality rates globally" : being ranked fourth may be more appropriate.
- page4, "One of the websites through a bioinformatics approach that discusses gene variation is Genome-Wide Association Studies (GWAS)" : GWAS should be changed to GWAS catalog. and a period is needed after the sentence.

**A1: Thank you for pointing out the areas needing improvement. We have thoroughly revised the language and corrected the specified sentences on [Page 2 and 4, Lines 38 and 79-80].**

*"Liver cancer is the fourth leading cause of death worldwide".*

*"One of the websites through a bioinformatics approach that documents genetic variants is GWAS catalog".*

**Q2:** Detailed explanations are not provided.

- Only HaploReg results are provided without any interim process.
- Aren't the rest of the significant SNPs from the GWAS catalog associated to any genes?

**A2: Thank you to the reviewers who have provided us with input. We have corrected it according to the feedback in [Page 4, lines 115-116 and 119-123]. The sentences are revised as below :**

*"Subsequently, HaploReg version 4.1 was utilized, and a p-value  $<10^{-8}$  was applied based on the number of SNPs obtained"*

*"Through our integrative bioinformatics approach, two variants with a missense mutation (rs2294915, rs2896019) that encoded the PNPLA3 genes were prioritized as the biological risk SNPs for Liver Cancer. Primary liver cancer, also known as hepatocellular carcinoma, is a pathological condition characterised by the development of malignant cells within the hepatic tissues"*

**Q3:** Some presentations are not correct.

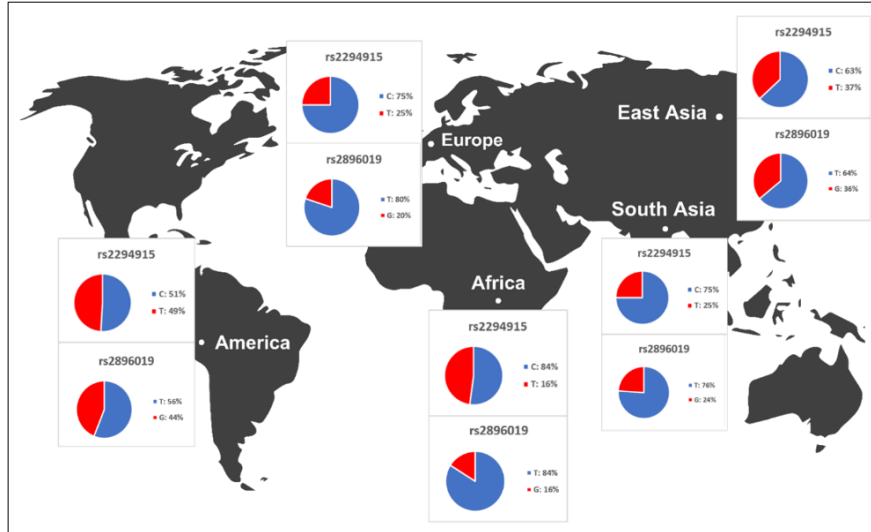
- In Fig 4, Africa pie chart are not matched with the proportions of the SNPs

**A3: We acknowledge this discrepancy and have revised the relevant sections, specifically Page 7, lines 179-186, to ensure accuracy in the tables and figures:**

**Table 4.** Analysis of allele frequencies for the *PNPLA3* gene from variant annotation (SNPnexus)

SNP	Gene	Location	Alle		Allele Frequency (N)				
			Ref	Eff*	AFR	AMR	EAS	EUR	SAS
rs2294915	<i>PNPLA3</i>	Missense	C	T	T: 0.163 (215)	T: 0.490 (340)	T: 0.365 (368)	T: 0.252 (254)	T: 0.246 (241)
rs2896019	<i>PNPLA3</i>	Missense	T	G	G: 0.158 (209)	G: 0.438 (304)	G: 0.364 (367)	G: 0.199 (200)	G: 0.236 (231)

Note : *PNPLA3*, patatin like phospholipase 3; SNP, single nucleotide polymorphism; Ref, Reference; Alt, alternate; AFR, Africa; AMR, America; EAS, East Asia; EUR, Europe; SAS, Southeast Asia.



**Figure 4.** The results of the distribution of allele frequencies affecting *PNPLA3* across various populations. EUR, Europe; AFR, Africa; SAS, South Asia; EAS, East Asia.

Q4: Interpretation is not enough

- In the last step, they provide allele frequencies for each of continents, but the association between allele frequencies and some statistics for the liver cancer is not provided.

**A4: Thank you for highlighting this oversight. We have enhanced our interpretation and revised the related sections, particularly on Pages 7 and 8, lines 169-176 and 195-199, to address this issue.**

*“The results of the research we have done, we have identified variants associated with liver cancer gene expression and carried out allele frequency analysis in various populations. As shown in Table 4, allele variant frequencies were evaluated in multiple people from Europe, America, East Asia, South Asia, and Africa. Allele frequencies across populations varied for each SNP, as depicted in Figure 4. Table 4 and Figure 4 show the gene expression levels at higher frequencies of the rs2294915 related allele (C) and the rs2896019 related allele (T). At the population frequency of the rs2294915 (C) allele, populations in Europe and South Asia were expressed at much higher levels than America, Africa, and East Asia”.*

*“The allele frequencies of the T and G alleles “rs2294915” and “rs2896019” in African populations were expressed at much lower levels compared to American, European, and Southeast Asian people. Overall, the allele frequencies of the “rs2294915” and “rs2896019” variant alleles suggest a contribution to the prevalence of the variants for gene expression of *PNPLA3*”.*

## Recommendation Reviewer 2:

This submitted article suggested a pipeline that can draw more meaningful results using several bioinformatics databases and analysis tools. However, the writing of the manuscript is not clear to understand and should be improved more. I listed some of points to be improved to make it better manuscript.

**Answer: We appreciate the reviewer's detailed feedback and have made revisions to enhance the clarity of our manuscript.**

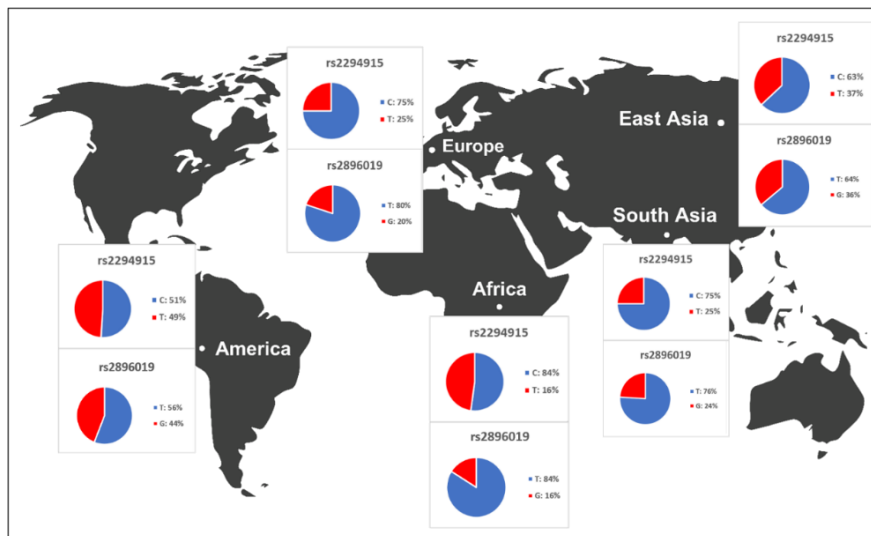
**Q1:** In the part of providing statistics of SNPs from multiple continents, information in Table 4 is not matched with the information in Figure 4

**A1: Thank you for highlighting this issue. We have revised the relevant sections to ensure consistency between Table 4 and Figure 4 [Page 7, lines 179-186], as below:**

**Table 4.** Analysis of allele frequencies for the *PNPLA3* gene from variant annotation (SNPnexus)

SNP	Gene	Location	Alle		Allele Frequency (N)				
			Ref	Eff*	AFR	AMR	EAS	EUR	SAS
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rs2896019	<i>PNPLA3</i>	Missense	T	G	G: 0.158 (209)	G: 0.438 (304)	G: 0.364 (367)	G: 0.199 (200)	G: 0.236 (231)

Note : *PNPLA3*, patatin like phospholipase 3; SNP, single nucleotide polymorphism; Ref, Reference; Alt, slternate; AFR, Africa; AMR, America; EAS, East Asia; EUR, Europe; SAS, Southeast Asia.



**Figure 4.** The results of the distribution of allele frequencies affecting *PNPLA3* across various populations. EUR, Europe; AFR, Africa; SAS, South Asia; EAS, East Asia.

**Q2:** No interpretation is provided for continent specific allele information to the disease

**A2: We acknowledge the need for clearer interpretation and have revised the relevant sections to better articulate the association between allele frequencies and liver cancer across different continents [Page 7 and 8. Lines 169-176 and 195-199]. The sentences are revised as below:**

*“The results of the research we have done, we have identified variants associated with liver cancer gene expression and carried out allele frequency analysis in various populations. As shown in Table 4, allele variant frequencies were evaluated in multiple people from Europe, America, East Asia, South Asia, and Africa. Allele frequencies across populations varied for each SNP, as depicted in Figure 4. Table 4 and Figure 4 show the*

*gene expression levels at higher frequencies of the rs2294915 related allele (C) and the rs2896019 related allele (T). At the population frequency of the rs2294915 (C) allele, populations in Europe and South Asia were expressed at much higher levels than America, Africa, and East Asia”.*

*“The allele frequencies of the T and G alleles “rs2294915” and “rs2896019” in African populations were expressed at much lower levels compared to American, European, and Southeast Asian people. Overall, the allele frequencies of the “rs2294915” and “rs2896019” variant alleles suggest a contribution to the prevalence of the variants for gene expression of PNPLA3”.*

**Q3:** In the eQTL analysis, I could not understand how the two tissues in the Table 3 are associated to the disease.

**A3:** **Thank you for your valuable feedback. We have addressed your concerns in our revision, specifically in the eQTL analysis section on Pages 5 and 6, Lines 148-153 and 159-163. The relevant sentences have been revised for clarity as follows:**

*“The result in a correlation between the Gene Expression of PNPLA3 and eQTL, to identify eQTLs associated with liver cancer gene expression, the GTEx database was used. We have identified minor alleles related to liver cancer, as presented in Table 3. Uniquely, several types of SNPs we found have high expression in skin tissue, namely rs2294915 and rs2896019. The CC type genotypes of rs2294915 and rs2896019 were associated with higher expression in suprapubic and underarm skin compared with the CT and TT type genotypes (Figure 3)”.*

*“The research results show that the genomic database can identify gene variations with the highest potential in the pathogenesis of liver cancer. Liver cancer is characterized by eyes and skin that appear yellow (Fitrianti et al., 2022). According to Nessa et. al. (2017), the severity of the liver can be gauged by the diminishing quality of liver function. The quality of liver function can be assessed from total bilirubin levels, serum albumin, and PT (partial thromboplastin time)”.*



1 **A Genomic and Bioinformatic-based Approach to Identify Genetic Variants**  
2 **for Liver Cancer across Multiple Continents**

3  
4 **Muhammad Ma'ruf<sup>1</sup>, Lalu Muhammad Irham<sup>1\*</sup>, Wirawan Adikusuma<sup>2</sup>, Made Ary Sarasmita<sup>3,4</sup>**  
5 **Sabiah Khairi<sup>5</sup>, Barkah Djaka Purwanto<sup>6,7</sup>, Rockie Chong<sup>8</sup>, Maulida Mazaya<sup>9</sup>, Lalu Muhammad**  
6 **Harmain Siswanto<sup>10</sup>**

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## Abstract

Liver cancer is the fourth leading cause of death worldwide. Well-known risk factors include hepatitis B virus (HBV) and hepatitis C virus (HCV), along with exposure to aflatoxins, excessive alcohol consumption, obesity, and type 2 diabetes.. Genomic variants play a crucial role in mediating liver cancer among these factors. However, specific variants involved in this process remain under-explored. This study utilizes a bioinformatics approach to identify genetic variants associated with liver cancer from across various continents. The single nucleotide polymorphisms (SNPs) associated with liver cancer were retrieved from the Genome-Wide Association Studies (GWAS) catalog. The subsequent prioritization was performed using functional annotation with HaploReg v4.1 and the Ensembl database. Our results indicate that two variants, rs2294915 and rs2896019, encoded by the PNPLA3 gene, were found to be highly expressed in the liver tissue, as well as in the skin, cell-cultured fibroblasts and adipose-subcutaneous tissue, all of which contribute to the risk of liver cancer. We highlight the utility of this population-specific *PNPLA3* genetic variant for genetic association studies and for early prognosis and treatment of liver cancer. This study highlights the potential of integrating genomic databases with bioinformatic analysis to identify genetic variations involved in the pathogenesis of liver cancer. We recommend that future research prioritize the validation of these variations in clinical settings.

**Keywords:** *Genomic Variants, Liver Cancer, Bioinformatics*

## Introduction

Liver cancer, a type of carcinoma, has the highest mortality rate in the world every year (McGlynn et al., 2021). In 2018, there were 841,000 cases of liver cancer, of which the death rate caused by liver cancer reached 782,000 (Bray et al., 2018). The average number of liver cancer cases and associated deaths can be two to three times higher in men in certain parts of the world. The Global Cancer Statistics (GLOBOCAN) in 2020 ranked liver cancer as the third most deadly cancer, accounting for 8.3% of cancer-related deaths. In 2020, liver cancer incidences reached 905,000, with a mortality rate of 830,000 (Sung et al., 2021). In Indonesia, liver cancer is the second most common in men, amounting to 12.4 per 100,000 of the Indonesian population, with an average death rate of 7.6 per 100,000 (Kemenkes RI, 2019).

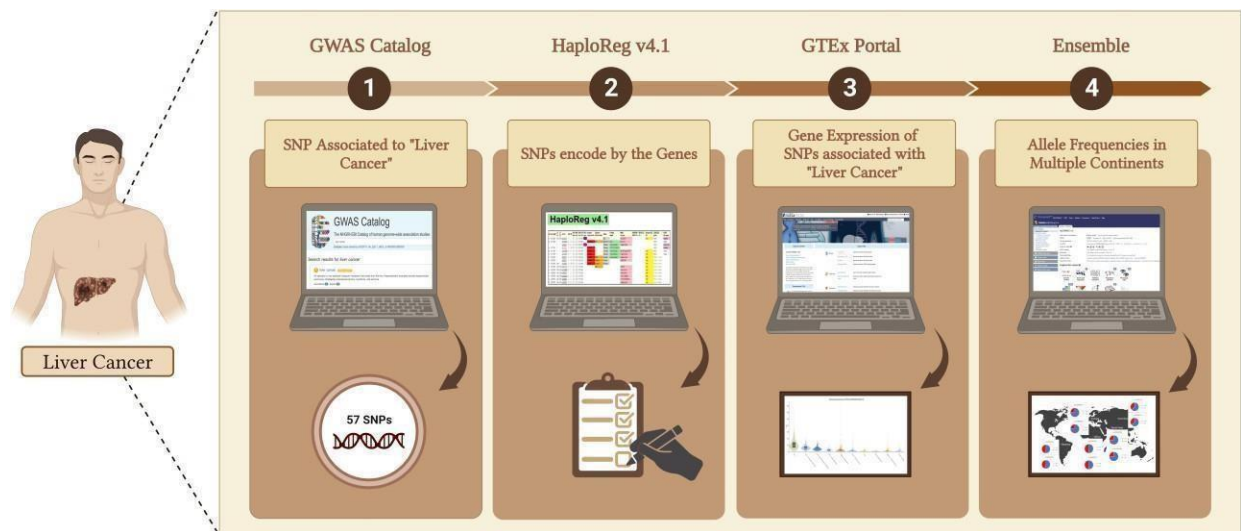
Factors that cause liver cancer include chronic infection with hepatitis B virus (HBV), hepatitis C virus (HCV), the result of contamination with aflatoxin, alcohol consumption, history of obesity, history of type 2 diabetes, and smoking addiction (Bray et al., 2018). According to Villanueva (2019), other risk factors are thought to exacerbate the occurrence of liver cancer, such as an unhealthy lifestyle, geographic conditions, gender, age, family history of the disease, and the severity of damage to the liver. Liver cancer is also found in areas that have cases of hepatitis B. In these areas, liver cancer is prevalent at a young age. This is because some infected with hepatitis B are obtained vertically through the delivery process (Mittal & El-Serag, 2013).

Patients often felt complaints in the form of fatigue, pain, diarrhea, skin abnormalities, and decreased appetite, all of which have affected their quality of life (Waller et al., 2015). Therefore, detecting the presence of disease symptoms in liver cancer can be done by examining deoxyribonucleic acid (DNA). Gene variation can be associated with disease progression and pathogenesis, which includes liver cancer. One of the websites through a bioinformatics

80 approach that documents genetic variation is the GWAS catalog. GWAS catalog is a database  
81 with single nucleotide polymorphism (SNP) search results that has identified several variants  
82 associated with liver fat content, circulating liver enzymes, and the development of Non-  
83 Alcoholic Fatty Liver Disease (NAFLD) as well as genetic markers used in predicting a disease  
84 disorder (Wang et al., 2021).

85 Genetic identification in humans aims to identify inherited genetic risk factors for liver cancer.  
86 This study uses the GWAS catalog database to map genes from genetic variations across several  
87 populations that play an essential role in the pathogenesis of liver cancer. The most significant  
88 gene variations based on their function in protein changes will be further verified.  
89

## 90 Methods



91  
92 **Figure 1.** Analysis methodology for integrated bioinformatic, database and genomic analysis of  
93 genetic variation that affect liver cancer. The figure was created with BioRender.com under  
94 agreement number “FM2500073C”  
95

96 In this study, we adopted the method used by Ma’ruf et al (2023) and Puspitaningrum et  
97 al (2022), as depicted in Figure 1. Liver cancer-associated SNPs were obtained from the GWAS  
98 Catalog database (<http://www.ebi.ac.uk/gwas>; accessed on 15-02-2023). Subsequently, we  
99 performed further analysis using HaploReg (version 4.1) applying a p-value  $< 10^{-8}$  to account  
100 for multiple tests in the GWAS catalog. This threshold is commonly used to identify  
101 associations between common genetic variants and traits with adjacent gene expression (Chen et  
102 al., 2021). Furthermore, to evaluate the relationships between various genetic variants and gene  
103 expression profiles, we utilized e-QTL analysis with data sourced from the GTEx Portal database  
104 (<http://www.gtexportal.org/home/>; accessed on 16-02-2023), considering gene expression across  
105 various tissues in humans. Additionally, we confirmed the identified variants using the Ensembl  
106 Genome Browser (<https://www.ensembl.org/index.html>; accessed on 17-02-2023). Our study  
107 considered allele frequencies in populations from Europe, Africa, America, East Asia, and

108 Southeast Asia. To comprehend the functionalities of different gene variants, we performed  
109 evaluations using the SNP nexus database (<https://www.snp-nexus.org>; accessed on 20-02-2023).  
110

## 111 Results and Discussion

### 112 1. Identification of Genomic Variants of Liver Cancer

113 This study identified SNPs associated with liver cancer from the GWAS catalog. Among  
114 these SNPs, 29 of them were further confirmed through SNP duplication, as shown in Table 1.  
115 Subsequently, HaploReg version 4.1 was utilized, and a p-value  $<10^{-8}$  was applied based on the  
116 number of SNPs obtained. Based on the findings presented in Table 2, we found the risk of two  
117 genes for “Liver Cancer” disease. This study analyzed tissue expression affecting liver cancer,  
118 focusing on the missense variant *PNPLA3*.

119 Through our integrative bioinformatics approach, two variants with a missense mutation (rs  
120 rs2294915, rs2896019) that encoded the *PNPLA3* genes were prioritized as the biological risk  
121 SNPs for Liver Cancer. Primary liver cancer, also known as hepatocellular carcinoma, is a  
122 pathological condition characterised by the development of malignant cells within the hepatic  
123 tissues. The development of cancer in extraneous anatomical sites that then metastasizes to the  
124 liver does not constitute primary liver cancer. Primary liver cancer encompasses many kinds,  
125 including hepatocellular carcinoma (HCC), intrahepatic cholangiocarcinoma (iCCA), and less  
126 frequent varieties such as mixed hepatocellular cholangiocarcinoma (HCC-CCA), fibrolamellar  
127 HCC (FLC), and the paediatric neoplasm hepatoblastoma (Wage et al., 2021).  
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**Table 1.** SNPs from the GWAS catalog with p-value  $<10^{-8}$

No.	Variation and risk allele	p-value
1	rs2856723	$3 \times 10^{-43}$
2	rs34675408	$1 \times 10^{-32}$
3	rs9272105	$5 \times 10^{-22}$
4	rs913493	$5 \times 10^{-20}$
5	rs2294915	$2 \times 10^{-19}$
6	rs17401966	$2 \times 10^{-18}$
7	rs3096380	$1 \times 10^{-17}$
8	rs9275319	$3 \times 10^{-17}$
9	rs584368	$2 \times 10^{-14}$
10	rs2596542	$4 \times 10^{-13}$
11	rs1110446	$9 \times 10^{-13}$
12	rs58489806	$3 \times 10^{-12}$
13	rs6078460	$2 \times 10^{-11}$
14	rs2523961	$6 \times 10^{-11}$
15	rs7574865	$2 \times 10^{-10}$
16	rs1110446	$3 \times 10^{-10}$
17	rs455804	$5 \times 10^{-10}$
18	rs58542926	$6 \times 10^{-10}$
19	rs2523961	$6 \times 10^{-10}$
20	rs8107030	$8 \times 10^{-10}$
21	rs10272859	$9 \times 10^{-10}$
22	rs190121281	$4 \times 10^{-9}$
23	rs9275572	$6 \times 10^{-9}$
24	rs2242652	$6 \times 10^{-9}$
25	rs188273166	$1 \times 10^{-8}$
26	rs708113	$1 \times 10^{-8}$
27	rs2896019	$2 \times 10^{-8}$
28	rs17047200	$3 \times 10^{-8}$
29	rs541860626	$5 \times 10^{-8}$

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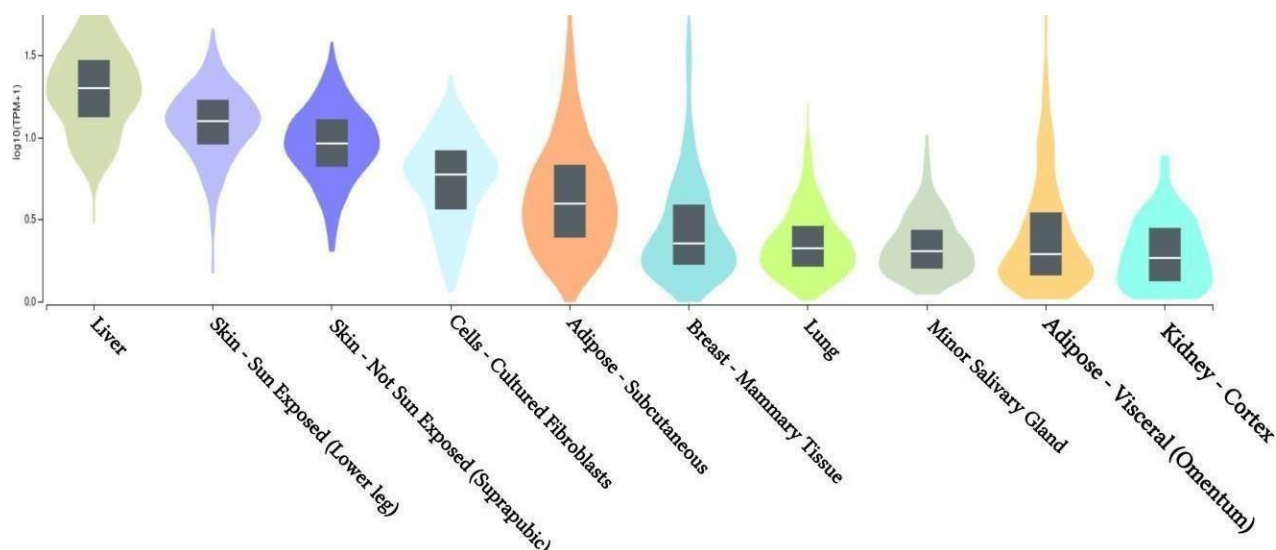
**Table 2.** Variants and risk alleles of liver cancer encoding prioritized SNPs

Variation and risk alleles	Variants near risk allele ( $r^2 > 0.8$ )	$p$ -value	GENCODE	Type of allele
rs2294915	rs738409	$2 \times 10^{-19}$	<i>PNPLA3</i>	missense
rs2896019	rs3761472	$2 \times 10^8$	<i>PNPLA3</i>	missense

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## 2. Gene expression of *PNPLA3* across 10 human tissues

The results of *PNPLA3* gene expression across 10 human tissues comprise the most apparent functional consequences of genetic variation. Liver, sun-exposed skin (lower legs), non-sun-exposed skin (suprapubic), and adipose subcutaneous fibroblasts and cell cultures showed the highest *PNPLA3* gene expression in the 10 human tissues analyzed from GTEx (Figure 2). In addition, we have found that the SNP IDs rs2294915 and rs2896019 have similar gene expression variations in Sun-Exposed skin (lower legs). The exciting thing about these findings is that liver cancer patients often experience complaints that their skin appears yellow. Further results showed that the *PNPLA3* gene has high expression in suprapubic and underarm skin.



143

144 **Figure 2.** *PNPLA3* gene expression associated with liver cancer across human tissues based on  
145 GTEx Portal analysis  
146

## 147 3. Correlation between Gene Expression of *PNPLA3* and eQTL

148 The result in a correlation between the Gene Expression of *PNPLA3* and eQTL, to identify  
149 eQTLs associated with liver cancer gene expression, the GTEx database was used. We have  
150 identified minor alleles related to liver cancer, as presented in Table 3. Uniquely, several types of  
151 SNPs we found have high expression in skin tissue, namely rs2294915 and rs2896019. The CC  
152 type genotypes of rs2294915 and rs2896019 were associated with higher expression in  
153 suprapubic and underarm skin compared with the CT and TT type genotypes (Figure 3).

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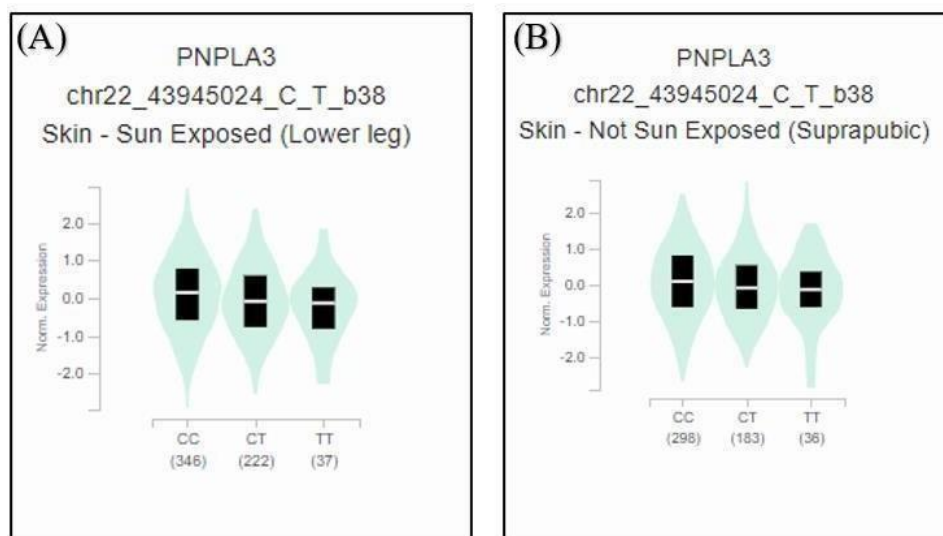
**Table 3.** Results of e-QTL in liver cancer from the GTEx portal database

SNP	Gencode ID (ENSG000000-)	Gene Symbol	p-value	Effect Size	Tissue	Expression Level
rs2294915	100344.10	<i>PNPLA3</i>	$2.8 \times 10^{-8}$	-0.15	Skin - Sun Exposed (Lower leg)	CC>CT>TT
	100344.10	<i>PNPLA3</i>	$5 \times 10^{-8}$	-0.50	Skin - Not Sun Exposed (Suprapubic)	CC>CT>TT
rs2896019	100344.10	<i>PNPLA3</i>	$6.7 \times 10^{-11}$	-0.19	Skin - Sun Exposed (Lower leg)	CC>CT>TT
	100344.10	<i>PNPLA3</i>	$2 \times 10^{-9}$	-0.22	Skin - Not Sun Exposed (Suprapubic)	CC>CT>TT

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Source: Expression Quantitative Trait Loci (eQTL) obtained from the GTEx Portal.  
Note : *PNPLA3*, patatin like phospholipase 3; SNP, single nucleotide polymorphism; eQTL, expression quantitative trait loci; SNP, single nucleotide polymorphism.

159 The research results show that the genomic database can identify gene variations with the  
160 most potential in the pathogenesis of liver cancer. Liver cancer is characterized by eyes and skin  
161 that appear yellow (Fitrianti et al., 2022). According to Nessa et al (2017), the severity of the  
162 liver can be seen from the decreasing quality of the liver. The quality of the liver can be assessed  
163 from total bilirubin levels, serum albumin, and PT (partial thromboplastin time).



164  
165 **Figure 3.** Patatin like phospholipase domain containing protein 3 (*PNPLA3*) gene expression for  
166 each genotype of the single nucleotide polymorphism (SNP): (A) rs2294915 and (B) rs2896019.  
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#### 4. Allele frequencies of candidate variants in populations in different continents

169 The results of the research we have done, we have identified variants associated with liver  
170 cancer gene expression and carried out allele frequency analysis in various populations. As  
171 shown in Table 4, allele variant frequencies were evaluated in multiple people from Europe,  
172 America, East Asia, South Asia, and Africa. Allele frequencies across populations varied for each  
173 SNP, as depicted in Figure 4. Table 4 and Figure 4 show the gene expression levels at higher  
174 frequencies of the rs2294915 related allele (C) and the rs2896019 related allele (T). At the  
175 population frequency of the rs2294915 (C) allele, populations in Europe and South Asia were  
176 expressed at much higher levels than America, Africa, and East Asia.

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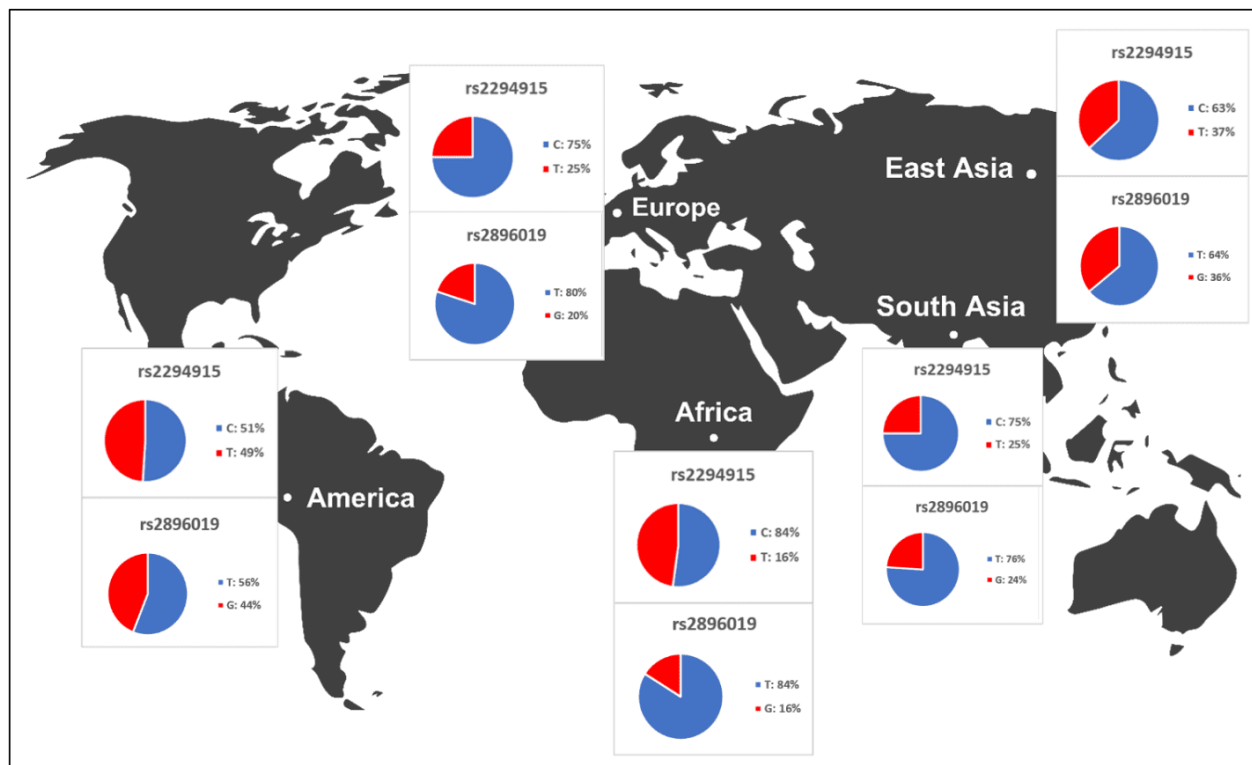
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**Table 4.** Analysis of allele frequencies for the *PNPLA3* gene from variant annotation (SNPnexus)

SNP	Gene	Location	Allele		Allele Frequency (N)				
			Ref	Eff*	AFR	AMR	EAS	EUR	
rs2294915	<i>PNPLA3</i>	Missense	C	T	T: 0.163 (215)	T: 0.490 (340)	T: 0.365 (368)	T: 0.252 (254)	
rs2896019	<i>PNPLA3</i>	Missense	T	G	G: 0.158 (209)	G: 0.438 (304)	G: 0.364 (367)	G: 0.199 (200)	

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Note : *PNPLA3*, patatin like phospholipase domain containing protein 3; SNP, single nucleotide polymorphism; Ref, Reference; Alt, Alternate; AFR, Africa; AMR, America; EAS, East Asia; EUR, Europe; SAS, Southeast Asia.



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**Figure 4.** The results of the distribution of allele frequencies affecting *PNPLA3* across various populations. EUR, Europe; AFR, Africa; SAS, South Asia; EAS, East Asia.

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Based on this finding, rs2294915 and rs2896019 are potentially related to the susceptibility to “liver cancer with effect size the highest score of -0.50 can be interpreted on the skin not exposed to sunlight (suprapubic).” According to Poggiali & Vercelli (2023), this condition is characterized by a disruption in the heme biosynthesis pathway due to reduced hepatic uroporphyrinogen decarboxylase (UROD) activity. The consequence of this phenomenon is the buildup of light-sensitive by-products, such as uroporphyrinogen, leading to the development of fragility and blistering of sun-exposed skin, as well as impairment of liver function.

The allele frequencies of the T and G alleles “rs2294915” and “rs2896019” in African populations were expressed at much lower levels compared to American, European, and Southeast Asian people. Overall, the allele frequencies of the “rs2294915” and “rs2896019” variant alleles suggest a contribution to the prevalence of the variants for gene expression of *PNPLA3*.

Across human populations, the frequency of the T allele at rs2294915 was associated with a high expression of *PNPLA3* in liver cancer, which is much lower in African populations (16%)



202 compared to South Asians (25%), Europeans (25%), East Asians (37%) and America (49%). In  
203 contrast, the frequency of the C allele at rs2296019 was considerably higher in African (84%),  
204 European (80%), South Asian (76%), East Asian (64%), and American (56%) populations.

205 Patients with liver cancer who have a history of alcohol addiction to an amount of  $\geq 3$  drinks  
206 per day can increase the risk of liver cancer by 16% in the general population; diabetics and  
207 people with central obesity also increase the risk of liver cancer by 2 times (McGlynn et al.,  
208 2021). The diagnosis of liver cancer in patients often involves using serological testing in  
209 conjunction with imaging techniques, which is considered the established approach for  
210 identifying liver carcinoma. Nevertheless, the sensitivity of the often-used serological test,  
211 specifically designed to detect alpha-fetoprotein (AFP), is at around 60%. Imaging modalities,  
212 including magnetic resonance imaging (MRI), computed tomography (CT), and ultrasonography  
213 (US), exhibit notable levels of sensitivity and specificity in the identification of liver cancer,  
214 particularly in individuals afflicted with liver cirrhosis (Huang et al., 2022).

215 Variant alleles (rs2294915 and rs2896019) are associated with liver cancer. Accordingly,  
216 populations from the continents of Africa, America, East Asia, Europe, and South Asia show  
217 associated *PNPLA3* expression, resulting in a higher susceptibility to liver cancer. Identifying  
218 unique and pathogenic gene variations for a disease is very interesting for both research and  
219 clinical validation. These variants not only offer insights into disease susceptibility but also  
220 serve as potential diagnostic and prognostic biomarkers (Irham et al., 2020). Additionally, they  
221 can facilitate the identification of drug target candidates, a concept known as genomic-driven  
222 drug repurposing (Afief et al., 2022). We anticipate that the discovery of candidate gene  
223 variations in *PNPLA3* will pave the way for successful clinical validation, potentially  
224 establishing this as a promising diagnostic and prognostic biomarker for liver cancer.

225 It is important to acknowledge that the genetic variants identified in this study as potentially  
226 pathogenic are based on preliminary investigations using genomic and bioinformatics databases.  
227 However, these findings offer crucial insights for future researchers intending to validate these  
228 genetic variants in liver cancer patients. We strongly recommend future research to incorporate  
229 further functional annotations, which would further aid in prioritizing these pathogenic genetic  
230 variants.

231

## 232 **Conclusion**

233 This study identifies genetic variants influencing “liver cancer” and reveals the significance of  
234 the *PNPLA3* gene in liver tissue, as well as in skin regions exposed to the sun (lower legs), skin  
235 regions not exposed to the sun (suprapubic), cultured fibroblasts, and adipose-subcutaneous  
236 tissue. These findings collectively contribute to an increased risk of liver cancer development.  
237 The observed variations in allele frequencies of the two identified variants, rs2294915, and  
238 rs2896019, across populations from Africa, America, East Asia, Europe, and South Asia,  
239 significantly impact *PNPLA3* gene expression. Consequently, these population groups exhibit  
240 varying susceptibilities to liver cancer based on associated *PNPLA3* expression levels. These  
241 discoveries highlight the critical relevance of understanding genomic variations in precision  
242 medicine and designing screening strategies for liver cancer across diverse populations on  
243 different continents.

244

245

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**Pada tanggal 28 November 2023 artikel mendapatkan revisian yang kedua oleh reviewers.**

::: 2nd Review :::	
<b>Editorial Comment</b>	
Invited date	Nov 28, 2023
Complete Review Date	Nov 28, 2023
Recommendation	<b>Accept after minor revision</b>
Comments to the Author	<ul style="list-style-type: none"><li>- Although the reviewers pointed out that Figure 4 was not correct, they did not fully correct the figure, for example, rs2294915 in Africa, the pie chart seems still not correct. Please correct it.</li><li>- The interpretation using the allele information from the continents are still not clear. There are not enough information about association between allele frequency and the prevalence of the liver cancer of each continent. Please add some more contents based on the actual prevalence information.</li></ul>

**Artikel hasil dari revisi yang kedua setelah mendapatkan masukan dan saran oleh 2 reviewers pada tanggal 28 November 2023**

**November 29, 2023**

**Dear Editors,**

We are pleased to submit our revised manuscript titled “A Genomic and Bioinformatic-based Approach to Identify Genetic Variants for Liver Cancer across Multiple Continents” for consideration as an original research article in Genomics & Informatics (GI23067). We are grateful for your encouraging feedback on our manuscript. Enclosed is the revised version, addressing the comments provided by the reviewers. The revised sections of the manuscript are highlighted in yellow. We would like to express our gratitude for the opportunity to refine our manuscript and hope these revisions meet your expectations. Your review and assistance are invaluable, and we look forward to your feedback.

**Sincerely yours,**

**Apt. Lalu Muhammad Irham M.Farm Ph.D.**

Faculty of Pharmacy,

Universitas Ahmad Dahlan, Yogyakarta, Indonesia

Jl. Prof. DR. Soepomo SH, Warungboto,

Kec. Umbulharjo, Kota Yogyakarta, Daerah Istimewa Yogyakarta

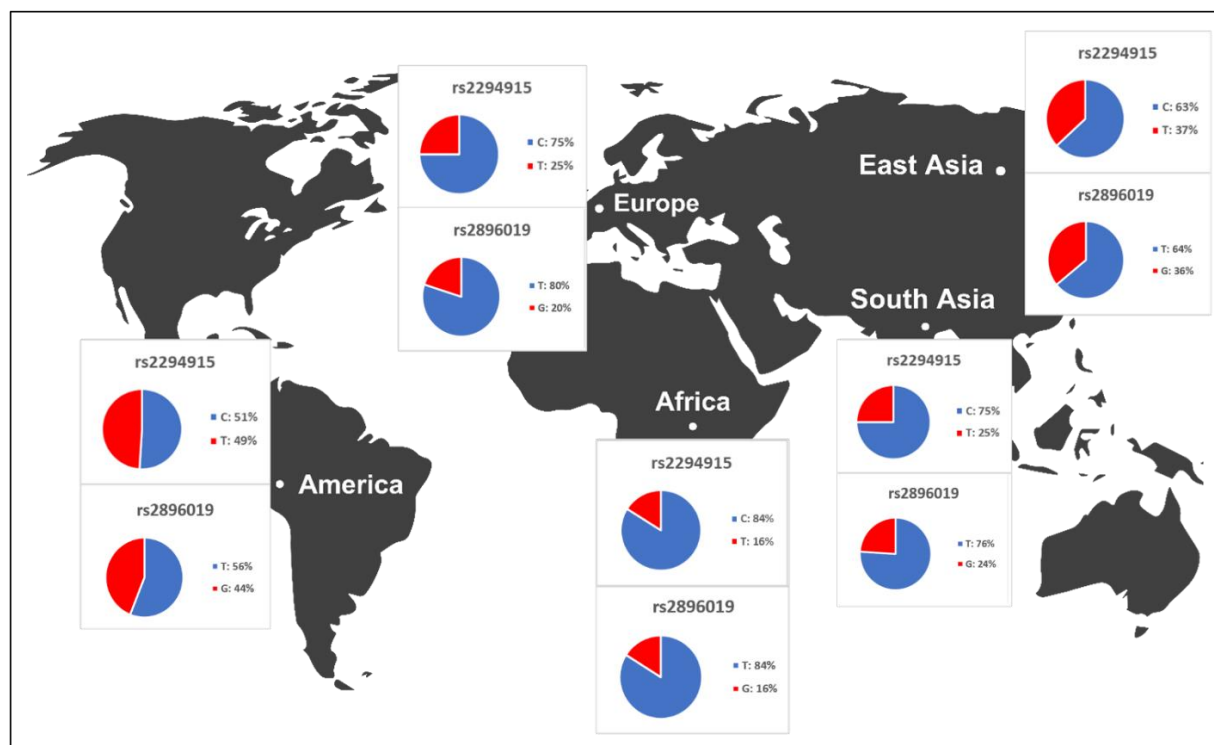
## Recommendation Reviewer 2:

Accept after minor revision

**Answer:** We are grateful for the detailed review and constructive feedback. Efforts have been made to address the highlighted concerns.

**Q1:** Although the reviewers pointed out that Figure 4 was not correct, they did not fully correct the figure, for example, rs2294915 in Africa, the pie chart seems still not correct. Please correct it.

**A1:** We acknowledge this discrepancy and have revised the relevant sections, specifically Page 7, lines 184, to ensure accuracy in the figures:



**Figure 4.** The results of the distribution of allele frequencies affecting *PNPLA3* across various populations. EUR, Europe; AFR, Africa; SAS, South Asia; EAS, East Asia.

**Q2:** The interpretation using the allele information from the continents are still not clear. There are not enough information about association between allele frequency and the prevalence of the liver cancer of each continent. Please add some more contents based on the actual prevalence information.

**A2:** Thank you for highlighting this oversight. We have enhanced our interpretation and revised the association between allele frequency and the prevalence of the liver cancer of each continent, in each of the following continents:

### Abstract [Page 2, lines 51-54]

“We further obtained that these two SNPs (*rs2294915* and *rs2896019*) were positively correlated with the prevalence rate. Positive association of prevalence rates were underlined more frequent in East Asian and African population.”

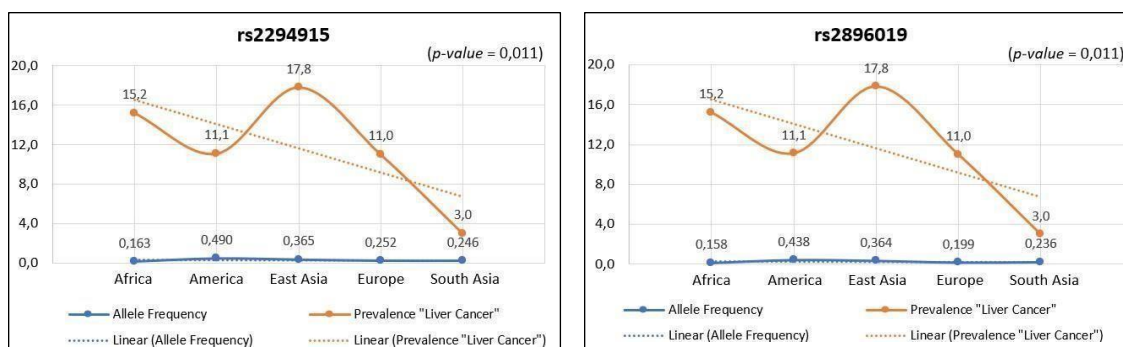


## Methods [Page 3, Lines 116-121]

“Furthermore, epidemiological and genomic data of the prevalence of liver cancer rates were obtained from Li et al (2022). The prevalence rates and allele frequencies of the variants in multiple continents were evaluated using IBM SPSS Statistics 25.0 with the Pearson Correlation test. After the procedure was evaluated, the values of p-value 95% CI were obtained. All plots were created using line charts. A p-value less than 0.05 ( $P \leq 0.05$ ) was considered as statistically significant in current study”

## Result and Discussion [Page 8, Lines 216-227]

“Next, the association between allele frequency and the prevalence of the liver cancer of each continent was evaluated. Data on the prevalence of liver cancer in the continents were obtained from Li et al (2022) (Li et al., 2023). Herein, two SNPs (*rs2294915* and *rs2896019*) were positively correlated with the prevalence rate of liver cancer in multiple continents (Africa, America, East Asia, Europe, South Asia) based on the Pearson's correlation analysis ( $P$ -value < 0.011) (Figure 5). Populations with higher frequencies of variant alleles of these polymorphisms are thought to have a higher prevalence. We highlighted that these two variants (*rs2294915* and *rs2896019*) were more frequent in the East Asian and Africa, which performed the higher aggressiveness of liver cancer in East Asian and African compared to America, Europe and South Asia. This study might give an insight that East Asian and African with carriers variants *rs2294915* and *rs2896019* might be more highly susceptible to suffer the liver cancer”.



**Figure 5.** The association between allele frequency and the prevalence of liver cancer of each continent

## Conclusion [Page 10, Lines 266-271]

“Our study also demonstrated that these two SNPs (*rs2294915* and *rs2896019*) were positively correlated with the prevalence rate. Positive association of prevalence rates were underlined more frequent in East Asian and African population. The higher frequency of the variants allele of these polymorphisms in population, the higher the estimated prevalence rates. The variants investigated in this study were likely to predispose to liver cancer and could play a role in its progression and aggressiveness”

1 **A Genomic and Bioinformatic-based Approach to Identify Genetic Variants**  
2 **for Liver Cancer across Multiple Continents**

3  
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## Abstract

Liver cancer is the fourth leading cause of death worldwide. Well-known risk factors include hepatitis B virus (HBV) and hepatitis C virus (HCV), along with exposure to aflatoxins, excessive alcohol consumption, obesity, and type 2 diabetes. Genomic variants play a crucial role in mediating liver cancer among these factors. However, specific variants involved in this process remain under-explored. This study utilizes a bioinformatics approach to identify genetic variants associated with liver cancer from across various continents. The single nucleotide polymorphisms (SNPs) associated with liver cancer were retrieved from the Genome-Wide Association Studies (GWAS) catalog. The subsequent prioritization was performed using functional annotation with HaploReg v4.1 and the Ensembl database. The prevalence and allele frequencies of each variants were evaluated by using pearson correlation. Our results indicate that two variants, rs2294915 and rs2896019, encoded by the *PNPLA3* gene, were found to be highly expressed in the liver tissue, as well as in the skin, cell-cultured fibroblasts and adipose-subcutaneous tissue, all of which contribute to the risk of liver cancer. We further obtained that these two SNPs (rs2294915 and rs2896019) were positively correlated with the prevalence rate. Positive association of prevalence rates were underlined more frequent in East Asian and African population. We highlight the utility of this population-specific *PNPLA3* genetic variant for genetic association studies and for early prognosis and treatment of liver cancer. This study highlights the potential of integrating genomic databases with bioinformatic analysis to identify genetic variations involved in the pathogenesis of liver cancer. The genetic variants investigated in this study were likely to predispose to liver cancer and could affect its progression and aggressiveness. We recommend that future research prioritize the validation of these variations in clinical settings.

**Keywords:** *Genomic Variants, Liver Cancer, Bioinformatics*

## Introduction

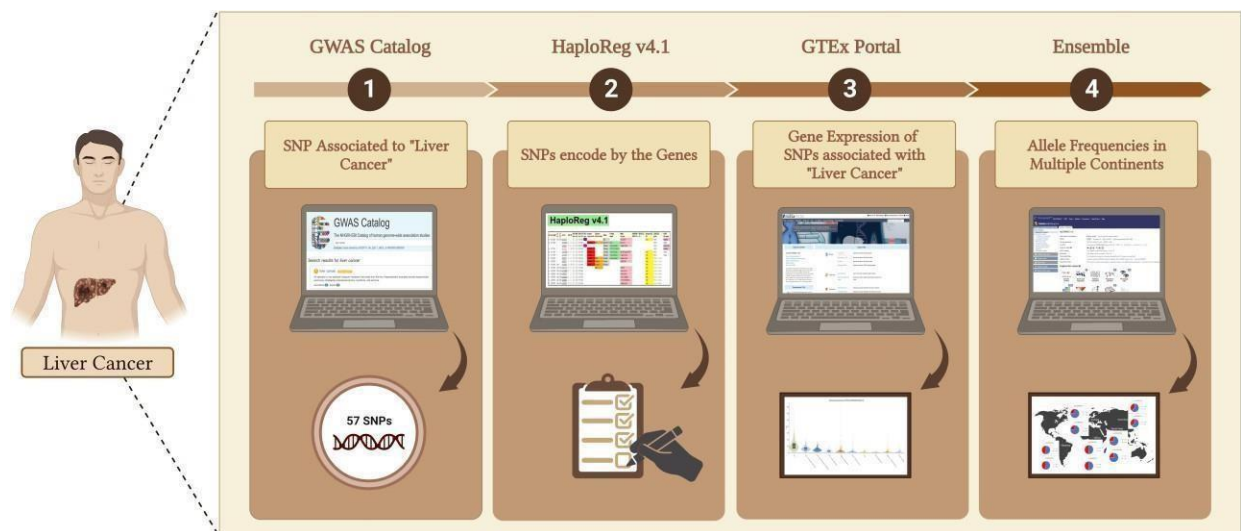
Liver cancer, a type of carcinoma, has the highest mortality rate in the world every year (McGlynn et al., 2021). In 2018, there were 841,000 cases of liver cancer, of which the death rate caused by liver cancer reached 782,000 (Bray et al., 2018). The average number of liver cancer cases and associated deaths can be two to three times higher in men in certain parts of the world. The Global Cancer Statistics (GLOBOCAN) in 2020 ranked liver cancer as the third most deadly cancer, accounting for 8.3% of cancer-related deaths. In 2020, liver cancer incidences reached 905,000, with a mortality rate of 830,000 (Sung et al., 2021). In Indonesia, liver cancer is the second most common in men, amounting to 12.4 per 100,000 of the Indonesian population, with an average death rate of 7.6 per 100,000 (Kemenkes RI, 2019).

Factors that cause liver cancer include chronic infection with hepatitis B virus (HBV), hepatitis C virus (HCV), the result of contamination with aflatoxin, alcohol consumption, history of obesity, history of type 2 diabetes, and smoking addiction (Bray et al., 2018). According to Villanueva (2019), other risk factors are thought to exacerbate the occurrence of liver cancer, such as an unhealthy lifestyle, geographic conditions, gender, age, family history of the disease, and the severity of damage to the liver. Liver cancer is also found in areas that have cases of hepatitis B. In these areas, liver cancer is prevalent at a young age. This is because some infected with hepatitis B are obtained vertically through the delivery process (Mittal & El-Serag, 2013).

81 Patients often felt complaints in the form of fatigue, pain, diarrhea, skin abnormalities, and  
82 decreased appetite, all of which have affected their quality of life (Waller et al., 2015). Therefore,  
83 detecting the presence of disease symptoms in liver cancer can be done by examining  
84 deoxyribonucleic acid (DNA). Gene variation can be associated with disease progression and  
85 pathogenesis, which includes liver cancer. One of the websites through a bioinformatics  
86 approach that documents genetic variation is the GWAS catalog. GWAS catalog is a database  
87 with single nucleotide polymorphism (SNP) search results that has identified several variants  
88 associated with liver fat content, circulating liver enzymes, and the development of Non-  
89 Alcoholic Fatty Liver Disease (NAFLD) as well as genetic markers used in predicting a disease  
90 disorder (Wang et al., 2021).

91 Genetic identification in humans aims to identify inherited genetic risk factors for liver  
92 cancer. This study uses the GWAS catalog database to map genes from genetic variations across  
93 several populations that play an essential role in the pathogenesis of liver cancer. The most  
94 significant gene variations based on their function in protein changes will be further verified.  
95

## 96 Methods



97  
98 **Figure 1.** Analysis methodology for integrated bioinformatic, database and genomic analysis of  
99 genetic variation that affect liver cancer. The figure was created with BioRender.com under  
100 agreement number "FM250O073C"  
101

102 In this study, we adopted the method used by Ma'ruf et al (2023) and Puspitaningrum et  
103 al (2022), as depicted in Figure 1. Liver cancer-associated SNPs were obtained from the GWAS  
104 Catalog database (<http://www.ebi.ac.uk/gwas>; accessed on 15-02-2023). Subsequently, we  
105 performed further analysis using HaploReg (version 4.1) applying a  $p$ -value  $< 10^{-8}$  to account  
106 for multiple tests in the GWAS catalog. This threshold is commonly used to identify associations  
107 between common genetic variants and traits with adjacent gene expression (Chen et al., 2021).  
108 Furthermore, to evaluate the relationships between various genetic variants and gene expression  
109 profiles, we utilized e-QTL analysis with data sourced from the GTEx Portal database

110 (<http://www.gtexportal.org/home/>; accessed on 16-02-2023), considering gene expression across  
 111 various tissues in humans. Additionally, we confirmed the identified variants using the Ensembl  
 112 Genome Browser (<https://www.ensembl.org/index.html>; accessed on 17-02-2023). Our study  
 113 considered allele frequencies in populations from Europe, Africa, America, East Asia, and  
 114 Southeast Asia. To comprehend the functionalities of different gene variants, we performed  
 115 evaluations using the SNP nexus database (<https://www.snp-nexus.org>; accessed on 20-02-2023).  
 116 Furthermore, epidemiological and genomic data of the prevalence of liver cancer rates were  
 117 obtained from Li et al (2022). The prevalence rates and allele frequencies of the variants in  
 118 multiple continents were evaluated using IBM SPSS Statistics 25.0 with the Pearson Correlation  
 119 test. After the procedure was evaluated, the values of *p-value* 95% CI were obtained. All plots  
 120 were created using line charts. A *p-value* less than 0.05 ( $P \leq 0.05$ ) was considered as statistically  
 121 significant in current study.

122

## 123 Results and Discussion

### 124 1. Identification of Genomic Variants of Liver Cancer

125 This study identified SNPs associated with liver cancer from the GWAS catalog. Among  
 126 these SNPs, 29 of them were further confirmed through SNP duplication, as shown in Table 1.  
 127 Subsequently, HaploReg version 4.1 was utilized, and a  $p$ -value  $< 10^{-8}$  was applied based on the  
 128 number of SNPs obtained. Based on the findings presented in Table 2, we found the risk of two  
 129 genes for “Liver Cancer” disease. This study analyzed tissue expression affecting liver cancer,  
 130 focusing on the missense variant *PNPLA3*.

131 Through our integrative bioinformatics approach, two variants with a missense mutation (rs  
 132 rs2294915, rs2896019) that encoded the *PNPLA3* genes were prioritized as the biological risk  
 133 SNPs for Liver Cancer. Primary liver cancer, also known as hepatocellular carcinoma, is a  
 134 pathological condition characterised by the development of malignant cells within the hepatic  
 135 tissues. The development of cancer in extraneous anatomical sites that then metastasizes to the  
 136 liver does not constitute primary liver cancer. Primary liver cancer encompasses many kinds,  
 137 including hepatocellular carcinoma (HCC), intrahepatic cholangiocarcinoma (iCCA), and less  
 138 frequent varieties such as mixed hepatocellular cholangiocarcinoma (HCC-CCA), fibrolamellar  
 139 HCC (FLC), and the paediatric neoplasm hepatoblastoma (Wage et al., 2021).

140

141

**Table 1.** SNPs from the GWAS catalog with  $p$ -value  $< 10^{-8}$

No.	Variation and risk allele	p-value
1	rs2856723	$3 \times 10^{-13}$
2	rs34675408	$1 \times 10^{-32}$
3	rs9272105	$5 \times 10^{-22}$
4	rs913493	$5 \times 10^{-20}$
5	rs2294915	$2 \times 10^{-19}$
6	rs17401966	$2 \times 10^{-18}$
7	rs3096380	$1 \times 10^{-17}$
8	rs9275319	$3 \times 10^{-17}$
9	rs584368	$2 \times 10^{-14}$
10	rs2596542	$4 \times 10^{-13}$
11	rs1110446	$9 \times 10^{-13}$
12	rs58489806	$3 \times 10^{-12}$
13	rs6078460	$2 \times 10^{-11}$
14	rs2523961	$6 \times 10^{-11}$
15	rs7574865	$2 \times 10^{-10}$

16	rs1110446	3x10 <sup>-10</sup>
17	rs455804	5x10 <sup>-10</sup>
18	rs58542926	6x10 <sup>-10</sup>
19	rs2523961	6x10 <sup>-10</sup>
20	rs8107030	8x10 <sup>-10</sup>
21	rs10272859	9x10 <sup>-10</sup>
22	rs190121281	4x10 <sup>-9</sup>
23	rs9275572	6x10 <sup>-9</sup>
24	rs2242652	6x10 <sup>-9</sup>
25	rs188273166	1x10 <sup>-8</sup>
26	rs708113	1x10 <sup>-8</sup>
27	rs2896019	2x10 <sup>-8</sup>
28	rs17047200	3x10 <sup>-8</sup>
29	rs541860626	5x10 <sup>-8</sup>

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**Table 2.** Variants and risk alleles of liver cancer encoding prioritized SNPs

Variation and risk alleles	Variants near risk allele ( $r^2 > 0.8$ )	<i>p</i> -value	GENCODE	Type of allele
rs2294915	rs738409	2x10 <sup>-19</sup>	<i>PNPLA3</i>	missense
rs2896019	rs3761472	2x10 <sup>-8</sup>	<i>PNPLA3</i>	missense

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## 146 2. Gene expression of *PNPLA3* across 10 human tissues

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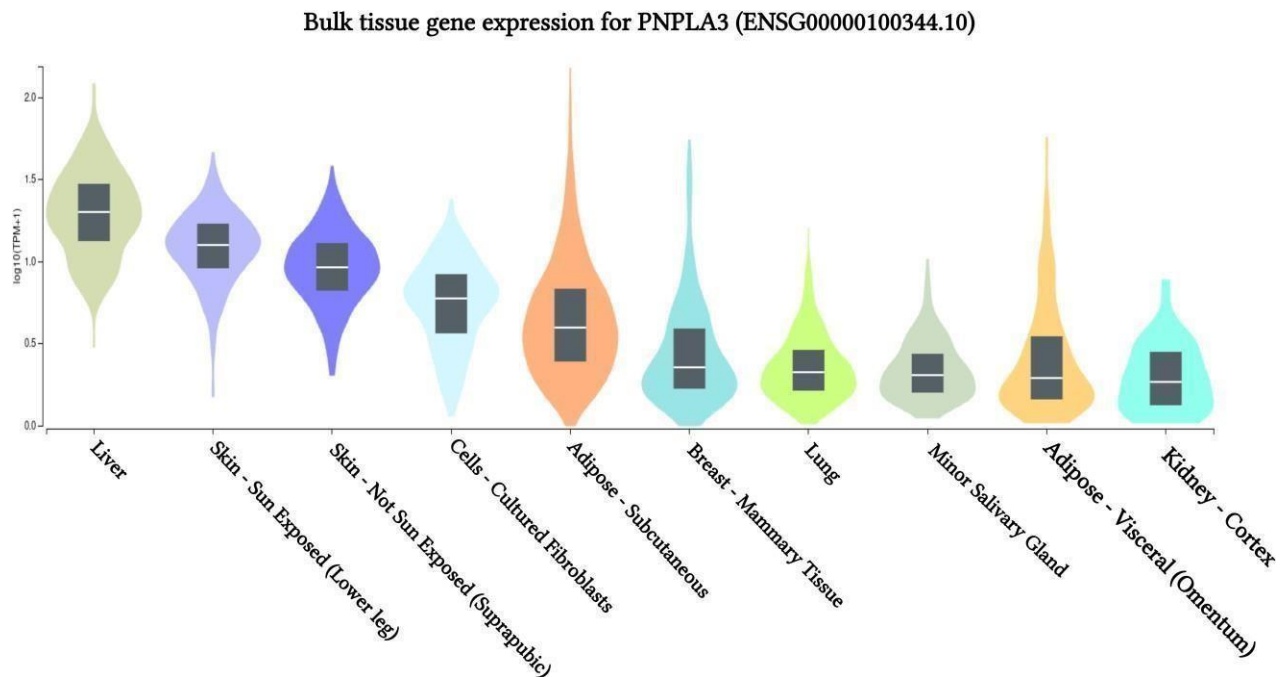
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The results of *PNPLA3* gene expression across 10 human tissues comprise the most apparent functional consequences of genetic variation. Liver, sun-exposed skin (lower legs), non-sun-exposed skin (suprapubic), and adipose subcutaneous fibroblasts and cell cultures showed the highest *PNPLA3* gene expression in the 10 human tissues analyzed from GTEx (Figure 2). In addition, we have found that the SNP IDs rs2294915 and rs2896019 have similar gene expression variations in Sun-Exposed skin (lower legs). The exciting thing about these findings is that liver cancer patients often experience complaints that their skin appears yellow. Further results showed that the *PNPLA3* gene has high expression in suprapubic and underarm skin.





**Figure 2.** *PNPLA3* gene expression associated with liver cancer across human tissues based on GTEx Portal analysis

### 3. Correlation between Gene Expression of *PNPLA3* and eQTL

The result in a correlation between the Gene Expression of *PNPLA3* and eQTL, to identify eQTLs associated with liver cancer gene expression, the GTEx database was used. We have identified minor alleles related to liver cancer, as presented in Table 3. Uniquely, several types of SNPs we found have high expression in skin tissue, namely rs2294915 and rs2896019. The CC type genotypes of rs2294915 and rs2896019 were associated with higher expression in suprapubic and underarm skin compared with the CT and TT type genotypes (Figure 3).

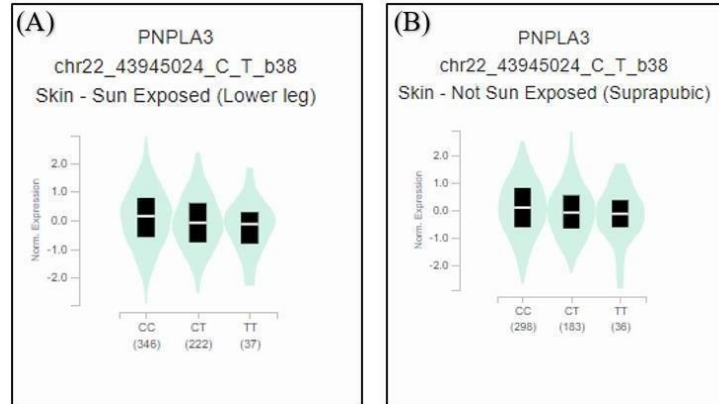
**Table 3.** Results of e-QTL in liver cancer from the GTEx portal database

SNP	Gencode ID (ENSG00000-)	Gene Symbol	<i>p</i> -value	Effect Size	Tissue	Expression Level
rs2294915	100344.10	<i>PNPLA3</i>	$2.8 \times 10^{-8}$	-0.15	Skin - Sun Exposed (Lower leg)	CC>CT>TT
	100344.10	<i>PNPLA3</i>	$5 \times 10^{-8}$	-0.50	Skin - Not Sun Exposed (Suprapubic)	CC>CT>TT
rs2896019	100344.10	<i>PNPLA3</i>	$6.7 \times 10^{-11}$	-0.19	Skin - Sun Exposed (Lower leg)	CC>CT>TT
	100344.10	<i>PNPLA3</i>	$2 \times 10^{-9}$	-0.22	Skin - Not Sun Exposed (Suprapubic)	CC>CT>TT

Source: Expression Quantitative Trait Loci (eQTL) obtained from the GTEx Portal.

Note : *PNPLA3*, patatin like phospholipase 3; SNP, single nucleotide polymorphism; eQTL, expression quantitative trait loci; SNP, single nucleotide polymorphism.

The research results show that the genomic database can identify gene variations with the most potential in the pathogenesis of liver cancer. Liver cancer is characterized by eyes and skin that appear yellow (Fitrianti et al., 2022). According to Nessa et al (2017), the severity of the liver can be seen from the decreasing quality of the liver. The quality of the liver can be assessed from total bilirubin levels, serum albumin, and PT (partial thromboplastin time).



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**Figure 3.** Patatin like phospholipase domain containing protein 3 (*PNPLA3*) gene expression for each genotype of the single nucleotide polymorphism (SNP): (A) rs2294915 and (B) rs2896019.

181 **4. Allele frequencies of candidate variants in populations in different continents**

182 The results of the research we have done, we have identified variants associated with liver  
183 cancer gene expression and carried out allele frequency analysis in various populations. As  
184 shown in Table 4, allele variant frequencies were evaluated in multiple people from Europe,  
185 America, East Asia, South Asia, and Africa. Allele frequencies across populations varied for each  
186 SNP, as depicted in Figure 4. Table 4 and Figure 4 show the gene expression levels at higher  
187 frequencies of the rs2294915 related allele (C) and the rs2896019 related allele (T). At the  
188 population frequency of the rs2294915 (C) allele, populations in Europe and South Asia were  
189 expressed at much higher levels than America, Africa, and East Asia.

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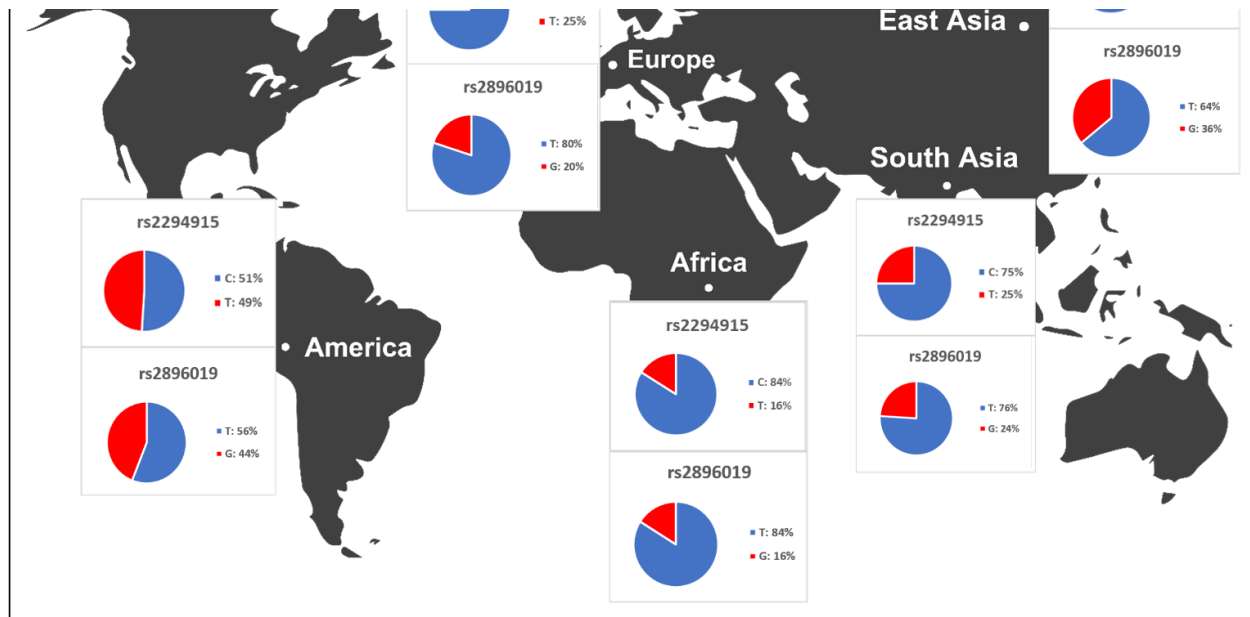
**Table 4.** Analysis of allele frequencies for the *PNPLA3* gene from variant annotation (SNPnexus)

SNP	Gene	Location	Allele		Allele Frequency (N)				
			Ref	Eff*	AFR	AMR	EAS	EUR	SAS
rs2294915	<i>PNPLA3</i>	Missense	C	T	T: 0.163 (215)	T: 0.490 (340)	T: 0.365 (368)	T: 0.252 (254)	T: 0.246 (241)
rs2896019	<i>PNPLA3</i>	Missense	T	G	G: 0.158 (209)	G: 0.438 (304)	G: 0.364 (367)	G: 0.199 (200)	G: 0.236 (231)

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Note : *PNPLA3*, patatin like phospholipase domain containing protein 3; SNP, single nucleotide polymorphism; Ref, Reference; Alt, Alternate; AFR, Africa; AMR, America; EAS, East Asia; EUR, Europe; SAS, Southeast Asia.





197 **Figure 4.** The results of the distribution of allele frequencies affecting *PNPLA3* across various  
 198 populations. EUR, Europe; AFR, Africa; SAS, South Asia; EAS, East Asia.

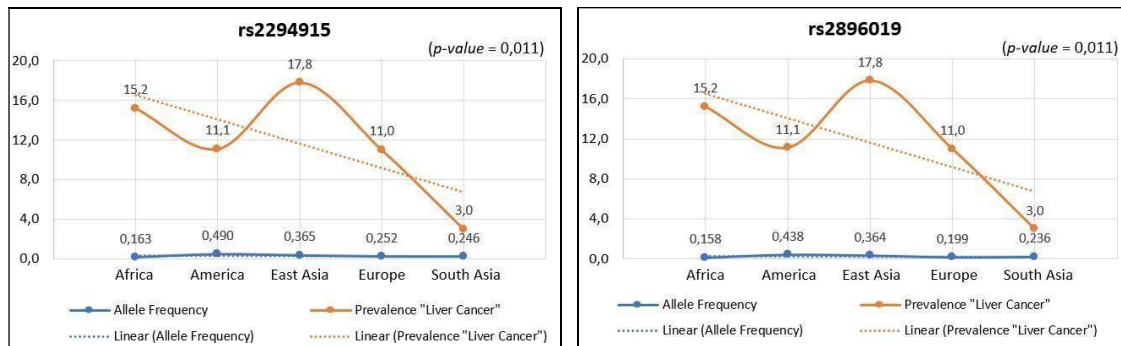
199

200 Based on this finding, rs2294915 and rs2896019 are potentially related to the susceptibility to  
 201 “liver cancer with effect size the highest score of -0.50 can be interpreted on the skin not exposed  
 202 to sunlight (suprapubic).” According to Poggiali & Vercelli (2023), this condition is  
 203 characterized by a disruption in the heme biosynthesis pathway due to reduced hepatic  
 204 uroporphyrinogen decarboxylase (UROD) activity. The consequence of this phenomenon is the  
 205 buildup of light-sensitive by-products, such as uroporphyrinogen, leading to the development of  
 206 fragility and blistering of sun-exposed skin, as well as impairment of liver function.

207 The allele frequencies of the T and G alleles “rs2294915” and “rs2896019” in African  
 208 populations were expressed at much lower levels compared to American, European, and  
 209 Southeast Asian people. Overall, the allele frequencies of the “rs2294915” and “rs2896019”  
 210 variant alleles suggest a contribution to the prevalence of the variants for gene expression of  
 211 *PNPLA3*.

212 Across human populations, the frequency of the T allele at rs2294915 was associated with a  
 213 high expression of *PNPLA3* in liver cancer, which is much lower in African populations (16%)  
 214 compared to South Asians (25%), Europeans (25%), East Asians (37%) and America (49%). In  
 215 contrast, the frequency of the C allele at rs2296019 was considerably higher in African (84%),  
 216 European (80%), South Asian (76%), East Asian (64%), and American (56%) populations. Next,  
 217 the association between allele frequency and the prevalence of the liver cancer of each continent  
 218 was evaluated. Data on the prevalence of liver cancer in the continents were obtained from Li et  
 219 al (2022) (Li et al., 2023). Herein, two SNPs (rs2294915 and rs2896019) were positively  
 220 correlated with the prevalence rate of liver cancer in multiple continents (Africa, America, East

221 Asia, Europe, South Asia) based on the pearson's correlation analysis ( $P$ -value<0.011) (Figure  
 222 5). Populations with higher frequencies of variant alleles of these polymorphisms are thought to  
 223 have a higher prevalence. We highlighted that these two variants (rs2294915 and rs2896019)  
 224 were more frequent in the East Asian and Africa, which performed the higher aggressiveness of  
 225 liver cancer in East Asian and African compared to America, Europe and South Asia. This study  
 226 might give an insight that East Asian and African with carriers variants rs2294915 and rs2896019  
 227 might be more highly susceptible to suffer the liver cancer.  
 228



229 **Figure 5.** The association between allele frequency and the prevalence of liver cancer of each  
 230 continent  
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233 Patients with liver cancer who have a history of alcohol addiction to an amount of  $\geq 3$  drinks  
 234 per day can increase the risk of liver cancer by 16% in the general population; diabetics and  
 235 people with central obesity also increase the risk of liver cancer by 2 times (McGlynn et al.,  
 236 2021). The diagnosis of liver cancer in patients often involves using serological testing in  
 237 conjunction with imaging techniques, which is considered the established approach for  
 238 identifying liver carcinoma. Nevertheless, the sensitivity of the often-used serological test,  
 239 specifically designed to detect alpha-fetoprotein (AFP), is at around 60%. Imaging modalities,  
 240 including magnetic resonance imaging (MRI), computed tomography (CT), and ultrasonography  
 241 (US), exhibit notable levels of sensitivity and specificity in the identification of liver cancer,  
 242 particularly in individuals afflicted with liver cirrhosis (Huang et al., 2022).

243 Variant alleles (rs2294915 and rs2896019) are associated with liver cancer. Accordingly,  
 244 populations from the continents of Africa, America, East Asia, Europe, and South Asia show  
 245 associated *PNPLA3* expression, resulting in a higher susceptibility to liver cancer. Identifying  
 246 unique and pathogenic gene variations for a disease is very interesting for both research and  
 247 clinical validation. These variants not only offer insights into disease susceptibility but also  
 248 serve as potential diagnostic and prognostic biomarkers (Irham et al., 2020). Additionally, they  
 249 can facilitate the identification of drug target candidates, a concept known as genomic-driven  
 250 drug repurposing (Afief et al., 2022). We anticipate that the discovery of candidate gene  
 251 variations in *PNPLA3* will pave the way for successful clinical validation, potentially  
 252 establishing this as a promising diagnostic and prognostic biomarker for liver cancer.

253 It is important to acknowledge that the genetic variants identified in this study as potentially  
 254 pathogenic are based on preliminary investigations using genomic and bioinformatics databases.  
 255 However, these findings offer crucial insights for future researchers intending to validate these  
 256 genetic variants in liver cancer patients. We strongly recommend future research incorporate

257 further functional annotations, which would further aid in prioritizing these pathogenic genetic  
258 variants.

259

## 260 **Conclusion**

261 This study identifies genetic variants influencing “liver cancer” and reveals the significance of  
262 the *PNPLA3* gene in liver tissue. consequently, these population groups exhibit varying  
263 susceptibilities to liver cancer based on associated *PNPLA3* expression levels. . The observed  
264 variations in allele frequencies of the two identified variants, rs2294915, and rs2896019, across  
265 populations from Africa, America, East Asia, Europe, and South Asia, significantly impact  
266 *PNPLA3* gene expression. Our study also demonstrated that these two SNPs (rs2294915 and  
267 rs2896019) were positively correlated with the prevalence rate. Positive association of prevalence  
268 rates were underlined more frequent in East Asian and African population. The higher frequency of the  
269 variants allele of these polymorphisms in population, the higher the estimated prevalence rates. The  
270 variants investigated in this study were likely to predispose to liver cancer and could play a role  
271 in its progression and aggressiveness. These discoveries highlight the critical relevance of  
272 understanding genomic variations in precision medicine and designing screening strategies for  
273 liver cancer across diverse populations on different continents.

274

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**Pada tanggal 29 November 2023 hasil revisian yang kedua telah diterima oleh editor jurnal. Dan pada tanggal 30 November 2023 artikel telah diterima / accepted oleh editor.**

#### Editorial Comment

Invited date	Nov 29, 2023
Complete Review Date	Nov 30, 2023
Recommendation	Accept as it is
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<p><b>Q5</b></p>	<p>Authors' Contribution            Conceptualization:            Data curation:            Formal analysis:            Funding acquisition:            Methodology:</p>	<p>Conceptualization: MM, LMI, WA.            Data curation: MM, LMI            Formal analysis: MM, LMI, WA            Methodology: MM, LMI, WA.            Writing – original draft: MM, LMI            Writing – review &amp; editing: MM, LMI, WA, BDP, MAS, SK, RC, MM, LMHS.</p>

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<b>Q8</b>	<p>If possible, please change the corresponding text as follows:</p> <p>SNPs associated with liver cancer</p> <p>SNPs encoded by the genes</p> <p>Gene expression of SNPs associated with liver cancer</p> <p>Allele frequencies on multiple continents</p>	These terms were correct

# A genomic and bioinformatic-based approach to identify genetic variants for liver cancer across multiple continents

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Liver cancer is the fourth leading cause of death worldwide. Well-known risk factors include hepatitis B virus and hepatitis C virus, along with exposure to aflatoxins, excessive alcohol consumption, obesity, and type 2 diabetes. Genomic variants play a crucial role in mediating the associations between these risk factors and liver cancer. However, the specific variants involved in this process remain under-explored. This study utilized a bioinformatics approach to identify genetic variants associated with liver cancer from various continents. Single-nucleotide polymorphisms associated with liver cancer were retrieved from the genome-wide association studies catalog. Prioritization was then performed using functional annotation with HaploReg v4.1 and the Ensembl database. The prevalence and allele frequencies of each variant were evaluated using Pearson correlation coefficients. Two variants, rs2294915 and rs2896019, encoded by the *PNPLA3* gene, were found to be highly expressed in the liver tissue, as well as in the skin, cell-cultured fibroblasts, and adipose-subcutaneous tissue, all of which contribute to the risk of liver cancer. We further found that these two SNPs (rs2294915 and rs2896019) were positively correlated with the prevalence rate. Positive associations with the prevalence rate were more frequent in East Asian and African populations. We highlight the utility of this population-specific *PNPLA3* genetic variant for genetic association studies and for the early prognosis and treatment of liver cancer. This study highlights the potential of integrating genomic databases with bioinformatic analysis to identify genetic variations involved in the pathogenesis of liver cancer. The genetic variants investigated in this study are likely to predispose to liver cancer and could affect its progression and aggressiveness. We recommend future research prioritizing the validation of these variations in clinical settings.

**Keywords:** bioinformatics, genomic variants, liver neoplasms

## Introduction

Patients often report symptoms such as fatigue, pain, diarrhea, skin abnormalities, and decreased appetite, all of which can ad-

**Fig. 1.** Analytical methodology for integrated bioinformatic, database, and genomic analysis of genetic variations that affect liver cancer. The figure was created with BioRender.com under agreement number “FM2500073C”.

ciated SNPs were obtained from the GWAS Catalog database (<http://www.ebi.ac.uk/gwas>; accessed on 15-02-2023). Subse-

two variants with missense mutations (rs2294915 and rs2896019) that encode the *PNPLA3* gene as biological risk SNPs for liver can-

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rs2896019	rs3761472	$2 \times 10^{-8}$	<i>PNPLA3</i>	Missense
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SNP, single-nucleotide polymorphism

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Based on these findings, rs2294915 and rs2896019 may be associated with an increased susceptibility to liver cancer, with the high-

rs2896019 FNFLAS missense T G G 0.136 (207) G 0.436 (304) G 0.304 (307) G 0.177 (200) G 0.230 (231)

SNP, single-nucleotide polymorphism; Ref, reference; Alt, alternate; AFR, Africa; AMR, America; EAS, East Asia; EUR, Europe; SAS, Southeast Asia.



allele at rs2296019 is considerably higher in African (84%), European (80%), South Asian (76%), East Asian (64%), and American

[12,18]. In this context, two SNPs (rs2294915 and rs2896019) were found to be positively correlated with the prevalence rate of



liver cancer across multiple continents (Africa, America, East Asia, Europe, South Asia), as determined by Pearson's correlation analysis ( $p = 0.011$ ) (Fig. 5). Populations with higher frequencies of variant alleles of these polymorphisms are thought to have a higher prevalence of liver cancer. We highlighted that these two variants (rs2294915 and rs2896019) are more frequent in East Asian and African populations, which exhibit higher aggressiveness of liver cancer compared to America, Europe, and South Asia. This study suggests that individuals in East Asian and African populations carrying the variant alleles rs2294915 and rs2896019 may be more susceptible to liver cancer.

Patients with liver cancer who also have a history of alcohol abuse, consuming  $\geq 3$  drinks per day, have a 16% increased risk of developing liver cancer compared to the general population. Additionally, individuals with diabetes and those with central obesity are at twice the risk of developing liver cancer [1]. The diagnosis of liver cancer typically involves serological testing combined with imaging techniques, which is the standard approach for detecting liver carcinoma. However, the sensitivity of the commonly used serological test, which is designed to detect alpha-fetoprotein, is only about 60%. Imaging modalities such as magnetic resonance imaging, computed tomography, and ultrasonography demonstrate high levels of sensitivity and specificity in detecting liver cancer, especially in patients with liver cirrhosis [19].

Variant alleles (rs2294915 and rs2896019) are associated with liver cancer. Populations from Africa, America, East Asia, Europe, and South Asia exhibit associated *PNPLA3* expression, which leads to an increased susceptibility to liver cancer. The identification of unique and pathogenic gene variations for a disease is of great interest for both research and clinical validation. These variants provide insights into disease susceptibility and also act as potential diagnostic and prognostic biomarkers [20]. Furthermore, they can aid in the identification of drug target candidates, an approach referred to as genomic-driven drug repurposing [21]. We expect that the discovery of candidate gene variations in *PNPLA3* will facilitate successful clinical validation, potentially establishing it as a promising diagnostic and prognostic biomarker for liver cancer.

It is important to acknowledge that the genetic variants identified in this study as potentially pathogenic are based on preliminary investigations using genomic and bioinformatics databases. While these findings provide crucial insights for future researchers aiming to validate these genetic variants in liver cancer patients, it is important to proceed with caution. We strongly recommend that future research includes additional functional annotations to aid in the prioritization of these pathogenic genetic variants.

 This study identified genetic variants that influence liver cancer,

highlighting the importance of the *PNPLA3* gene in liver tissue. Consequently, these population groups exhibit varying susceptibilities to liver cancer based on the associated *PNPLA3* expression levels. The observed variations in allele frequencies of the two identified variants, rs2294915 and rs2896019, across populations from Africa, America, East Asia, Europe, and South Asia, significantly impact *PNPLA3* gene expression. Our study also demonstrated that these two SNPs (rs2294915 and rs2896019) were positively correlated with the prevalence rate. The positive association of prevalence rates was more frequently observed in East Asian and African populations. The higher the frequency of the variant alleles of these polymorphisms in a population, the higher the estimated prevalence rates. The variants investigated in this study are likely to predispose individuals to liver cancer and could play a role in its progression and aggressiveness. These findings highlight the critical importance of understanding genomic variations for precision medicine and for designing targeted screening strategies for liver cancer across diverse populations on different continents.

### Authors' Contribution

- Conceptualization:
- Data curation:
- Formal analysis:
- Funding acquisition:
- Methodology:
- Writing - original draft:
- Writing - review & editing:

### Conflicts of Interest

No potential conflict of interest relevant to this article was reported.

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**Conclusion**  
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## A genomic and bioinformatic-based approach to identify genetic variants for liver cancer across multiple continents

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Sabiah Khairi <sup>5</sup>, Barkah Djaka Purwanto <sup>6,7</sup>, Rockie Chong <sup>8</sup>, Maulida Mazaya <sup>9</sup>, Lalu Muhammad  
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# A genomic and bioinformatic-based approach to identify genetic variants for liver cancer across multiple continents

Muhammad Ma'ruf<sup>1</sup>, Lalu Muhammad Irham<sup>1\*</sup>, Wirawan Adikusuma<sup>2</sup>, Made Ary Sarasmita<sup>3,4</sup>, Sabiah Khairi<sup>5</sup>, Barkah Djaka Purwanto<sup>6,7</sup>, Rockie Chong<sup>8</sup>, Maulida Mazaya<sup>9</sup>, Lalu Muhammad Harmain Siswanto<sup>10</sup>

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Liver cancer is the fourth leading cause of death worldwide. Well-known risk factors include hepatitis B virus and hepatitis C virus, along with exposure to aflatoxins, excessive alcohol consumption, obesity, and type 2 diabetes. Genomic variants play a crucial role in mediating the associations between these risk factors and liver cancer. However, the specific variants involved in this process remain under-explored. This study utilized a bioinformatics approach to identify genetic variants associated with liver cancer from various continents. Single-nucleotide polymorphisms associated with liver cancer were retrieved from the genome-wide association studies catalog. Prioritization was then performed using functional annotation with HaploReg v4.1 and the Ensembl database. The prevalence and allele frequencies of each variant were evaluated using Pearson correlation coefficients. Two variants, rs2294915 and rs2896019, encoded by the *PNPLA3* gene, were found to be highly expressed in the liver tissue, as well as in the skin, cell-cultured fibroblasts, and adipose-subcutaneous tissue, all of which contribute to the risk of liver cancer. We further found that these two SNPs (rs2294915 and rs2896019) were positively correlated with the prevalence rate. Positive associations with the prevalence rate were more frequent in East Asian and African populations. We highlight the utility of this population-specific *PNPLA3* genetic variant for genetic association studies and for the early prognosis and treatment of liver cancer. This study highlights the potential of integrating genomic databases with bioinformatic analysis to identify genetic variations involved in the pathogenesis of liver cancer. The genetic variants investigated in this study are likely to predispose to liver cancer and could affect its progression and aggressiveness. We recommend future research prioritizing the validation of these variations in clinical settings.

**Keywords:** bioinformatics, genomic variants, liver cancer, *PNPLA3*, rs2294915, rs2896019



## Introduction

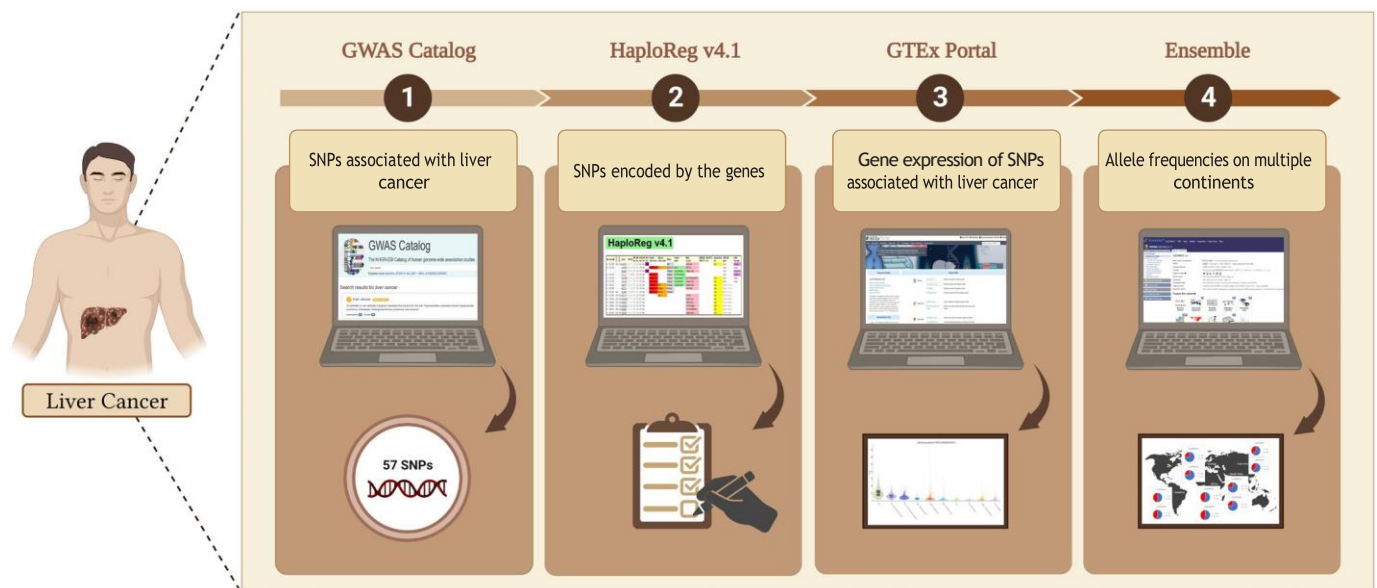
Liver cancer, a type of carcinoma, has the highest mortality rate in the world each year [1]. In 2018, there were 841,000 new cases of liver cancer, and the death toll reached 782,000 [2]. The average incidence of liver cancer and the associated mortality rate can be two to three times higher in men, particularly in certain regions of the world. According to the Global Cancer Statistics (GLOBOCAN) in 2020, liver cancer was ranked as the third most deadly cancer, responsible for 8.3% of all cancer-related deaths. In that year, there were 905,000 new cases of liver cancer, with a mortality rate of 830,000 [3]. In Indonesia, liver cancer is the second most common cancer among men, with an incidence rate of 12.4 per 100,000 of the population and an average mortality rate of 7.6 per 100,000 [4]. Factors that contribute to liver cancer include chronic infection with hepatitis B virus and hepatitis C virus, exposure to aflatoxin contamination, alcohol consumption, a history of obesity, type 2 diabetes, and smoking addiction [2]. Villanueva [5] notes that additional risk factors may exacerbate the incidence of liver cancer, including an unhealthy lifestyle, geographic conditions, gender, age, family history of the disease, and the extent of liver damage. Liver cancer is also prevalent in regions with high rates of hepatitis B infection. In these areas, the disease often manifests at a younger age, partly because hepatitis B can be transmitted vertically from mother to child during childbirth [6].

Patients often report symptoms such as fatigue, pain, diarrhea, skin abnormalities, and decreased appetite, all of which can adversely affect their quality of life [7]. Consequently, the detection of disease symptoms in liver cancer can involve examining DNA. Variations in genes may be linked to the progression and pathogenesis of diseases, including liver cancer. The genome-wide association studies (GWAS) Catalog is a resource that employs a bioinformatics approach to document genetic variations. This database contains search results for single-nucleotide polymorphisms (SNPs) and has identified several variants associated with liver fat content, circulating liver enzymes, and the development of non-alcoholic fatty liver disease, as well as genetic markers useful in predicting disease disorders [8].

Genetic identification studies in humans aim to identify inherited genetic risk factors for various conditions, including liver cancer. This study used the GWAS catalog database to map genes from genetic variations across several populations that play an essential role in the pathogenesis of liver cancer. The most significant gene variations based on their function in protein changes were further verified.

## Methods

In this study, we adopted the method used by Ma'rif et al. [9] and Puspitaningrum et al. [10], as depicted in Fig. 1. Liver cancer-asso-



**Fig. 1.** Analytical methodology for integrated bioinformatic, database, and genomic analysis of genetic variations that affect liver cancer. The figure was created with BioRender.com under agreement number "FM2500073C". SNP, single-nucleotide polymorphism; GWAS, genome-wide association study.

ciated SNPs were obtained from the GWAS Catalog database (<http://www.ebi.ac.uk/gwas>; accessed on 15-02-2023). Subsequently, we performed further analysis using HaploReg (version 4.1) applying a  $p < 10^{-8}$  to account for multiple tests in the GWAS catalog. This threshold is commonly used to identify associations between common genetic variants and traits with adjacent gene expression [11]. Furthermore, to evaluate the relationships between various genetic variants and gene expression profiles, we conducted an analysis of expression quantitative trait loci (eQTLs) with data sourced from the GTEx Portal database (<http://www.gtexportal.org/home/>; accessed on 16 Feb 2023), considering gene expression across various tissues in humans. Additionally, we confirmed the identified variants using the Ensembl Genome Browser (<https://www.ensembl.org/index.html>; accessed on 17 Feb 2023). Our study considered allele frequencies in populations from Europe, Africa, America, East Asia, and Southeast Asia. To explore the functionalities of different gene variants, we performed evaluations using the SNP nexus database (<https://www.snp-nexus.org>; accessed on 20 Feb 2023). Furthermore, epidemiological and genomic data on the prevalence of liver cancer rates were obtained from Li et al. [12]. The prevalence rates and allele frequencies of the variants in multiple continents were evaluated using IBM SPSS Statistics 25.0 (IBM Corp., Armonk, NY, USA) with the Pearson correlation test. After the procedure was evaluated, the p-values were obtained. All plots were created using line charts. A  $p < 0.05$  was considered statistically significant in the current study.

## Results and Discussion

### Identification of genomic variants of liver cancer

This study identified SNPs associated with liver cancer from the GWAS catalog. Of these SNPs, 29 were further confirmed through SNP genotyping, as shown in Table 1. Subsequently, HaploReg version 4.1 was utilized, applying a p-value threshold of  $<10^{-8}$  based on the number of SNPs obtained. The findings presented in Table 2 indicate an increased risk associated with two genes for liver cancer. The study also analyzed tissue expression impacting liver cancer, with a focus on missense variants of *PNPLA3* (patatin-like phospholipase domain-containing 3).

Through our integrative bioinformatics approach, we prioritized two variants with missense mutations (rs2294915 and rs2896019) that encode the *PNPLA3* gene as biological risk SNPs for liver cancer. Primary liver cancer is a pathological condition characterized by the development of malignant cells within the hepatic tissues. The development of cancer at extraneous anatomical sites that subsequently metastasizes to the liver does not constitute primary liver

**Table 1.** SNPs from the GWAS catalog with a  $p < 10^{-8}$

No.	Variation and risk allele	p-value
1	rs2856723	$3 \times 10^{-43}$
2	rs34675408	$1 \times 10^{-32}$
3	rs9272105	$5 \times 10^{-22}$
4	rs913493	$5 \times 10^{-20}$
5	rs2294915	$2 \times 10^{-19}$
6	rs17401966	$2 \times 10^{-18}$
7	rs3096380	$1 \times 10^{-17}$
8	rs9275319	$3 \times 10^{-17}$
9	rs584368	$2 \times 10^{-14}$
10	rs2596542	$4 \times 10^{-13}$
11	rs1110446	$9 \times 10^{-13}$
12	rs58489806	$3 \times 10^{-12}$
13	rs6078460	$2 \times 10^{-11}$
14	rs2523961	$6 \times 10^{-11}$
15	rs7574865	$2 \times 10^{-10}$
16	rs1110446	$3 \times 10^{-10}$
17	rs455804	$5 \times 10^{-10}$
18	rs58542926	$6 \times 10^{-10}$
19	rs2523961	$6 \times 10^{-10}$
20	rs8107030	$8 \times 10^{-10}$
21	rs10272859	$9 \times 10^{-10}$
22	rs190121281	$4 \times 10^{-9}$
23	rs9275572	$6 \times 10^{-9}$
24	rs2242652	$6 \times 10^{-9}$
25	rs188273166	$1 \times 10^{-8}$
26	rs708113	$1 \times 10^{-8}$
27	rs2896019	$2 \times 10^{-8}$
28	rs17047200	$3 \times 10^{-8}$
29	rs541860626	$5 \times 10^{-8}$

SNP, single-nucleotide polymorphism; GWAS, genome-wide association study.

**Table 2.** Variants and risk alleles of the prioritized SNPs for liver cancer

Variation and risk alleles	Variants near risk allele ( $r^2 > 0.8$ )	p-value	GENCODE	Type of allele
rs2294915	rs738409	$2 \times 10^{-19}$	<i>PNPLA3</i>	Missense
rs2896019	rs3761472	$2 \times 10^{-8}$	<i>PNPLA3</i>	Missense

SNP, single-nucleotide polymorphism.

cancer. Primary liver cancer includes several types, such as hepatocellular carcinoma (HCC), intrahepatic cholangiocarcinoma, and less common varieties like mixed hepatocellular cholangiocarcinoma, fibrolamellar HCC, and the pediatric neoplasm hepatoblastoma [13].

**Gene expression of *PNPLA3* across 10 human tissues**

The results of *PNPLA3* gene expression across 10 human tissues revealed significant functional consequences of genetic variation. The highest levels of *PNPLA3* gene expression were observed in the liver, sun-exposed skin (lower legs), non-sun-exposed skin (suprapubic), and adipose-subcutaneous fibroblasts and cell cultures, according to analyses of the 10 human tissues from the GTEx database (Fig. 2). Additionally, we found that the SNP IDs rs2294915 and rs2896019 exhibited similar patterns of gene expression variation in sun-exposed skin (lower legs). Notably, patients with liver

cancer often report that their skin appears yellow, which may be related to these findings. Further analysis indicated that the *PNPLA3* gene is also highly expressed in suprapubic and underarm skin.

**Correlation between gene expression of *PNPLA3* and eQTLs**

The study revealed a correlation between the gene expression of *PNPLA3* and eQTLs. To identify eQTLs associated with liver cancer gene expression, we utilized the GTEx database. We identified minor alleles that are related to liver cancer, as detailed in Table 3 [14]. Notably, we discovered that several SNPs, specifically rs2294915 and rs2896019, exhibit high expression in skin tissue. The CC genotype of both rs2294915 and rs2896019 was associated with increased expression in suprapubic and underarm skin compared to the CT and TT genotypes, as shown in Fig. 3.

The research results show that the genomic database could be used to identify gene variations with significant potential in the

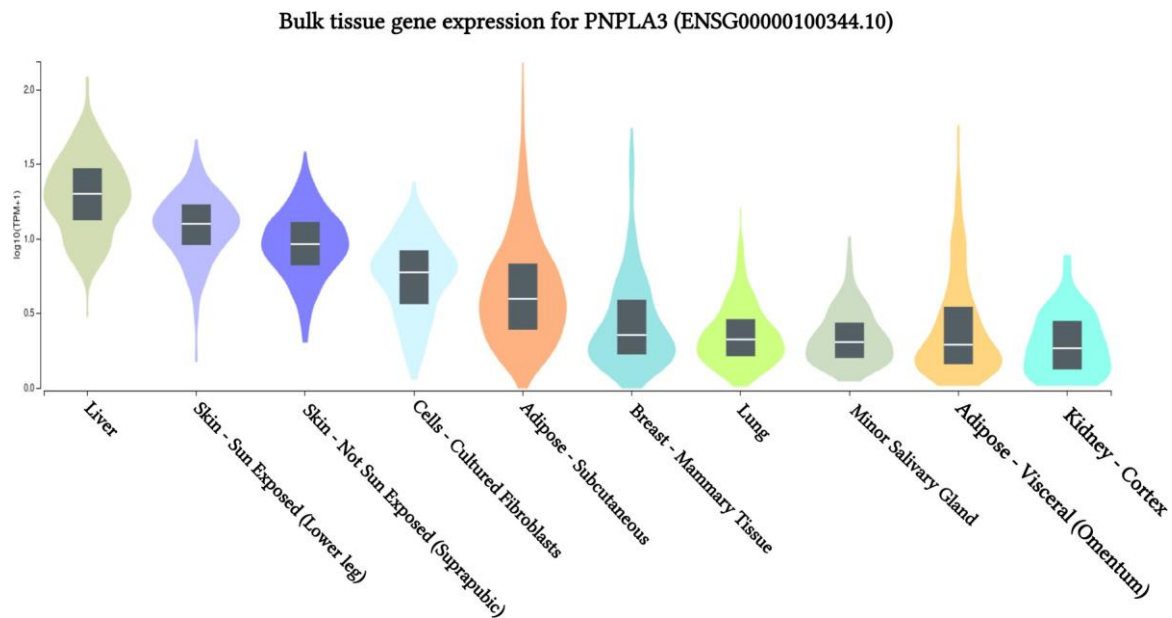


Fig. 2. *PNPLA3* gene expression associated with liver cancer across human tissues based on GTEx Portal analysis.

Table 3. Results of eQTLs in liver cancer from the GTEx Portal database

SNP	Gencode ID (ENSG00000-)	Gene symbol	p-value	Effect size	Tissue	Expression level
rs2294915	100344.1	<i>PNPLA3</i>	$2.8 \times 10^{-8}$	-0.15	Skin - sun exposed (lower leg)	CC > CT > TT
	100344.1	<i>PNPLA3</i>	$5 \times 10^{-8}$	-0.50	Skin - not sun exposed (suprapubic)	CC > CT > TT
rs2896019	100344.1	<i>PNPLA3</i>	$6.7 \times 10^{-11}$	-0.19	Skin - sun exposed (lower leg)	CC > CT > TT
	100344.1	<i>PNPLA3</i>	$2 \times 10^{-9}$	-0.22	Skin - not sun exposed (suprapubic)	CC > CT > TT

Source: expression quantitative trait loci (eQTLs) obtained from the GTEx Portal [14]. eQTL, expression quantitative trait loci; SNP, single-nucleotide polymorphism.

pathogenesis of liver cancer. Liver cancer is marked by the yellowing of the eyes and skin [15]. Nessa et al. [16] note that the severity of liver disease can be gauged by the declining quality of liver function. This quality can be evaluated by measuring total bilirubin levels, serum albumin, and prothrombin time.

**Allele frequencies of candidate variants in populations in different continents**

We identified variants associated with liver cancer gene expression and conducted allele frequency analysis across various populations. As indicated in Table 4, we evaluated the frequency of allele variants in individuals from Europe, America, East Asia, South Asia, and Africa. The allele frequencies for each SNP differed among these populations, as illustrated in Fig. 4. Both Table 4 and Fig. 4 demonstrate that gene expression levels are higher for populations with increased frequencies of the rs2294915 (C) allele and the rs2896019 (T) allele. Specifically, the gene expression associated with the rs2294915 (C) allele was significantly higher in European and South Asian populations compared to those in America, Africa, and East Asia.

Based on these findings, rs2294915 and rs2896019 may be asso-

ciated with an increased susceptibility to liver cancer, with the highest effect size of -0.50 observed on skin not exposed to sunlight, such as the suprapubic area. Poggiali and Vercelli [17] describe this condition as being characterized by a disruption in the heme biosynthesis pathway, which is due to decreased activity of hepatic uroporphyrinogen decarboxylase. This disruption leads to an accumulation of light-sensitive by-products, including uroporphyrinogen, resulting in the development of skin fragility and blistering in areas exposed to the sun, as well as impaired liver function.

The allele frequencies of the T and G alleles at loci rs2294915 and rs2896019 were significantly lower in African populations compared to those in American, European, and Southeast Asian populations. Overall, the allele frequencies of the variant alleles rs2294915 and rs2896019 suggest they may contribute to the prevalence of variants affecting the gene expression of *PNPLA3*.

Across human populations, the frequency of the T allele at rs2294915 is associated with high expression of *PNPLA3* in liver cancer. This frequency is much lower in African populations (16%) compared to South Asians (25%), Europeans (25%), East Asians (37%), and Americans (49%). Conversely, the frequency of the C

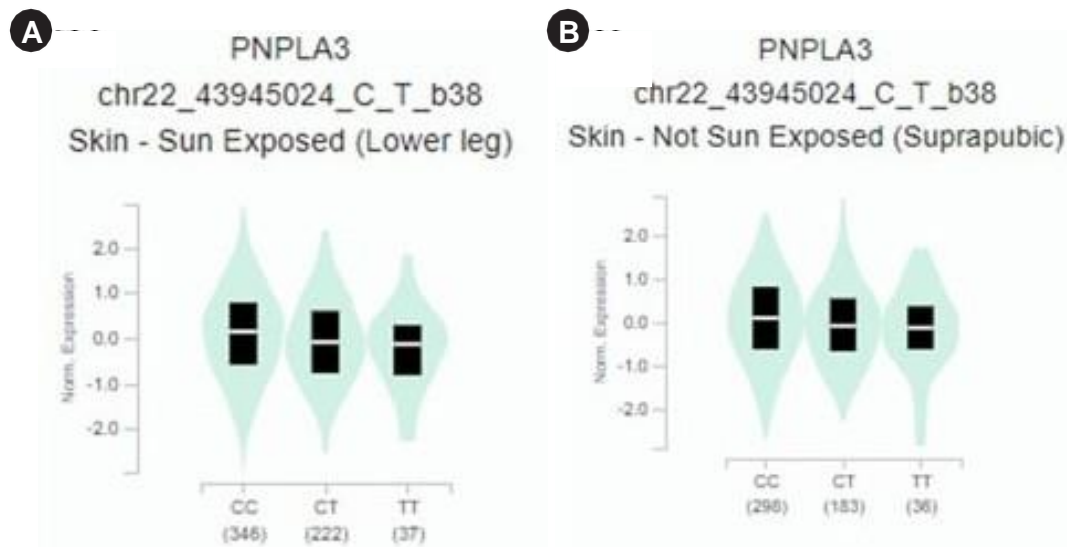


Fig. 3. *PNPLA3* gene expression for each genotype of the single-nucleotide polymorphisms: (A) rs2294915 and (B) rs2896019.

Table 4. Analysis of allele frequencies for the *PNPLA3* gene from variant annotation (SNPnexus)

SNP	Gene	Location	Allele		Allele frequency (n)				
			Ref	Eff*	AFR	AMR	EAS	EUR	SAS
rs2294915	<i>PNPLA3</i>	Missense	C	T	T: 0.163 (215)	T: 0.490 (340)	T: 0.365 (368)	T: 0.252 (254)	T: 0.246 (241)
rs2896019	<i>PNPLA3</i>	Missense	T	G	G: 0.158 (209)	G: 0.438 (304)	G: 0.364 (367)	G: 0.199 (200)	G: 0.236 (231)

SNP, single-nucleotide polymorphism; Ref, reference; Eff, alternate; AFR, Africa; AMR, America; EAS, East Asia; EUR, Europe; SAS, Southeast Asia.



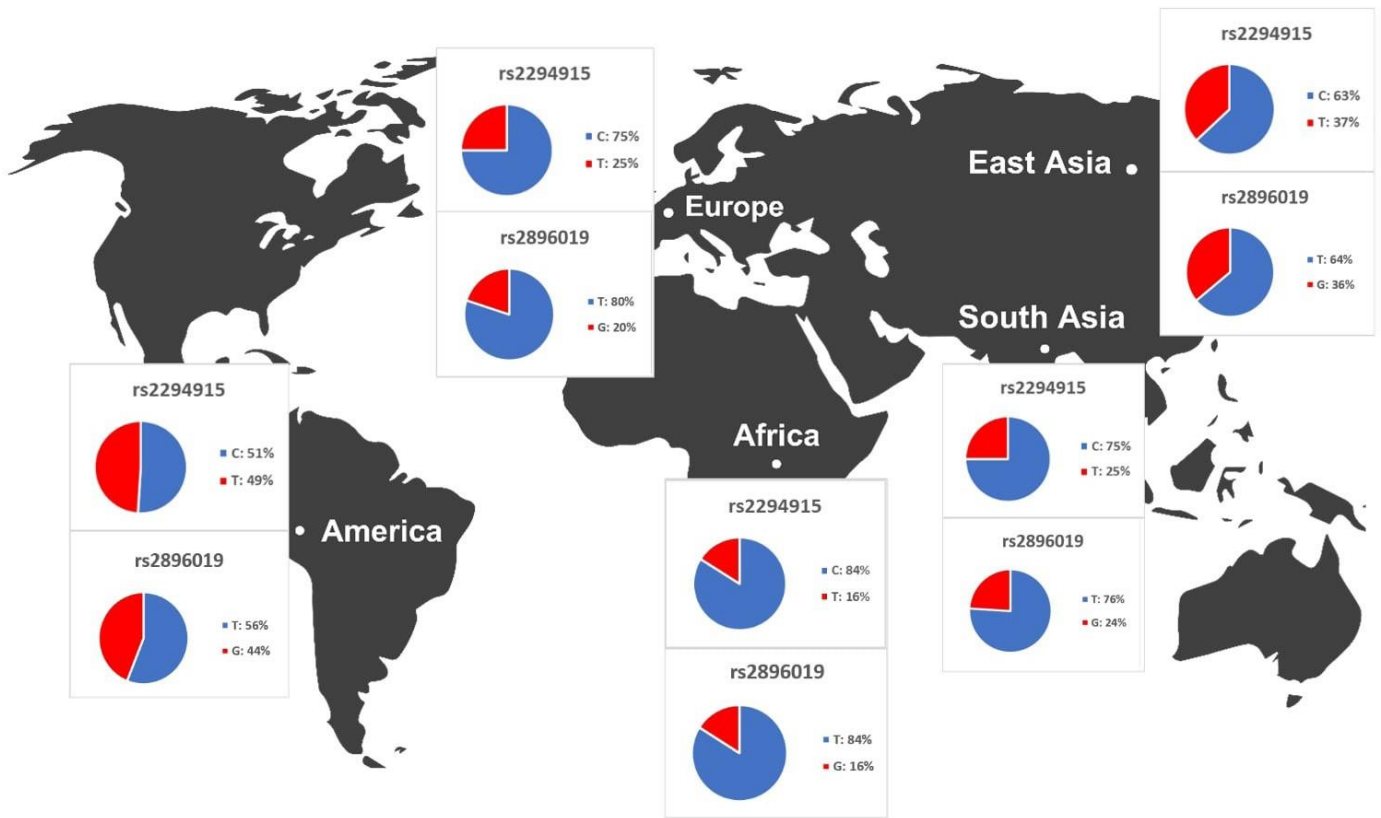


Fig. 4. The results of the distribution of *PNPLA3* allele frequencies across various populations.

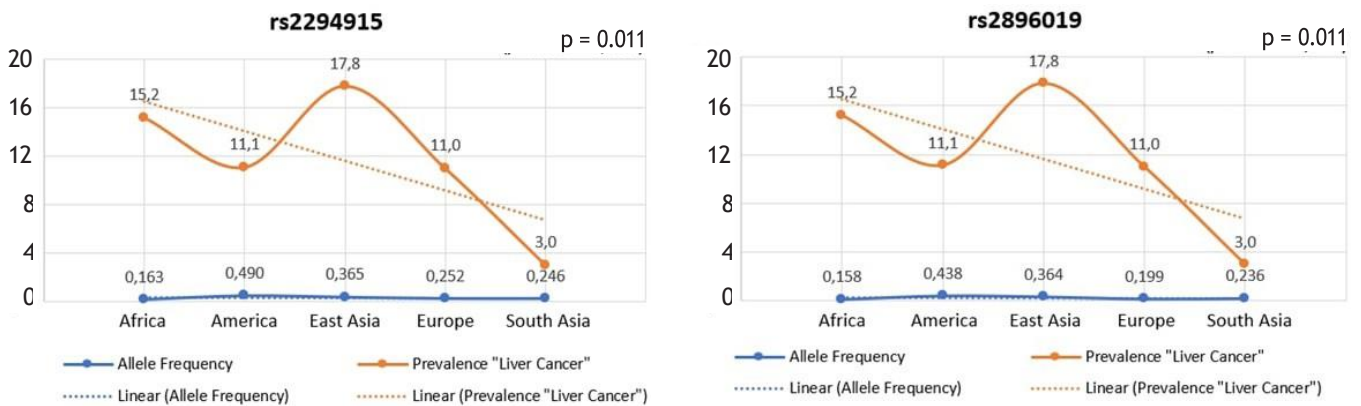


Fig. 5. The association between allele frequency and the prevalence of liver cancer on each continent.

allele at rs2296019 is considerably higher in African (84%), European (80%), South Asian (76%), East Asian (64%), and American (56%) populations. Next, we evaluated the association between allele frequency and the prevalence of liver cancer on each continent. Data on liver cancer prevalence were obtained from Li et al.

[12,18]. In this context, two SNPs (rs2294915 and rs2896019) were found to be positively correlated with the prevalence rate of liver cancer across multiple continents (Africa, America, East Asia, Europe, South Asia), as determined by Pearson's correlation analysis ( $p = 0.011$ ) (Fig. 5). Populations with higher frequencies of

variant alleles of these polymorphisms are thought to have a higher prevalence of liver cancer. We highlighted that these two variants (rs2294915 and rs2896019) are more frequent in East Asian and African populations, which exhibit higher aggressiveness of liver cancer compared to America, Europe, and South Asia. This study suggests that individuals in East Asian and African populations carrying the variant alleles rs2294915 and rs2896019 may be more susceptible to liver cancer.

Patients with liver cancer who also have a history of alcohol abuse, consuming  $\geq 3$  drinks per day, have a 16% increased risk of developing liver cancer compared to the general population. Additionally, individuals with diabetes and those with central obesity are at twice the risk of developing liver cancer [1]. The diagnosis of liver cancer typically involves serological testing combined with imaging techniques, which is the standard approach for detecting liver carcinoma. However, the sensitivity of the commonly used serological test, which is designed to detect alpha-fetoprotein, is only about 60%. Imaging modalities such as magnetic resonance imaging, computed tomography, and ultrasonography demonstrate high levels of sensitivity and specificity in detecting liver cancer, especially in patients with liver cirrhosis [19].

Variant alleles (rs2294915 and rs2896019) are associated with liver cancer. Populations from Africa, America, East Asia, Europe, and South Asia exhibit associated *PNPLA3* expression, which leads to an increased susceptibility to liver cancer. The identification of unique and pathogenic gene variations for a disease is of great interest for both research and clinical validation. These variants provide insights into disease susceptibility and also act as potential diagnostic and prognostic biomarkers [20]. Furthermore, they can aid in the identification of drug target candidates, an approach referred to as genomic-driven drug repurposing [21]. We expect that the discovery of candidate gene variations in *PNPLA3* will facilitate successful clinical validation, potentially establishing it as a promising diagnostic and prognostic biomarker for liver cancer.

It is important to acknowledge that the genetic variants identified in this study as potentially pathogenic are based on preliminary investigations using genomic and bioinformatics databases. While these findings provide crucial insights for future researchers aiming to validate these genetic variants in liver cancer patients, it is important to proceed with caution. We strongly recommend that future research includes additional functional annotations to aid in the prioritization of these pathogenic genetic variants.

This study identified genetic variants that influence liver cancer, highlighting the importance of the *PNPLA3* gene in liver tissue. Consequently, these population groups exhibit varying susceptibilities to liver cancer based on the associated *PNPLA3* expression

levels. The observed variations in allele frequencies of the two identified variants, rs2294915 and rs2896019, across populations from Africa, America, East Asia, Europe, and South Asia, significantly impact *PNPLA3* gene expression. Our study also demonstrated that these two SNPs (rs2294915 and rs2896019) were positively correlated with the prevalence rate. The positive association of prevalence rates was more frequently observed in East Asian and African populations. The higher the frequency of the variant alleles of these polymorphisms in a population, the higher the estimated prevalence rates. The variants investigated in this study are likely to predispose individuals to liver cancer and could play a role in its progression and aggressiveness. These findings highlight the critical importance of understanding genomic variations for precision medicine and for designing targeted screening strategies for liver cancer across diverse populations on different continents.

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Conceptualization: MM (Muhammad Ma'ruf), LMI, WA. Data curation: MM (Muhammad Ma'ruf), LMI. Formal analysis: MM (Muhammad Ma'ruf), LMI, WA. Methodology: MM (Muhammad Ma'ruf), LMI, WA. Writing – original draft: MM (Muhammad Ma'ruf), LMI. Writing – review & editing: MM (Muhammad Ma'ruf), LMI, WA, BDP, MAS, SK, RC, MM (Maulida Mazaya), LMHS.

## Conflicts of Interest

No potential conflict of interest relevant to this article was reported.

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# Profil Genomics & Informatics Journal

(Sumber: Resurchify.com)



**Note:** The **impact score** or **impact index** shown here is equivalent to the average number of times documents published in a journal/conference in the past two years have been cited in the current year (i.e., Cites / Doc. (2 years)). It is based on Scopus data and can be a little higher or different compared to the impact factor (IF) produced by Journal Citation Report. Please refer to the Web of Science data source to check the exact journal impact factor™ (Thomson Reuters) metric.

## Important Metrics and Factor

<b>Title</b>	Genomics and Informatics
<b>Abbreviation</b>	Genom. Inform.
<b>Publication Type</b>	Journal
<b>Subject Area, Categories, Scope</b>	Ecology, Evolution, Behavior and Systematics (Q2); Genetics (Q3); Health Informatics (Q3)
<b>h-index</b>	10
<b>Overall Rank/Ranking</b>	11649
<b>SCImago Journal Rank (SJR)</b>	0.437
<b>Impact Score</b>	1.03
<b>Publisher</b>	Korea Genome Organization
<b>Country</b>	South Korea
<b>ISSN</b>	22340742
<b>Best Quartile</b>	Q2
<b>Coverage History</b>	2019-2022

## Genomics and Informatics Impact IF 2023 Prediction

Impact IF 2022 of **Genomics and Informatics** is **1.03**. If the same downward trend persists, Impact IF may **fall** in 2023 as well.

## Impact IF Trend

Year wise Impact IF of Genomics and Informatics. Based on Scopus data.



Year	Impact IF
2023/2024	Coming Soon
2022	1.03
2021	1.73
2020	1.47
2019	0.00

## Genomics and Informatics h-index



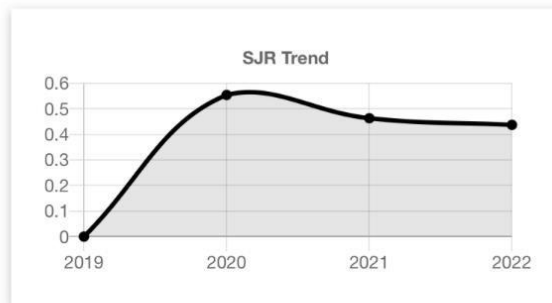
**Genomics and Informatics** has an h-index of **10**. It means **10** articles of this **journal** have more than **10** number of citations. The h-index is a way of measuring the productivity and citation impact of the publications. The h-index is defined as the maximum value of h such that the given journal/author has published h papers that have each been cited at least h number of times.



## Genomics and Informatics Rank and SCImago Journal Rank (SJR)

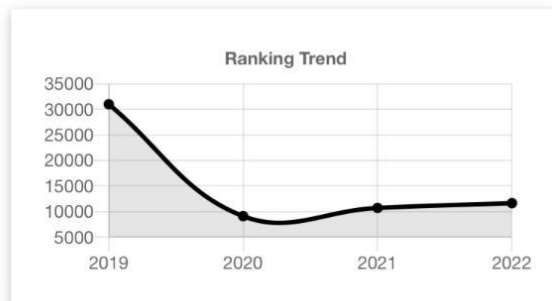
The overall rank of **Genomics and Informatics** is **11649**. According to SCImago Journal Rank (SJR), this journal is ranked **0.437**. SCImago Journal Rank is an indicator, which measures the scientific influence of journals. It considers the number of citations received by a journal and the importance of the journals from where these citations come.

### SJR of Genomics and Informatics by Year



Year	SJR
2023/2024	Coming Soon
2022	0.437
2021	0.463
2020	0.554
2019	

### Ranking of Genomics and Informatics by Year



Year	Ranking
2023/2024	Coming Soon
2022	11649
2021	10713
2020	9112
2019	30992