Research Article

Discovery of Functional SNPs via Genome-Wide Exploration of Malaysian Pigmented Rice Varieties

Rabiatul-Adawiah Zainal-Abidin,1,2 Norliza Abu-Bakar,2 Yun-Shin Sew,2 Sanimah Simoh,2 and Zeti-Azura Mohamed-Hussein1,3

1Centre for Bioinformatics Research, Institute of Systems Biology (INBIOSIS), Universiti Kebangsaan Malaysia (UKM), 43600 UKM Bangi, Selangor, Malaysia
2Malaysian Agricultural Research & Development Institute (MARDI), Persiaran MARDI-UPM, 43300 Serdang, Selangor, Malaysia
3Centre for Frontier Sciences, Faculty of Science & Technology (FST), Universiti Kebangsaan Malaysia (UKM), 43600 UKM Bangi, Selangor, Malaysia

Correspondence should be addressed to Zeti-Azura Mohamed-Hussein; zeti.hussein@ukm.edu.my

Received 1 March 2019; Revised 1 August 2019; Accepted 19 August 2019; Published 10 October 2019

Academic Editor: Corey Nislow

Copyright © 2019 Rabiatul-Adawiah Zainal-Abidin et al. This is an open access article distributed under the Creative Commons Attribution License, which permits unrestricted use, distribution, and reproduction in any medium, provided the original work is properly cited.

Recently, rice breeding program has shown increased interests on the pigmented rice varieties due to their benefits to human health. However, the genetic variation of pigmented rice varieties is still scarce and remains unexplored. Hence, we performed genome-wide SNP analysis from the genome resequencing of four Malaysian pigmented rice varieties, representing two black and two red rice varieties. The genome of four pigmented varieties was mapped against Nipponbare reference genome sequences, and 1.9 million SNPs were discovered. Of these, 622 SNPs with polymorphic sites were identified in 258 protein-coding genes related to metabolism, stress response, and transporter. Comparative analysis of 622 SNPs with polymorphic sites against six rice SNP datasets from the Ensembl Plants variation database was performed, and 70 SNPs were identified as novel SNPs. Analysis of SNPs in the flavonoid biosynthetic genes revealed 40 nonsynonymous SNPs, which has potential as molecular markers for rice seed colour identification. The highlighted SNPs in this study show effort in producing valuable genomic resources for application in the rice breeding program, towards the genetic improvement of new and improved pigmented rice varieties.

1. Introduction

Rice (Oryza sativa L.) is the most crucial staple food crops in Asian countries. The most consumed rice is white rice, which resulted from the white pericarp. The coloured pericarp such as black, red, and brown has become more popular. Coloured pericarp accumulates secondary metabolites such as flavonoid, anthocyanin, and proanthocyanidin and usually are associated as potent antioxidants. Previous study has found that food sources with high antioxidant properties can lower the risk of chronic diseases such as type II diabetes, cardiovascular disease, and cancers [1]. Hence, this finding has accelerated the development of pigmented rice varieties.

Previous efforts have been performed to elucidate the genetic basis of black and red rice varieties [2–4]. In red rice variety, Rc is responsible for the accumulation of proanthocyanidins in red pericarp, but it has to interact with Rd gene that encodes for dihydroflavonol-4-reductase (DFR) that involved in the catalysis activity of dihydroflavonol to leucoanthocyanidin [2, 3]. However, without this interaction, brown rice will be produced whilst Rd alone has no phenotype change. Rc is also known as domestication gene [5] and has been widely used to investigate the domestication process in rice subspecies [6–8]. Kala4, a transcription factor in basic helix-loop-helix (bHLH) family, is involved in black rice pigmentation [4]. Ectopic expression in Kala4 causes the upregulation of LDOX in pericarp, accumulates the anthocyanin, and produces black pericarp [4].

To further investigate the genetic basis of pigmented rice varieties, many efforts have been performed using omics
technologies and bioinformatics. For instance, several studies on the phytochemical diversity of the coloured or pigmented rice from landraces, varieties, and wild relatives have been widely conducted using a metabolomics approach to reveal their antioxidant properties and variabilities [9–14]. Previous studies on the transcriptome sequencing of pigmented rice varieties were conducted to identify single-nucleotide polymorphisms (SNPs) and regulatory genes, which might be responsible in the accumulation of anthocyanin [15, 16]. An integrative omics approach, combining proteomics and transcriptomics sequencing, was conducted to identify the flavonoid biosynthetic genes in the black and red rice varieties [17] and potential biomarkers responsible to the accumulation of flavonoid in rice varieties by linking the SNP located in the flavonoid biosynthetic genes to flavonoid accumulation [18]. Meanwhile, genome resequencing of pigment rice varieties has been performed to identify potential SNPs located in the biosynthetic genes, which can be developed as molecular markers for nutritional quality traits such as high antioxidant [19, 20] and high amylose content [21].

All these efforts showed the importance of mining genetic variant, biosynthetic genes, and transcription factors in order to understand the interactions that will affect and influence the biosynthesis of antioxidant contents in rice varieties.

Molecular marker is a DNA fragment with phenotypic expression that is associated with a certain location within the genome [22]. Several types of molecular markers such as random amplified polymorphic DNA (RAPD), restriction fragment length polymorphism (RFLP), and microsatellite (SSR) are widely used in the genetic improvement of rice [23]. Recently, the application of SNP in rice breeding improvement is rapidly expanding. The combinatorial approach between the next-generation sequencing technology (NGS) and bioinformatics has greatly assisted SNPs discovery from the genome, followed by the validation of SNPs conducted using current genotyping technology [24]. Thus, the application of bioinformatics in predicting SNPs from the genome sequences is crucial to accelerate the implementation of genome-based breeding approaches for the development of rice varieties with desirable agronomical traits [25].

SNP is defined as a single base difference in DNA sequence and the most common type of genetic variation to distinguish individuals [26]. The abundance of SNPs in the genome can be used in the improvement of high-resolution genetic map that will lead to the association of SNP with agronomic traits of interest [27]. Interestingly, SNPs located in the genic region could affect the phenotypic expression of crops and are applicable for gene functional analysis and marker-assisted selection (MAS) [28]. SNPs have been applied to investigate the evolution and domestication of rice [29–31] and the identification of functional SNP in genes related to various agronomic traits such as domestication trait [32], seed size [33], salinity tolerance [34] and response to stress [35], diversity analysis among cultivars [36–39], and seed purity assessments [40]. These efforts showed the utilisation of SNP for rice breeding improvement. However, not much effort has been conducted to explore the genetic variation in Malaysian pigmented rice varieties using single-nucleotide polymorphism (SNP). As a result, this has to limit genetic understanding of pigmented rice that is crucial for the genetic improvement of pigmented rice varieties.

Here, we report the genome-wide SNP analysis on the whole genome resequencing of two black rice varieties (Bali and Pulut Hitam 9) and two red rice varieties (MRM16 and MRQ100). A landrace rice variety, while Pulut Hitam 9 (PH9), MRM16, and MRQ100 are modern rice varieties. All of them were from indica subspecies. These four varieties were chosen due to their nutritional trait that was enriched with antioxidant properties [14]. Figure 1 shows the whole grains of Bali, Pulut Hitam 9, MRM16, and MRQ100.

We mined the SNPs from the genomes of four pigmented Malaysian rice varieties to search for the SNPs with polymorphic sites and candidate SNPs associated with the flavonoid biosynthetic genes. Additionally, we have identified 70 novel SNPs after comparing with SNP data from Ensembl Plants variation [41], comprising the variation data from six large-scale SNP studies. The SNPs highlighted in this study are suggested as potential molecular markers for further validation using a genotyping platform, towards genetic improvement of pigmented rice varieties.

2. Materials and Methods

2.1. Plant Materials. Plant materials consisted of four pigmented rice varieties from Malaysian, i.e., Bali, PH9, MRM16, and MRQ100. Four varieties were selected based on (a) the presence of high antioxidant contents and (b) released variety. Seeds of Bali, PH9, MRM16, and MRQ100 were obtained from MARDI Seberang Perai, Penang, Malaysia. Seeds were sterilized, incubated at 42°C overnight, and soaked in water for two days before being placed onto wet tissues or directly sowed into the soil.

2.2. DNA Isolation and Genome Sequencing. Total DNA of each variety was extracted from leaves of two-week-old germinated seedling using Mutou et al.’s protocol [42] and Sigma DNA extraction kit. DNA quality and quantity were analysed using NanoDrop spectrophotometer. The integrity of DNA samples was determined using 0.8% agarose gel. The DNA samples were sequenced using Illumina HiSeq 4000 sequencing (Illumina, Inc., San Diego, CA, USA). Standard Illumina protocol was used for the sequencing process.

2.3. Reads Mapping and Identification of SNPs. The pair-end sequencing reads from Bali, PH9, MRM16, and MRQ100 with the read length of 150 bp at each end were aligned with Nipponbare genome sequences [43] using Burrows-Wheeler Aligner (BWA) [44] software using default parameters except for “-mem -m 10000 -o 1 -e 10 -t 4”. All genomes were individually aligned. The mapped reads were merged and indexed as BAM files. The mapped reads from each variety were then processed for mark duplicate reads, fixing mate-pair information, and adding or replacing read groups using PICARD version 0.7.12.

We followed the GATK best-practices pipeline for SNP calling [45]. This SNP-calling pipeline has been used in rice SNP discovery [31, 34, 46, 47] and development of SNP panel using genotyping platforms [48–50]. Local realignment and
mapping quality, (b) >30 for mapping quality, (b) >50 for variant quality, and (c) >10 for the number of supporting reads for every base. Another two criteria also were performed after SNPs calling, i.e., (i) distance between SNP and another SNP is >150bp and (ii) SNP with a PASS score.

2.4. Annotation and Functional Classification of SNPs. SnpEff [53] version 4.1 was used to annotate SNPs into intergenic and genic. The genic SNPs were classified as coding sequences (CDS), untranslated region (UTR), and intron. SNPs in the CDS region were further divided into synonymous and nonsynonymous amino acid substitutions. Annotated SNPs were filtered accordingly with reference to the above criteria using R packages (dplyr, sqldf, and tidyr). Genomic distribution of SNPs was performed using R scripts and visualised using Flapjack [54]. Unique SNP in each variety was extracted using R scripts. The number of SNPs in CDS was counted using R scripts.

2.5. Enrichment Analysis. Gene ontology enrichment analysis of genes containing 622 SNPs with polymorphic sites was performed using PANTHER (protein annotation through evolutionary relationship) classification system [55] (http://www.pantherdb.org) with FDR cutoff selected at ≤0.05. Gene Ontology database for Oryza sativa was selected for this analysis.

2.6. Identification of SNP Genes Involved in the Flavonoid Biosynthetic Genes (FBGs). The flavonoid biosynthetic genes (FBGs) were obtained from the similarity and bibliomic search. The list of FBGs is provided in the Supplementary Dataset S1. Genic SNPs from each variety were compared to the flavonoid biosynthetic genes by matching with the Oryza sativa gene identification (OsID) using R scripts.

3. Results and Discussion

3.1. Mapping of Bali, PH9, MRM16, and MRQ100 Genome Data onto the Nipponbare Reference Genome. Genome sequencing of Bali, PH9, MRM16, and MRQ100 has produced 101.71, 99.98, 98.76, and 99.99 million reads, respectively. The average read lengths of 2 × 150 bp were generated with 30× depth of sequencing. This 30× depth of sequencing was chosen as it provides sufficient coverage in identifying high-quality genetic variations such as SNP, single-nucleotide variation (SNV), and insertion-deletion (InDel) [56]. Therefore, the relationship between the depth of sequencing and identification of SNPs is a key factor in obtaining high-quality SNPs. A total of 96.47% of Bali, 95.97% of PH9, 98.07% of MRM16, and 94.42% of MRQ100 million clean reads was obtained after the sequence read cleaning process. The clean reads for each variety were then mapped against the Nipponbare reference genome. Nipponbare was used as a reference genome because it is well-assembled and annotated genome [34, 35, 57]. The mapped reads against Nipponbare genome showed that almost 96% of the reads were successfully mapped onto the rice genome. Low divergence of genetic differences between indica and japonica varieties might be a contributing factor that caused the highest mapped rate. Table 1 represents a summary of the sequence reads and mapping data in four pigmented rice varieties.

3.2. Identification of SNPs and SNPs with Polymorphic Sites. Table 2 provides statistics of raw and high-quality SNPs for Bali, PH9, MRM16, and MRQ100 genome. MRM16 contained the highest variation among the genomes, suggesting that MRM16 has a distant relationship to Nipponbare.

Figure 2 shows the distribution of 662 SNPs with polymorphic sites on 12 rice chromosomes. SNPs with polymorphic sites are defined as the presence of SNP in the individual but with several different alleles. A set of SNPs with polymorphic sites indicates that the SNP is highly informative, thus suitable as a potential candidate for genetic marker development [58]. Supplementary Figure 1 shows the character of SNPs with polymorphic sites.

Distribution of these polymorphic sites on the 12 rice chromosomes shows that chromosome 11 consisted of the highest number of SNPs with polymorphic sites (82), followed by chromosome 1 (80) and chromosome 2 (80).
These values demonstrate the random distribution of SNPs with polymorphic sites within the 12 rice chromosomes. Interestingly, 70/10% of the SNPs with polymorphic sites were novel SNPs based on the comparison against *Oryza sativa* Ensembl Plants variation database as of October 2017 (Figure 1). The SNP datasets in the *Oryza sativa* Ensembl Plants variation were from six large-scale SNP studies [59–64]. This finding indicates that many SNPs have been discovered from various rice cultivars by rice genome-sequencing effort from time to time. The 70 novel SNPs with polymorphic sites can be suggested as molecular markers for varietal identification.

### 3.3. Annotation of SNPs and SNPs with Polymorphic Sites

The annotation of SNPs in four pigmented rice varieties has revealed that most of the SNPs were located in the intergenic regions. The numbers and types of SNPs vary across different rice cultivars (Table 2). The distribution of SNPs across the rice chromosomes is shown in Figure 2. Of these, 70 novel SNPs (10%) were detected when compared against *Oryza sativa japonica* Ensembl Plants variation database.

Table 1: Summary of sequence reads and mapping statistics in Bali, PH9, MRM16, and MRQ100 genome.

|            | Bali             | PH9             | MRM16            | MRQ100            |
|------------|------------------|-----------------|------------------|-------------------|
| Total reads (bp) | 101,710,572     | 99,980,328      | 98,764,058       | 99,998,624        |
| Number of clean reads (bp) | 99,865,228 (98.18%) | 99,380,446 (99.40%) | 98,078,122 (99.30%) | 94,428,632 (99.43%) |
| Genome coverage (30×) | 88.59%          | 88.45%          | 88.45%           | 88.49%           |
| Total mapped reads | 96,479,796      | 95,971,696      | 94,870,967       | 91,170,844        |
| Percentage of total mapped reads | 96.61%          | 96.57%          | 96.73%           | 96.55%           |

Table 2: Summary of SNP identification and annotation in Bali, PH9, MRM16, and MRQ100 when compared against Nipponbare reference genome. The number of total annotated SNPs was higher than the total number of quality SNPs due to more than one annotation in a single SNP.

|                | Bali       | PH9        | MRM16       | MRQ100       | Total       |
|----------------|------------|------------|-------------|--------------|-------------|
| Number of raw SNPs | 2,394,592  | 2,227,819  | 2,740,764   | 2,380,079    | 9,743,254   |
| Number of high-quality SNPs | 436,322    | 412,791    | 469,782     | 435,382      | 1,754,277   |
| Intergenic SNPs | 328,261    | 310,712    | 349,786     | 327,021      | 1,315,780   |
| Genic SNPs      | 149,232    | 140,677    | 165,124     | 149,903      | 604,936     |

![Figure 2: Distribution of 662 SNPs with polymorphic sites on 12 rice chromosomes. Of these, 70 novel SNPs (10%) were detected when compared against *Oryza sativa japonica* Ensembl Plants variation database.](image-url)
region (1,315,780; 64%) while fewer SNPs are located within the genic region (604,936; 29%) (Table 2). This finding corroborates with the results obtained by Tatarinova et al. where the SNP rate is higher in the intergenic regions compared to that in the genic regions [65]. This finding is common in SNP discovery as the coding regions are more conserved than intergenic regions [65].

Analysis of the SNP differences between rice varieties showed that MRM16 (165,124) has a higher number of SNPs in the genic region whereas PH9 (140,677) has the least number of SNPs in the genic region. High number of SNPs in the genic region of MRM16 suggested the introgression, and recombination have occurred through human-guided artificial selection during rice breeding activity. Previous studies by Sang et al. and Tatarinova et al. suggested that artificial selection in developing modern rice varieties has shaped the present of SNP frequency and gene pool in the rice genome [5, 65].

Functional annotation analysis was performed to explore the effect of 662 SNPs with polymorphic sites on gene function. SNPs with polymorphic sites in the genic region will be valuable if associated with phenotypic expression or important agronomical trait [28]. Enrichment analysis based on the Gene Ontology (GO) terms was conducted on the 662 SNPs with polymorphic sites for functional annotation towards investigating their effect on the gene function. The top ten GO terms from biological processes and molecular function terms have been chosen for further discussion (Table 3).

GO:0009987 (cellular process) and GO:0008152 (metabolic process) were assigned for all genes that carry the SNPs with polymorphic sites in Bali, PH9, MRM16, and MRQ100 varieties suggesting their involvement in various physiological functions. Cellular process plays essential roles in cell communication while the metabolic process involved in the anabolism and catabolism of biosynthesis pathway. In the molecular function category, the SNPs with polymorphic sites were assigned to the binding function (heterocyclic compound binding, organic cyclic compound binding, ion binding, small molecule binding, and carbohydrate derivative binding) and catalytic activity suggesting their possible involvement in the formation of molecule and enzymatic activities related to abiotic stress [34], several biochemical pathways, and disease trait [66].

The biological interpretation of genes in the SNPs with polymorphic sites was further examined using the information obtained from the Reactome pathway analysis [67]. In total, three major pathways were found to be correlated with the top 10 GO terms, such as metabolism and regulation (R-OSA-2744345), secondary metabolite biosynthesis (R-OSA-2744341), and hormone biosynthesis, signalling, and transport (R-OSA-2744341) (Table 3). This finding corroborates with a study by Lin et al. that most of the SNPs and genes in the pigmented rice varieties were abundant in metabolic pathways such as flavonoid and anthocyanin biosynthetic pathways [19]. Hence, SNPs with polymorphic sites and genes in the pigmented rice genome might play an important role in the production of anthocyanin and proanthocyanidin. Our finding confirms the existence of phenotypic characteris-

cic in pigmented rice (Bali, PH9, MRM16, and MRQ100) that are highly abundant with their antioxidant properties [14].

Functional annotation of the SNPs with polymorphic sites was further conducted using Pfam analysis on the 23 nonsynonymous SNPs (nsSNPs). Usually, nonsynonymous SNPs can affect the function of a gene to encode for the right protein, hence will affect its function. 13 nsSNPs were assigned into several functional gene classifications such as metabolism, stress response, and transporter and 10 nonsynonymous SNPs were assigned to the domain of unknown function (DUF). Table 4 shows the annotation of 13 nsSNPs into their gene classifications.

Parida et al. discovered the involvement of Os01g0128000, Os07g0117000, Os09g0314200, Os10g0371100, and Os11g0539000 genes in plant resistance, pathogenesis, and abiotic stress mechanism [66]. Our analysis has identified that all the above genes have one nsSNP while two nsSNPs were found in Os01g0147001 that encodes for glycosyltransferase family 43 enzymes (important in the biosynthesis of cell wall [68] and Os02g0503900 that encodes for a cytochrome P450 (involved in xylan biosynthesis [69], two nsSNPs were also found in Os06g0695800 that encoded for ATP-binding cassette (ABC) transporter genes (important in iron intake for the improvement of plant micronutrient content [70] and were involved in the transportation of molecules, secondary metabolites, and plant hormones [71]). Further investigation on these genes is recommended to reveal the specific role of these variants in plant development and defence system.

Besides, four nsSNPs were also identified in four transcription factor families such as Myb-like DNA-binding domain (Os01g0128000), AP2 domain (Os10g0371100), IQ calmodulin-binding motif (Os07g0562800), and SWItch/Sucrose Non-Fermentable (SWI/SNF2) family N-terminal domain (Os08g0180300). Interestingly, Os01g0128000 that encodes for the Myb-like DNA-binding domain has been identified to be involved in the uptake and higher accumulation of phosphate (Pi) [72]. In particular, this gene was observed as a regulator in the cross-talk between nutrient signalling and phytohormone signalling pathway. Li et al. has reported that Os08g0180300 encodes for SWI/SNF2 and it is able to suppress rice innate immunity thus remarkably important in the defence mechanism against pathogen attack [73]. Hence, variation in these genes might affect the disease resistance capability of rice.

On the contrary, not much study has been conducted to confirm the function of Os10g0371100 that encodes for the ethylene-responsive transcription factor (ERF) domain or AP2/ERF domain. However, Os10g0371100 is predicted to be involved in plant growth and development either as an activator or a repressor in the expression of stress-responsive genes that are related to the abiotic stress responses [74]. Similarly, not much work has been conducted on the function of Os07g0562800 that encodes for the IQ calmodulin-binding motif in rice. Nevertheless, this gene was predicted to play a role in regulating plant responses in the signal transduction pathway during biotic or abiotic stress condition [75]. Analysis of SNPs with polymorphic sites can facilitate the identification of candidate SNPs and genes for
functional markers in traits related to nutritional, nutraceutical and disease that can be used in the marker-assisted selection (MAS) of pigmented rice varieties.

3.4. Identification of SNPs Associated with Flavonoid Biosynthetic Genes (FBGs). Pigmented rice is significantly associated with higher antioxidant content due to the presence of anthocyanin and proanthocyanidin. The production of these secondary metabolites is controlled by a set of flavonoid biosynthetic genes such as DFR, LAR, ANR, UGT, and LDOX, which lead to the production of anthocyanin and proanthocyanidin. The difference between anthocyanin and proanthocyanidin synthesis is the inclusion of the catalysed enzymes LAR and ANR for proanthocyanidin, while catalysis...
of LDOX for anthocyanin. Besides, Kal44 gene activates LBG to produce anthocyanin whilst Rc gene activates DFR to produce proanthocyanidin. Rc is unable to regulate the production of proanthocyanidin alone; instead, it requires the presence of Rd gene which encodes DFR to activate the accumulation of proanthocyanidin.

In this study, a total of 99 flavonoid biosynthetic genes (FBGs) were selected from Nipponbare genome using similarity and bibliomic search [76–81]. Supplementary Table 1 shows the list of 99 FBGs into three groups, i.e., (i) general phenylpropanoid (phenylalanine ammonia-lyase (PAL); cinnamic acid 4-hydroxylase (C4H); 4-coumarate CoA ligase (4CL)); (ii) early biosynthetic genes (EBG) (chalcone synthase (CHS); chalcone isomerase (CHI); flavanone 3-hydroxylase (F3H); flavanone 3′-hydroxylase, F3′H); and (iii) late biosynthetic genes (LBG) (dihydroflavonol reductase (DFR); leucoanthocyanidin reductase (LAR); UDP-glucose flavonoid 3-O-glucosyl transferase (UGT); leucoanthocyanidin oxidase (LDOX)) [82, 83]. Three transcription factors involved in the production of anthocyanin and proanthocyanidin were selected, i.e., R2R3-MYB, Kal44, and Rc. R2R3-MYB (Os06g0205100) due to their role in activating the DFR gene in the upstream biosynthesis [84, 85]. Kal44 (Os04g0557500) encodes for a basic helix-loop-helix (bHLH) transcription factor, which plays a role in activating the LDOX gene in the regulation of black pigmentation [4]. Rc (Os07g0211500) has previously been shown as an activator for Rd (Os01g0633500) in the production of red pigmentation [2, 3].

A total of 1649 genic SNPs were found in the flavonoid biosynthetic genes, and 511 SNPs were identified in the genes related to the general phenylpropanoid, 463 SNPs in EBGs and 675 SNPs in LBGs (Table 5). A high number of variations was found in LBG due to a difference in patterns of evolutionary rate. A previous study has revealed that the upstream genes have been observed to evolve slower than downstream genes in the secondary metabolite biosynthesis [86]. A similar pattern has been observed in mango with a high number of variations in the downstream genes of the flavonoid biosynthetic pathway [87]. This finding suggests that mutations in the flavonoid biosynthetic genes could affect the accumulation of secondary metabolite end products such as anthocyanin and proanthocyanidin.

Interestingly, ten genic SNPs associated with UGT (Os02g0589400) were identified in this analysis. A previous study has reported that one SNP was strongly associated with UGT (Os02g0589400) and was suggested as a metabolite quantitative trait loci (mQTL) for antioxidant trait [88]. UDP-glucose flavonoid 3-O-glucosyl transferase (UGT) is an enzyme involved in the glycosylation process and is essential for pigment stabilisation and secondary metabolites storage [77]. For this reason, the variation in UGT might provide the possibility of finding the candidates for functional markers in the accumulation of antioxidant. However, further investigation is required to determine the actual function of these SNPs.

Two genic SNPs associated with UGT (Os01g0736300) at position 30712175 (chr01_30712175) and 30713739 (chr01_30713739) have been identified and were found as SNPs in the untranslated (UTR) region and CDS, respectively. This finding suggests that the mutation in the UGT can be used as potential genetic markers for the accumulation of antioxidant properties in the pigmented rice varieties as Dong et al. found that a mutation in Os01g0736300 was associated with 7-O-glycosylated flavonoids [18]. Furthermore, SNP (chr01_30713739) was predicted as a nonsynonymous SNP that is involved in amino acid substitution and might affect the protein function that leads to the phenotypic consequences.

In addition, there were 160 genic SNPs found in the transcription factor genes, i.e., 30 mutations in Rc (Os07g0211500), 38 mutations in R2R3-MYB genes, and 92 mutations in Kal44 (Os04g0557500). In comparison to the number of SNPs in the structural genes, fewer SNPs were found in the transcription factor, and this finding suggests that the character of the transcription factors are highly conserved compared to other classes of genes [89]. In conclusion, polymorphism in the transcription factor plays a crucial role in the biosynthetic pathway as it is responsible for regulating the functions of biosynthetic genes and affecting the production of secondary metabolites [86, 87].

3.5. Comparative Analysis on Genic SNPs in Flavonoid Biosynthetic Genes among Bali, PH9, MRM16, and MRQ100. This study also investigated the distribution of genic SNPs in four pigmented rice varieties. A total of 448, 420, 491, and 459 genic SNPs were identified in Bali, PH9, MRM16, and MRQ100, respectively (Figure 3). Of these, 94, 89, 103, and 88 nonsynonymous SNPs (nsSNPs) were identified from Bali, PH9, MRM16, and MRQ100, respectively (Figure 3).

SNPs are considered unique if they are present in one variety but absent in the other three varieties (Supplementary Figure 1). Hence, unique SNPs can be used to investigate the relationship between accessions and varieties [50]. In this study, a total of 40 nsSNPs in 39 flavonoid biosynthetic genes and one transcription factor was found unique to all four accessions (Figure 4 and Supplementary Table 2). Supplementary Table 2 provides list of 40 nsSNPs and their SNPs information (i.e., SNP identifier (SNP ID), gene identifier, reference allele, SNP allele, chromosome, and SNP position).

The proportion of unique nsSNPs in these four varieties is lower, which is 10%. This finding suggests that these four varieties might share a common ancestor and may share similar genetic characteristics. The impact of unique variants has been demonstrated in wild strawberry where the occurrence of the genetic changes has caused the yellow colour phenotypic differences in three strawberry accessions [50].

Four unique nsSNPs (m_UGT_12, m_UGT_13, b_UGT_6, and b_UGT_1) were identified at positions 26199225, 26199416, 26199448, and 26199529 in UGT (Os05g0527000), respectively, and one nsSNP (b_UGT_2) which occurred at position 10479849 in UGT (Os06g0288300) (Figure 4). Os05g0527000 and Os06g0288300 that encoded for UGT have been reported as potential markers to distinguish different accumulations of flavonoid in Indica subspecies [88]. Finally, one nonoverlap
nsSNP has been found in Os01g0305900 that encodes for R2R3-MYB (b_MYB_1), which is a transcription factor, and this unique nsSNP can only be found in the black rice variety Pulut Hitam 9. This unique nsSNP can be used as a potential genetic marker for rice seed colour identification.

Genomic variation among these four pigmented rice varieties provides a resource for genetic variability as well as generating new allelic variants towards the development of new and improved pigmented rice varieties. However, SNP validation must be conducted using a genotyping platform. This genome-wide gene-based SNP marker identification can provide a solution for breeders to effectively screen diverse accessions or interspecific hybrid breeding program for the genetic improvement in pigmented rice varieties.
4. Conclusions

Extensive bioinformatic analysis on next-generation sequencing (NGS) data has contributed to the identification of a high number of SNPs. From this study, the candidate SNPs associated with the essential functional genes and SNPs with polymorphic sites provide important insights into the genetic basis of four Malaysian pigmented rice varieties. Therefore, a genotyping experiment can be conducted on these SNPs for validation before progressing into genetic diversity study, cultivar identification, and marker-assisted selection (MAS), towards the development of new and improved pigmented rice varieties.

Data Availability

The raw sequencing reads data used to support the findings of this study have been deposited in the ENA database (https://www.ebi.ac.uk/ena). Accession numbers are ERR2831548(Bali), ERR2831549(PH9), ERR2831551(MRM16) and ERR2831550(MRQ100).

Conflicts of Interest

The authors declare no conflict of interest.

Acknowledgments

This work was supported by the MARDI Pembangunan project (P21003004010001-I) in collaboration with the Institute of Systems Biology, Universiti Kebangsaan Malaysia. The authors would like to thank Dr. Habibuddin Hashim for his constructive comments. The first author would like to thank MARDI for her PhD scholarship.

Supplementary Materials

Supplementary Table 1: list of 99 flavonoid biosynthetic genes. Supplementary Table 2: list of nonsynonymous SNPs of 16 flavonoid biosynthesis genes in four pigmented rice varieties (Bali, PH9, MRM16, and MRQ100). Supplementary Figure 1: unique SNP shows the allele present in one variety whilst SNPs with polymorphic sites show the presence of SNP in each variety but with several allele combinations. (Supplementary Materials)

References

[1] P. Goufo and H. Trindade, "Rice antioxidants: phenolic acids, flavonoids, anthocyanins, proanthocyanidins, tocopherols, tocotrienols, γ-oryzanol, and phytic acid," Food Science & Nutrition, vol. 2, no. 2, pp. 75–104, 2014.
[2] T. Furukawa, M. Maekawa, T. Oki et al., "The Rc and Rd genes are involved in proanthocyanidin synthesis in rice pericarp," The Plant Journal, vol. 49, no. 1, pp. 91–102, 2007.
[3] M. T. Sweeney, M. J. Thomson, B. E. Pfeil, and S. McCouch, "Caught red-handed: Rc encodes a basic helix-loop-helix protein conditioning red pericarp in rice," The Plant Cell, vol. 18, no. 2, pp. 283–294, 2006.
[4] T. Oikawa, H. Maeda, T. Oguchi et al., “The birth of a black rice gene and its local spread by introgression,” The Plant Cell, vol. 27, no. 9, pp. 2401–2414, 2015.
[5] T. Sang and S. Ge, "Understanding rice domestication and implications for cultivar improvement," Current Opinion in Plant Biology, vol. 16, no. 2, pp. 139–146, 2013.
[6] Y. Cui, B. K. Song, L.-F. Li et al., “Little white lies: pericarp color provides insights into the origins and evolution of southeast Asian weedy rice,” Genes Genomes Genetics, vol. 6, no. 12, pp. 4105–4114, 2016.
“Japonica rice by whole-genome sequencing,” *Plant and Cell Physiology*, vol. 52, no. 2, pp. 274–282, 2011.

[38] I.-S. Jeong, U. H. Yoon, G. S. Lee et al., “SNP-based analysis of genetic diversity in anther-derived rice by whole genome sequencing,” *Rice*, vol. 6, no. 1, p. 6, 2013.

[39] Y. Arai-Kichise, Y. Shiwa, K. Ebana et al., “Genome-wide DNA polymorphisms in seven rice cultivars of *Temperate* and *Tropical Japonica* groups,” *PLoS One*, vol. 9, no. 1, article e86312, 2014.

[40] B. C. Y. Collard and D. J. Mackill, “Marker-assisted selection: an approach for precision plant breeding in the twenty-first century,” *Philosophical Transactions of the Royal Society B: Biological Sciences*, vol. 363, no. 1491, pp. 557–572, 2007.

[41] S. E. Hunt, W. McLaren, L. Gil et al., “Ensembl variation resources,” *Database*, vol. 2018, article bay119, 2018.

[42] C. Mutou, K. Tanaka, and R. Ishikawa, “DNA extraction from rice endosperm (including a protocol for extraction of DNA from ancient seed samples),” in *Cereal Genomics: Methods and Protocols, Methods in Molecular Biology*, vol. 1099, R. Henry and A. Furtado, Eds., pp. 7–15, Humana Press, Totowa, NJ, USA, 2014.

[43] H. Sakai, S. S. Lee, T. Tanaka et al., “Rice annotation project database (RAP-DB): an integrative and interactive database for rice genomics,” *Plant and Cell Physiology*, vol. 54, no. 2, article e6, 2013.

[44] H. Li and R. Durbin, “Fast and accurate short read alignment with Burrows-Wheeler transform,” *Bioinformatics*, vol. 25, no. 14, pp. 1754–1760, 2009.

[45] G. A. van der Auwera, M. O. Carneiro, C. Hartl et al., “From FastQ data to high-confidence variant calls the genome analysis toolkit best practices pipeline,” *Current Protocols in Bioinformatics*, vol. 43, no. 1, pp. 11.10.1–11.10.33, 2013.

[46] H. B. Mahesh, M. D. Shirke, S. Singh et al., “Indica rice genome assembly, annotation and mining of blast disease resistance genes,” *BMC Genomics*, vol. 17, no. 1, p. 242, 2016.

[47] P. Civiáň, S. Ali, R. Batista-Navarro et al., “Origin of the *Aromatic* group of cultivated rice (*Oryza sativa* L.) traced to the Indian subcontinent,” *Genome Biology and Evolution*, vol. 11, no. 3, pp. 832–843, 2019.

[48] N. Li, H. Zheng, J. Cui et al., “Genome-wide association study and candidate gene analysis of alkalinity tolerance in japonica rice germplasm at the seedling stage,” *Rice*, vol. 12, no. 1, p. 24, 2019.

[49] M. M. Rana, T. Takamatsu, M. Baslam et al., “Salt tolerance improvement in rice through efficient SNP marker-assisted selection coupled with speed-breeding,” *International Journal of Molecular Sciences*, vol. 20, no. 10, p. 2585, 2019.

[50] C. Hawkins, J. Caruana, E. Schiknis, and Z. Liu, “Genome-scale DNA variant analysis and functional validation of a SNP underlying yellow fruit color in wild strawberry,” *Scientific Reports*, vol. 6, no. 1, article 29017, 2016.

[51] Q. Liu, Y. Guo, J. Li, J. Long, B. Zhang, and Y. Shyr, “Steps to ensure accuracy in genotype and SNP calling from Illumina sequencing data,” *BMC Genomics*, vol. 13, article S8, Supplement 8, 2012.

[52] Y. Guo, F. Ye, Q. Sheng, T. Clark, and D. C. Samuels, “Three-stage quality control strategies for DNA re-sequencing data,” *Briefings in Bioinformatics*, vol. 15, no. 6, pp. 879–889, 2014.

[53] P. Cingolani, A. Platts, L. L. Wang et al., “A program for annotating and predicting the effects of single nucleotide polymorphisms, SnPeff: SNPs in the genome of Drosophila melanogaster strain w1118; iso-2; iso-3,” *Fly*, vol. 6, no. 2, pp. 80–92, 2012.

[54] L. Milne, P. Shaw, G. Stephen et al., “Flapjack—graphical genotype visualization,” *Bioinformatics*, vol. 26, no. 24, pp. 3133–3134, 2010.

[55] H. Mi, A. Muruganujan, and P. D. Thomas, “PANTHER in 2013: modeling the evolution of gene function, and other gene attributes, in the context of phylogenetic trees,” *Nucleic Acids Research*, vol. 41, no. D1, pp. D377–D386, 2013.

[56] D. Sims, I. Sudbery, N. E. Iott, A. Heger, and C. P. Ponting, “Sequencing depth and coverage: key considerations in genomic analyses,” *Nature Reviews Genetics*, vol. 15, no. 2, pp. 121–132, 2014.

[57] P. Rathiinasabapathi, N. Purushothaman, R. VI. and M. Parani, “Whole genome sequencing and analysis of Swarna, a widely cultivated indica rice variety with low glycemic index,” *Scientific Reports*, vol. 5, no. 1, article 11303, 2015.

[58] Y. Shavrukov, R. Suchecki, S. Eliby, A. Abugalieva, S. Kenebayev, and P. Langridge, “Application of next-generation sequencing technology to study genetic diversity and identify unique SNP markers in bread wheat from Kazakhstan,” *BMC Plant Biology*, vol. 14, no. 1, p. 258, 2014.

[59] J. Yu, J. Wang, W. Lin et al., “The genomes of *Oryza sativa*: a history of duplications,” *PLoS Biology*, vol. 3, no. 2, article e38, 2005.

[60] K. L. McNally, K. L. Childs, R. Bohnert et al., “Genomewide SNP variation reveals relationships among landraces and modern varieties of rice,” *Proceedings of the National Academy of Sciences of the United State of America*, vol. 106, no. 30, pp. 12273–12278, 2009.

[61] J. L. Goicoechea, J. S. S. Ammiraju, P. R. Marri et al., “The future of rice genomics: sequencing the collective *Oryza* rice,” *Rice*, vol. 3, no. 2-3, pp. 89–97, 2010.

[62] K. Zhao, M. Wright, J. Kimball et al., “Genomic diversity and introgression in *O. sativa* reveal the impact of domestication and breeding on the rice genome,” *PLoS One*, vol. 5, no. 5, article e010780, 2010.

[63] N. Alexandrov, S. Tai, W. Wang et al., “SNP-seek database of SNPs derived from 3000 rice genomes,” *Nucleic Acids Research*, vol. 43, no. D1, pp. D1023–D1027, 2015.

[64] J. Duitama, A. Silva, Y. Sanabria et al., “Whole genome sequencing of elite rice cultivars as a comprehensive information resource for marker assisted selection,” *PLoS One*, vol. 10, no. 4, article e0124617, 2015.

[65] T. V. Tatarinova, E. Chekalin, Y. Nikolsky et al., “Nucleotide diversity analysis highlights functionally important genomic regions,” *Scientific Reports*, vol. 6, no. 1, article 35730, 2016.

[66] S. K. Parida, M. Mukerji, A. K. Singh, N. K. Singh, and T. Mohapatra, “SNPs in stress-responsive rice genes: validation, genotyping, functional relevance and population structure,” *BMC Genomics*, vol. 13, no. 1, p. 426, 2012.

[67] S. Naithani, J. Preece, P. D’Eustachio et al., “Plant Reactome: a resource for plant pathways and comparative analysis,” *Nucleic Acids Research*, vol. 41, no. 2, pp. 877, 2008.

[68] P. J. Cao, L. E. Bartley, K. H. Jung, and P. C. Ronald, “Construction of a rice glycosyltransferase phylogenomic database and identification of rice-diverged glycosyltransferases,” *Molecular Plant*, vol. 1, no. 5, pp. 858–877, 2008.

[69] C. Lee, Q. Teng, R. Zhong, Y. Yuan, and Z. H. Ye, “Functional roles of rice glycosyltransferase family GT43 in xylan
biosynthesis,” Plant Signaling & Behavior, vol. 9, no. 3, article e27809, 2014.

[70] T. Nozoye, S. Nagasaka, T. Kobayashi et al., “Phytosiderophore efflux transporters are crucial for iron acquisition in graminaceous plants,” Journal of Biological Chemistry, vol. 286, no. 7, pp. 5446–5454, 2011.

[71] S. Wilkens, “Structure and mechanism of ABC transporters,” F1000Prime Reports, vol. 7, 2015.

[72] M. Gu, J. Zhang, H. Li et al., “Maintenance of phosphate homeostasis and root development are coordinately regulated by MYB1, an R2R3-type MYB transcription factor in rice,” Journal of Experimental Botany, vol. 68, no. 13, pp. 3603–3615, 2017.

[73] X. Li, Y. Jiang, Z. Ji, Y. Liu, and Q. Zhang, “BRHIS1 suppresses rice innate immunity through binding to monoubiquitinated H2A and H2B variants,” EMBO Reports, vol. 16, no. 9, pp. 1192–1202, 2015.

[74] T. Nakano, K. Suzuki, T. Fujimura, and H. Shinshi, “Genome-wide analysis of the ERF gene family in arabidopsis and rice,” Plant Physiology, vol. 140, no. 2, pp. 411–432, 2006.

[75] W. A. Snedden and H. Fromm, “Calmodulin, calmodulin-related proteins and plant responses to the environment,” Trends in Plant Science, vol. 3, no. 8, pp. 299–304, 1998.

[76] P. Jaiswal, “Gramene: a bird’s eye view of cereal genomes,” Nucleic Acids Research, vol. 34, no. 90001, pp. D717–D723, 2006.

[77] J. H. Ko, B. G. Kim, H.-G. Hur, Y. Lim, and J.-H. Ahn, “Molecular cloning, expression and characterization of a glycosyltransferase from rice,” Plant Cell Reports, vol. 25, no. 7, pp. 741–746, 2006.

[78] J. H. Kim, Y. M. Cheon, B. G. Kim, and J. H. Ahn, “Analysis of flavonoids and characterization of the OsFNS gene involved in flavone biosynthesis in Rice,” Journal of Plant Biology, vol. 51, no. 2, pp. 97–101, 2008.

[79] J. H. Ko, B. G. Kim, J. H. Kim et al., “Four glucosyltransferases from rice: cDNA cloning, expression, and characterization,” Journal of Plant Physiology, vol. 165, no. 4, pp. 435–444, 2008.

[80] C. H. Shih, H. Chu, L. K. Tang et al., “Functional characterization of key structural genes in rice flavonoid biosynthesis,” Planta, vol. 228, no. 6, pp. 1043–1054, 2008.

[81] M. M. Rahman, K. E. Lee, E. S. Lee et al., “The genetic constitutions of complementary genes Pp and Pb determine the purple color variation in pericarps with cyanidin-3-O-glucoside depositions in black rice,” Journal of Plant Biology, vol. 56, no. 1, pp. 24–31, 2013.

[82] L. Lepiniec, I. Debeaujon, J.-M. Routaboul et al., “Genetics and biochemistry of seed flavonoids,” Annual Review of Plant Biology, vol. 57, no. 1, pp. 405–430, 2006.

[83] F. Quattrocchio, A. Baudry, L. Lepiniec, and E. Grotewold, “The regulation of flavonoid biosynthesis,” in The Science of Flavonoids, E. Grotewold, Ed., pp. 97–122, Springer-Verlag, New York, NY, USA, 2006.

[84] S. Li, “Transcriptional control of flavonoid biosynthesis,” Plant Signaling & Behavior, vol. 9, no. 1, article e27522, 2014.

[85] H. Maeda, T. Yamaguchi, M. Omoteno et al., “Genetic dissection of black grain rice by the development of a near isogenic line,” Breeding Science, vol. 64, no. 2, pp. 134–141, 2014.

[86] M. D. Rausher, “The evolution of flavonoids and their genes,” in The Science of Flavonoids, pp. 175–211, Springer, New York, NY, USA, 2006.

[87] V. L. T. Hoang, D. J. Innes, P. N. Shaw, G. R. Monteith, M. J. Gidley, and R. G. Dietzgen, “Sequence diversity and differential expression of major phenylpropanoid-flavonoid biosynthetic genes among three mango varieties,” BMC Genomics, vol. 16, no. 1, p. 561, 2015.

[88] W. Chen, Y. Gao, W. Xie et al., “Genome-wide association analyses provide genetic and biochemical insights into natural variation in rice metabolism,” Nature Genetics, vol. 46, no. 7, pp. 714–721, 2014.

[89] L. Zhang, W. Su, R. Tao et al., “RNA sequencing provides insights into the evolution of lettuce and the regulation of flavonoid biosynthesis,” Nature Communications, vol. 8, no. 1, p. 2264, 2017.