

## Assessment of genetic diversity in Thai upland rice varieties using SSR markers

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### Abstract

Upland rice (*Oryza sativa* L.) is precious genetic resource containing some valuable alleles not common in modern germplasm. In this study, genetic diversity and population structure of 98 upland rice varieties from northern part of Thailand were examined using nine simple sequence repeat markers. Number of alleles detected by the above primers was 50 with a minimum and maximum frequency of 2 to 10 alleles per locus, respectively. The polymorphic information content (PIC) values ranged from 0.375 to 0.714 with an average of 0.605 for the primers RM164 and RM1, respectively. Dendrogram cluster analysis of the SSR data distinctly classified all genotypes into three major groups (I, II and III), which corresponded to their places of collection. Population structure divided these genotypes into two distinct subpopulations. Subpopulation 1 consisted of upland rice varieties that collected from Chiang Rai province while the majority of subpopulation 2 were collected from Phayao and Phitsanulok provinces. Analysis of molecular variance revealed 68% variance among two subpopulations and 32% variance within subpopulations, suggesting a high genetic differentiation between the two subpopulations. The huge genetic variability of upland rice in northern part of Thailand can be used to complement the gene pool of modern genotypes in rice breeding program.

**Keywords:** *Oryza sativa* L. SSR markers upland rice genetic diversity.

### Introduction

Rice (*Oryza sativa* L.), one of the most important crops, is globally cultivated and feeds over all of the population in the world (Mohanty 2013). Especially in Asia, rice is a staple crop and often considered as a cash crop because of its potential for export. Thailand in particular is well known as the world's largest rice exporter and is also among the world's largest rice producers (Milovanovic and Smutka 2017). Some critical issues occur however, regarding the rice production. For example, unforeseen climatic changes (i.e., drought and flooding) affect directly the rice yield. On the other hand, the trend of the consumers has changed, especially for those with high incomes which focus on a premium quality rice. As a result, several breeding programs have been undertaken in order to improve rice with desired characteristics, corresponding to this tentative situation.

It has been suggested that rice was domesticated between 8,000 – 10,000 years ago from its wild ancestor, *Oryza rufipogon* (Oka 1988). Since then, the domestication process involving strong screening for desirable traits, causes in precipitous loss of the genetic diversity (Londo et al. 2006). Rice in particular is a good example for this situation. Modern rice varieties have been bred for high yield as well as high quality. These rice varieties being bred for such purposes tend to loss genetic diversity and thus this may have a series of effects from susceptibility to epidemic diseases or even cause a serious threat (i.e. rice extinction). Therefore, knowledge of genetic diversity in the genepool of rice is crucial considering that such information can be used

efficiently in the rice breeding program generating new varieties suitable to changing cultivated conditions.

Upland rice cultivars have been traditionally cultivated by minority people mainly in mountain areas of Southeast Asia (Oka 1988; Sato 1987, 1991). Upland rice is grown during rainy season without irrigation, depending only on rain. Most grains are consumed by the farmer family and the rest is sown in the next year without severe selection. Upland rice is considered as an important gene source for the resistance of insects, pathogens and abiotic stress (Ishikawa et al. 2006).

Recently, DNA technology has been successfully applied in the plant breeding program. One of the major applications is to introduce the DNA markers specific for the desirable traits of the plant cultivars allowing direct detection of these 'desired' plants in the breeding program. In rice, there are a large number of microsatellite markers with different simple sequence repeat (SSR) motifs available on databases (Akagi et al. 1996; Chen et al. 1997; Panaud et al. 1996; Temnykh et al. 1999; Wu and Tanksley 1993). The microsatellite markers are distributed uniformly throughout the genome and can detect a high level of allelic diversity in cultivated varieties and distantly related species that made it possible to investigate the incidence and variability of simple sequence repeats at the whole-genome level (Cho et al. 2000; McCouch et al. 1997). Many studies have used SSR markers to investigate the genetic diversity and population structure within rice (Pusadee et al. 2009; Salgotra et al. 2015;

Vilayheuang et al. 2016; Wunna et al. 2016). For example, Wunna et al. (2016) examined genetic variation of rice (*Oryza sativa* L.) germplasm in Myanmar using SSR markers and found that rice germplasm in Myanmar has high genetic diversity among ecosystems and areas. Furthermore, microsatellite DNA markers were used to study genetic diversity and population structure of 'Khao Kai Noi', a landrace rice, in Laos. The result showed that genetic variation was largest among accessions and smallest within accessions. 'Khao Kai Noi' accessions were clustered into three different genetic backgrounds (Vilayheuang et al. 2016). However, there was no report of genetic diversity of Thai upland rice. Therefore the genetic diversity of 98 upland rice varieties collected from northern of Thailand were analyzed using nine SSR markers to understand the present genetic diversity in Thailand and to be utilized in rice breeding program in the future.

## Results and Discussion

### Genetic diversity values among 98 upland rice varieties

A total of 50 alleles from 9 SSR primer pairs were detected across all 98 upland rice varieties in northern part of Thailand. The number of alleles per primer pair (locus) detected by microsatellite primers varied from 2 to 10 with an average of 5.556 alleles per locus with 33.33% and 22.22% of the loci having five and four alleles, respectively (Table 1). The average numbers of alleles per locus observed in this study correspond well to Cho et al. (2000) who reported that the average alleles per locus for various classes of microsatellites in rice germplasm were 2.0 - 5.5 alleles per locus. However, the mean of alleles per locus is in agreement with Brondani et al. (2006) who detected an average of 5.4 alleles per locus when 25 SSR markers were used to distinguish 20 and 10 cultivars of upland rice and commercial rice from Brasil, respectively. In addition, these results were similar to the previous report of Vilayheuang et al. (2016) who calculated an average of 5.7 alleles per locus among 70 accessions of Khao Kai Noi (Lao rice) from Laos.

Nine SSR primer pairs used in this study could generate polymorphic bands and the polymorphic information content (PIC) values that reflected allele diversity and frequency among the upland rice varieties. The PIC values are a good indication of the usefulness of markers for linkage analysis when defining the inheritance between offspring and parental genotypes (Shete et al. 2000). In this study, the PIC values ranged from 0.375 in RM164 to 0.714 in RM1 with an average of 0.605 (Table 1). Botstein et al. (1980) reported that the PIC value > 0.5 meaning the locus was high diversity. If the PIC value was between 0.25 and 0.50 meaning, the locus was intermediate diversity when PIC value < 0.25, the locus was low diversity. Our study showed that the PIC values for almost all the SSR markers (excepting RM164) were higher than 0.5 indicating that all the SSR markers were considered high informative markers. Similar results were also found in 175 accessions of upland and lowland rice in Myanmar, Thailand and Yunan in China, which had 0.75 PIC values (Wunna et al. 2016). Shannon's Information index (*I*) averaged 1.266 and ranged from 0.693 to 1.723. Expected heterozygosity (*He*) in the population varied from 0.5 (RM164) to 0.74 (RM1) with an average of

0.66, while observed heterozygosity (*Ho*) ranged from 0.000 (RM253) to 1.000 (RM22) with a mean of 0.472.

### Genetic relationship among upland rice varieties

All 50 SSR alleles scored were used to calculate the genetic similarity which used to determine the level of relatedness among the upland rice varieties. The Dice's similarity coefficients among upland rice varieties ranged from 0.1554 to 0.8000, indicating a high genetic diversity among the 98 upland rice varieties. This is in agreement with Wunna et al. (2016) who studied the genetic variation of rice germplasm in Myanmar, including landraces and improved types from upland and lowland rice using SSR markers and the results show that rice germplasm in Myanmar has high genetic diversity. In addition, Thailand, Myanmar, Laos and other countries in Southeast Asia are located at the center of diversity for rice (Nakagahra and Hayashi 1977).

Genetic similarity values among the upland rice varieties were then used to group the varieties and to construct a dendrogram based on the UPGMA cluster analysis using the R program. In the dendrogram (Fig. 1), all genotypes of upland rice varieties were distinctly separated into three major groups, designated as I, II and III. Group I contained 34 varieties (34.69% in total) at a similarity coefficient of 20%. This group could be further sub-clustered into 4 subgroups with varying levels of similarity coefficients. The major subgroup A at a similarity coefficient of about 24% comprised of 15 varieties that received from Phayao province. The remaining varieties from Phayao province were clustered in the subgroup D at a similarity coefficient of about 34%. Similarly, collections from Chiang Rai province were grouped into the subgroup B and C at similarity coefficients of about 28% and 31%, respectively (Fig. 1). Group II, the smallest group, comprised of eight varieties mainly received from Phitsanulok province which clustered at similarity coefficient of 30%. Group III contained 56 varieties (57.14% in total), most of them collected from Chiang Rai province. The similarity coefficients of this group ranged from 44% to 100%. This cluster could be divided into two groups. One major group with similarity coefficient of about 38%, comprising of 10 varieties, was mostly collected from the same village. The second group at the similarity coefficient of about 39% contained 46 varieties collected from many villages in Chiang Rai province. Based on the dendrogram, our data showed that upland rice varieties were well clustered with respect to their places/ geographic area of collection and the genetic diversity among upland rice varieties from the three areas in Thailand has high genetic diversity.

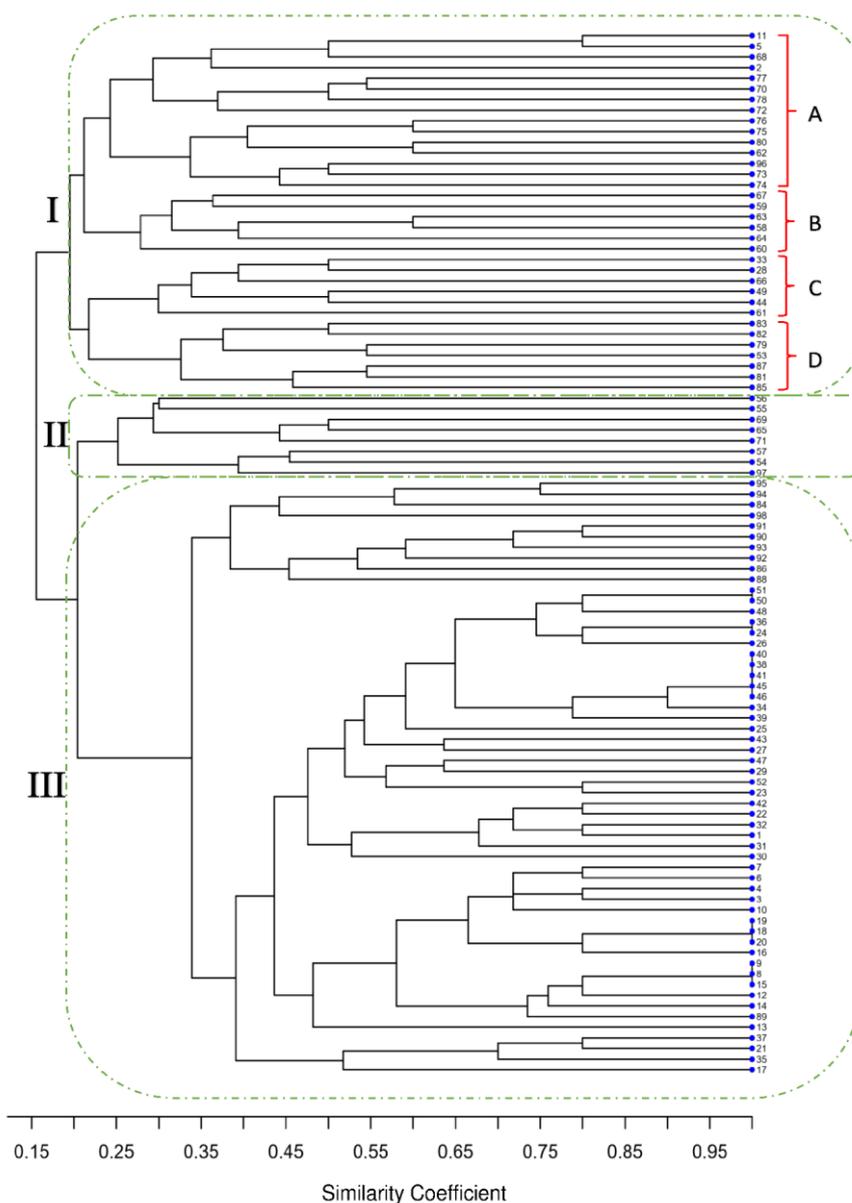
### Population structure analysis

The Bayesian model-based structure analysis was carried out by K values from 1 to 10 with 10 iterations using all 98 genotypes. In order to find the optimal K-value, the possible cluster numbers (K-value) were plotted against  $\Delta K$  which showed a clear maximum peak at K = 2 (Fig. 2A). A continuous gradual increase was observed in the log likelihood with the increased of K (Fig. 2B). The optimal K-value stratified that two subpopulations assigned to the subpopulation 1 and 2 showed the highest probability for population clustering. The subpopulation 1 (orange color,

**Table 1.** Nine SSR primer pairs information and the information of polymorphism obtained from 98 upland rice varieties. information content.

SSR primers	Chr.	SSR motif	Primers sequences (5'→3')	Number of alleles	He	Ho	I	PIC
RM1	1	(GA) <sub>26</sub>	F: GCG AAA ACA CAA TGC AAA AA R: GCG TTG GTT GGA CCT GAC	10	0.740	0.500	1.723	0.714
RM10	2	(GA) <sub>15</sub>	F:TTG TCA AGA GGA GGC ATC G R: CAG AAT GGG AAA TGG GTC C	5	0.717	0.133	1.342	0.658
RM19	4	(ATC) <sub>10</sub>	F: CAA AAA CAG AGC AGA TGA C R: CTC AAG ATG GAC GCC AAG A	4	0.611	0.480	1.125	0.556
RM22	3	(GA) <sub>22</sub>	F: GGT TTG GGA GCC CAT AAT CT R: CTG GGC TTC TTT CAC TCG TC	4	0.688	1.000	1.125	0.627
RM164	5	(GT) <sub>16</sub> TT(GT) <sub>4</sub>	F: TCT TGC CCG TCA CTG CAG ATA TCC R: GCA GCC CTA ATG CTA CAA TTC TTC	2	0.502	0.122	0.693	0.375
RM241	4	(CT) <sub>31</sub>	F: GAG CCA AAT AAG ATC GCT GA R: TGC AAG CAG CAG ATT TAG TG	9	0.742	0.980	1.552	0.698
RM252	4	(GA) <sub>19</sub>	F: TTC GCT GAC GTG ATA GGT TG R: ATG ACT TGA TCC CGA GAA CG	5	0.662	0.929	1.252	0.597
RM253	6	(GA) <sub>25</sub>	F: TCC TTC AAG AGT GCA AAA CC R: GCA TTG TCA TGT CGA AGC C	6	0.672	0.000	1.295	0.621
OSR28	9	(AGA) <sub>n</sub>	F: AGC AGC TAT AGC TTA GCT GG R: ACT GCA CAT GAG CAG AGA CA	5	0.641	0.102	1.287	0.599
Total				50				
Average				5.556	0.664	0.472	1.266	0.605

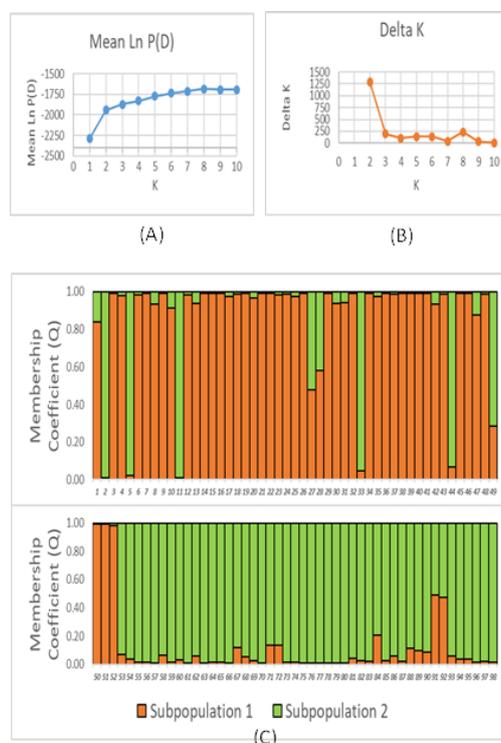
Chr.= chromosomes; He= expected heterozygosity; Ho= observed heterozygosity; I= Shanon information index; PIC= polymorphism.



**Fig 1.** UPGMA dendrogram showing three clusters (I, II and III) of all 98 upland rice varieties.

**Table 2.** Population structure results of 98 upland rice varieties for the fixation index (*Fst*), expected heterozygosity (*He*), number of genotypes in each subpopulation and inferred subpopulation.

Subpopulation	<i>Fst</i>	<i>He</i>	No. of genotypes	Inferred subpopulation
Subpopulation 1	0.4456	0.4222	45	0.481
Subpopulation 2	0.0280	0.6672	53	0.518

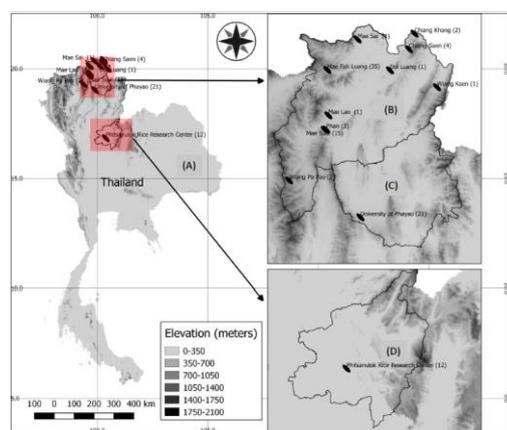


**Fig 2.** Population structure of 98 varieties of upland rice. (A) The relationship between  $\Delta K$  and *K* showing the maximum peak at *K* = 2 (B) The average log-likelihood of *K*-value against the number of *K* (C) The population structure of 98 upland rice varieties on *K* = 2. Varieties in orange color clustered into subpopulation 1 and the ones in green grouped into subpopulation 2.

**Table 3.** Analysis of molecular variance (AMOVA) among and within two subpopulations of 98 upland rice varieties.

Source of variance	<i>df</i>	SS	MS	Variance components	Variation (%)
Among subpopulation	1	215864.919	215864.919	4392.818	68
Within subpopulation	96	196919.816	2051.248	2051.248	32
Total	97	412784.735		6444.066	100

*df*: degrees of freedom; SS: Sums of squares, MS: Means squares.



**Fig 3.** Sampling locations of upland rice varieties from 3 provinces in Thailand. (A) Map of Thailand (B) Enlarged view of upland rice growing districts in Chiang Rai province (C) and (D) Maps of University of Phayao and Phitsanulok Rice Research Center, respectively. The numbers in parentheses stand for the total numbers of upland rice varieties collected from that place.

**Table 4.** Name, code, origin, latitude and longitude of provinces in Thailand and cluster based on the UPGMA clustering of the 98 upland rice varieties.

Code	Name	Place of collection	Latitude and longitude of provinces in Thailand	Height (m)	Cluster
1	Lap Chang	Mae Sai district, Chiang Rai	N 19 40 50.5 E 99 37 59.6	467	III
2	Khao Sim Khao	Phan district district, Chiang Rai	N 19 40 54.8 E 99 38 37.1	467	I (A)
3	Beu Mheu	Wiang Par Pao district, Chiang Rai	N 19 18 10.62 E 99 22 13.91	1097.83	III
4	Pi Ai Zoo	Wiang Par Pao district, Chiang Rai	N 19 18 10.62 E 99 22 13.91	1097.83	III
5	Hom Doi	Chiang Khong district, Chiang Rai	N 20 23 44.1 E 100 17 43.9	382	I (A)
6	Khao Kam09	Chiang Khong district, Chiang Rai	N 20 23 44.1 E 100 17 43.9	382	III
7	Khao Daeng	Chiang Saen district, Chiang Rai	N 20 16 48.7 E 100 15 15.3	440	III
8	Khao Kam012	Chiang Saen district, Chiang Rai	N 20 16 48.7 E 100 15 15.3	440	III
9	Khao Kam013	Chiang Saen district, Chiang Rai	N 20 16 48.7 E 100 15 15.3	440	III
10	Khao' Pleuak Kheaw	Chiang Saen district, Chiang Rai	N 20 16 48.7 E 100 15 15.3	440	III
11	Unknown015	Mae Fah Luang district, Chiang Rai	N 20 17 35.1 E 99 48 55.8	940	I (A)
12	Khao Jao Doi016	Mae Suai district, Chiang Rai	N 19 40 48.6 E 99 39 14.1	453	III
13	Khao' Jao Doi017	Mae Suai district, Chiang Rai	N 19 68 018 E 99 65 393	453	III
14	Unknown 018	Doi Luang district, Chiang Rai	N 20 7 22.0 E 100 6 51.0	379	III
15	A-Kha Ja Bue	Mae Fah Luang district, Chiang Rai	N 20 10 41.5 E 99 42 18.5	737	III
16	La Hae020	Mae Fah Luang district, Chiang Rai	N 20 10 41.5 E 99 42 18.5	737	III
17	Unknown021	Mae Fah Luang district, Chiang Rai	N 20 10 41.5 E 99 42 18.5	737	III
18	Chaw Miae Chae	Mae Fah Luang district, Chiang Rai	N 20 10 41.5 E 99 42 18.5	737	III
19	La Hae023	Mae Fah Luang district, Chiang Rai	N 20 10 41.5 E 99 42 18.5	737	III
20	Unknown024	Mae Fah Luang district, Chiang Rai	N 20 10 41.5 E 99 42 18.5	737	III
21	Khao Khao Chae Bah	Mae Fah Luang district, Chiang Rai	N 20 14 42.6 E 99 33 35.9	1028	III
22	Chae Mew	Mae Fah Luang district, Chiang Rai	N 20 14 42.6 E 99 33 35.9	1028	III
23	Chae Yah Yaw Ti	Mae Fah Luang district, Chiang Rai	N 20 14 42.6 E 99 33 35.9	1028	III
24	Chae Yah Yaw Heu	Mae Fah Luang district, Chiang Rai	N 20 14 42.6 E 99 33 35.9	1028	III
25	Jar Lo Mah	Mae Fah Luang district, Chiang Rai	N 20 14 42.6 E 99 33 35.9	1028	III
26	Kha Pah Chae Ne	Mae Fah Luang district, Chiang Rai	N 20 14 42.6 E 99 33 35.9	1028	III
27	Daw Choo	Mae Fah Luang district, Chiang Rai	N 20 14 42.6 E 99 33 35.9	1028	III
28	Chair Miaw Rae	Mae Fah Luang district, Chiang Rai	N 20 14 42.6 E 99 33 35.9	1028	I (C)
29	Unknown034	Wiang Kan district, Chiang Rai	N 19 59 47 E 100 27 44	446	III
30	Unknown035	Mae Sai district, Chiang Rai	N 20 35 198 E 99 876	400	III
31	Ja Naw Vuey	Mae Fah Luang district, Chiang Rai	N 20 7 30.9 E 99 39 7.6	894	III
32	Kaw Rue Sue	Mae Fah Luang district, Chiang Rai	N 20 7 30.9 E 99 39 7.6	894	III
33	Kaw Hom	Mae Fah Luang district, Chiang Rai	N 20 7 30.9 E 99 39 7.6	894	I (C)
34	Ja Beu Mah	Mae Fah Luang district, Chiang Rai	N 20 7 30.9 E 99 39 7.6	894	III
35	Khaw Mah Hah040	Mae Fah Luang district, Chiang Rai	N 20 7 30.9 E 99 39 7.6	894	III
36	Ja Hae	Mae Fah Luang district, Chiang Rai	N 20 7 30.9 E 99 39 7.6	894	III
37	Ja Seu Hae	Mae Fah Luang district, Chiang Rai	N 20 7 30.9 E 99 39 7.6	894	III
38	Ja Bi Ger or Ja Ber Ger	Mae Fah Luang district, Chiang Rai	N 20 7 30.9 E 99 39 7.6	894	III
39	Pae Hah Ja Naw	Mae Fah Luang district, Chiang Rai	N 20 7 30.9 E 99 39 7.6	894	III
40	Kaw Mah Hah045	Mae Fah Luang district, Chiang Rai	N 20 7 30.9 E 99 39 7.6	894	III
41	Ja Sue Mah	Mae Fah Luang district, Chiang Rai	N 20 7 30.9 E 99 39 7.6	894	III
42	Ja Na Gui	Mae Fah Luang district, Chiang Rai	N 20 7 30.9 E 99 39 7.6	894	III
43	Kaw Mah Hah Ja Chi	Mae Fah Luang district, Chiang Rai	N 20 7 30.9 E 99 39 7.6	894	III
44	O-Sa	Mae Fah Luang district, Chiang Rai	N 20 7 58.2 E 99 38 3.1	1107	I (C)
45	Khao Maw	Mae Fah Luang district, Chiang Rai	N 20 7 58.2 E 99 38 3.1	1107	III
46	Che Ba Ma	Mae Fah Luang district, Chiang Rai	N 20 7 58.2 E 99 38 3.1	1107	III
47	U-Mah Na	Mae Fah Luang district, Chiang Rai	N 20 7 58.2 E 99 38 3.1	1107	III
48	Che Bah Jui	Mae Fah Luang district, Chiang Rai	N 20 7 58.2 E 99 38 3.1	1107	III
49	Chae Sa	Mae Fah Luang district, Chiang Rai	N 20 7 58.2 E 99 38 3.1	1107	I (C)
50	Ka Moo	Mae Fah Luang district, Chiang Rai	N 20 7 58.2 E 99 38 3.1	1107	III
51	Unknown056	Mae Lao district, Chiang Rai	N 19 47 15.0 E 99 39 36.0	489	III
52	Khao Kum057	Phan district, Chiang Rai	N 19 40 54.8 E 99 38 37.1	467	III
53	Khao Sim Khao053	Phan district, Chiang Rai	N 19 40 54.8 E 99 38 37.1	467	I (D)
54	Chil Mae Jan	Wang Thong district, Phitsanulok	N 16 50 19.0 E 100 22 41.0	46	II
55	Jaow Num Roo	Wang Thong district, Phitsanulok	N 16 50 19.0 E 100 22 41.0	46	II
56	Law Take	Wang Thong district, Phitsanulok	N 16 50 19.0 E 100 22 41.0	46	II
57	Blae Klur	Wang Thong district, Phitsanulok	N 16 50 19.0 E 100 22 41.0	46	II
58	Ber Por Lo	Wang Thong district, Phitsanulok	N 16 50 19.0 E 100 22 41.0	46	I (B)
59	San Par Tong	Wang Thong district, Phitsanulok	N 16 50 19.0 E 100 22 41.0	46	I (B)
60	Bar Nhi	Wang Thong district, Phitsanulok	N 16 50 19.0 E 100 22 41.0	46	I (B)
61	Pa Ya Lurm Kang	Wang Thong district, Phitsanulok	N 16 50 19.0 E 100 22 41.0	46	I (C)
62	La Oup	Wang Thong district, Phitsanulok	N 16 50 19.0 E 100 22 41.0	46	I (A)
63	Khaow Tar Hong	Wang Thong district, Phitsanulok	N 16 50 19.0 E 100 22 41.0	46	I (B)
64	Hang Pla Lhai	Wang Thong district, Phitsanulok	N 16 50 19.0 E 100 22 41.0	46	I (B)
65	Mon Pu	Wang Thong district, Phitsanulok	N 16 50 19.0 E 100 22 41.0	46	II
66	Situ Patenggang	Muang Phayao district, Phayao	N 19 1 43.0 E 99 53 47.0	494	I (C)
67	Bue Nue Mu	Muang Phayao district, Phayao	N 19 1 43.0 E 99 53 47.0	494	I (B)

**Table 4** Continued.

Code	Name	Place of collection	Latitude and longitude of provinces in Thailand	Height (m)	Cluster
68	Khao Lueng Hom	Muang Phayao district, Phayao	N 19 1 43.0 E 99 53 47.0	494	I (A)
69	Mali Nam Nao	Muang Phayao district, Phayao	N 19 1 43.0 E 99 53 47.0	494	II
70	CPAC060014	Muang Phayao district, Phayao	N 19 1 43.0 E 99 53 47.0	494	I (A)
71	CPAC08043	Muang Phayao district, Phayao	N 19 1 43.0 E 99 53 47.0	494	II
72	Nam Ru	Muang Phayao district, Phayao	N 19 1 43.0 E 99 53 47.0	494	I (A)
73	IR78914-B-22-B-B-B	Muang Phayao district, Phayao	N 19 1 43.0 E 99 53 47.0	494	I (A)
74	IR81423-B-B-111-3	Muang Phayao district, Phayao	N 19 1 43.0 E 99 53 47.0	494	I (A)
75	IR7887-048-B-B-2	Muang Phayao district, Phayao	N 19 1 43.0 E 99 53 47.0	494	I (A)
76	IR71700-247-1-1-2	Muang Phayao district, Phayao	N 19 1 43.0 E 99 53 47.0	494	I (A)
77	PSL85051-14-2-1-2	Muang Phayao district, Phayao	N 19 1 43.0 E 99 53 47.0	494	I (A)
78	CNT86095-42-2-3	Muang Phayao district, Phayao	N 19 1 43.0 E 99 53 47.0	494	I (A)
79	Unknown UP-53	Muang Phayao district, Phayao	N 19 1 43.0 E 99 53 47.0	494	I (D)
80	IR13240-108-2-2-3	Muang Phayao district, Phayao	N 19 1 43.0 E 99 53 47.0	494	I (A)
81	IR15675-81-2-3	Muang Phayao district, Phayao	N 19 1 43.0 E 99 53 47.0	494	I (D)
82	IR15795-199-3-3	Muang Phayao district, Phayao	N 19 1 43.0 E 99 53 47.0	494	I (D)
83	Bue Wa	Muang Phayao district, Phayao	N 19 1 43.0 E 99 53 47.0	494	I (D)
84	Nam Ru	Muang Phayao district, Phayao	N 19 1 43.0 E 99 53 47.0	494	III
85	2R-43	Muang Phayao district, Phayao	N 19 1 43.0 E 99 53 47.0	494	I (D)
86	Ja Chi	Muang Phayao district, Phayao	N 19 1 43.0 E 99 53 47.0	494	III
87	Khao Sill	Mae Suai district, Chiang Rai	N 19 49 23.6 E 99 33 27.4	1142	I (D)
88	Jaa Ngee Si	Mae Suai district, Chiang Rai	N 19 49 23.6 E 99 33 27.4	1142	III
89	Jaa Bae Bae	Mae Suai district, Chiang Rai	N 19 49 23.6 E 99 33 27.4	1142	III
90	Jaa Da Mor	Mae Suai district, Chiang Rai	N 19 49 23.6 E 99 33 27.4	1142	III
91	Chep Pea	Mae Suai district, Chiang Rai	N 19 49 23.6 E 99 33 27.4	1142	III
92	Char-Ku-Lae	Mae Suai district, Chiang Rai	N 19 49 23.6 E 99 33 27.4	1142	III
93	Ta-Tae-Maa-Cha	Mae Suai district, Chiang Rai	N 19 49 23.6 E 99 33 27.4	1142	III
94	Jaa-Da-Ma	Mae Suai district, Chiang Rai	N 19 49 23.6 E 99 33 27.4	1142	III
95	A-The-Ma	Mae Suai district, Chiang Rai	N 19 49 23.6 E 99 33 27.4	1142	III
96	Khao-Neaw-LeeSaw	Mae Suai district, Chiang Rai	N 19 49 23.6 E 99 33 27.4	1142	I (A)
97	Lee-Su-Jaa	Mae Suai district, Chiang Rai	N 19 49 23.6 E 99 33 27.4	1142	II
98	Ta-The-Ma-Ja	Mae Suai district, Chiang Rai	N 19 49 23.6 E 99 33 27.4	1142	III

Fig. 2C) consisted of 45 varieties (45.92%) collected from Chiang Rai province and the subpopulation 2 (green color, Fig. 2C) included 53 (54.08%) of varieties collected from Phayao and Phitsanulok provinces. The structure analysis suggested differentiation between two subpopulations and clustered them with the geographic area. The fixation index ( $F_{st}$ ) for each of the subpopulation was estimated the genetic variation. Genetic differentiation of subpopulation 1 was very strong differentiation ( $F_{st} = 0.4456$ ). However, a low  $F_{st}$  value (0.0280) was found in the subpopulation 1 meaning little differentiation.

The subpopulations 1 and 2 had  $F_{st}$  values of 0.4456 and 0.0280, respectively, with an average value of 0.2368 (Table 2). suggested that there was significant divergence within the subpopulation 2.

#### **Analysis of molecular variance**

The two subpopulations generated from population structural analysis were also determined using analysis of molecular variance (AMOVA) to estimate the percentage of variation among subpopulation and within subpopulation of 98 upland rice varieties. The majority of the genetic variation in upland rice varieties based on structure was due to among subpopulation variation (68%) and the remaining 32% was attributed to individual differences within subpopulation (Table 3), indicating high genetic differentiation between the two subpopulation.

## **Materials and Methods**

### **Plant materials**

A total of 98 upland rice varieties were used in this study (Table 4). Sixty five varieties of upland rice were collected from farmers in 10 districts in Chiang Rai province, Thailand and 21 and 12 varieties were obtained from Dr. Vaiphot Kunjoo, University of Phayao and Phitsanulok Rice Research Center, Thailand, respectively (Fig. 3). Seeds of 98 varieties were planted on cultural tray filled with soil and grown at 25°C for two weeks.

### **Genomic DNA extraction**

Genomic DNA was extracted from bulk 14-day-old seeding leaves of each upland rice variety using the Cetyl Trimethyl Ammonium Bromide (CTAB) method previously described by Dolye and Dolye (1987). DNA was quantified by Nano-Drop 1000 spectrophotometer (Thermo Scientific, USA). Final concentration was adjusted to 50 ng/μl for SSR analysis.

### **PCR assay**

Nine SSR primer pairs (RM1, RM10, RM19, RM22, RM164, RM241, RM252, RM253 and ORS28) with relatively high polymorphism and distributed across the rice genome were selected for genetic diversity analysis on the basis of

published rice microsatellites. The chromosome positions, repeat motifs and primer sequences for these markers can be found in the rice genome database (<http://www.Gramene.org>) (Table 1). The polymerase chain reaction (PCR) was conducted in a total volume of 20 µl containing 50 ng of DNA template, 2 µl of 10x PCR buffer, 0.2 mM dNTPs, 2.5 mM MgCl<sub>2</sub>, 0.2 µM each primer, 0.5 units of *Taq* DNA polymerase (Vivantis, Malaysia). PCR reactions were carried out in Eppendorf Mastercycler Nexus Gradient GSX1 Thermal Cycler (USA). Thermal cycling program involved an initial denaturation at 94°C for 5 min, followed by 35 cycles of denaturation at 94°C for 1 min, annealing at 55 °C for 1 min (60°C for RM164) primer extension at 72°C for 30 sec, followed by a final extension at 72°C for 5 min. The PCR products were separated by electrophoresis in 6% (w/v) denaturing polyacrylamide gels in 1X Tris-borate-EDTA (TBE) buffer at 150 volts for 1 to 2 hours depending on the size of the PCR products. Gels were stained with RedSafe (iNtRON Biotechnology, USA) and visualized under UV light of the Gel document system. Allele sizes were estimated in comparison with 25 bp DNA ladder (Invitrogen, USA).

### Data analysis

The most intensively amplified bands for each SSR marker were scored. All upland rice varieties were scored for the presence (score '1') or absence (score '0') of the SSR band. Polymorphic information content (PIC), a measure of the allelic diversity at a locus, was calculated according to Anderson et al. (1993) using the following equation:  $PIC_i = 1 - \sum_j f_{ij}^2$  where  $f_{ij}$  is the frequency of the  $j^{th}$  pattern (present and absent) of the  $i^{th}$  band. Next, the PIC of each primer was calculated as:  $PIC = (\sum_{i=1}^n PIC_i) / n$  where  $n$  is the number of bands. Shannon information index ( $I$ ), expected heterozygosity ( $He$ ) and observed heterozygosity ( $Ho$ ) of each loci were calculated in GenAEx 6.502 software (Peakall and Smouse, 2006, 2012) and the Excel Microsatellite Toolkit (Park 2008), respectively. Genetic similarity among varieties was measured from the matrix of binary data using Jaccard coefficient. A dendrogram was constructed based on the resulting similarity coefficients using the unweighted pair-group method with the arithmetic averages (UPGMA) in the R program (Team, 2015). Analysis of molecular variance (AMOVA) was used to estimate variance among and within populations using GenAEx 6.502 software (Peakall and Smouse, 2006, 2012). Significance of variance was tested after 999 permutations. From AMOVA, the fixation index ( $F_{st}$ ) within the population obtained from GenAEx 6.502 software (Peakall and Smouse, 2006, 2012).  $F_{st}$  measures the amount of genetic variance. The  $F_{st}$  value of 0 indicates no differentiation between the subpopulation while the  $F_{st}$  value of 1 indicates complete differentiation (Bird et al. 2007). Populations were considered to have very strong differentiation when  $F_{st}$  values were greater than 0.25, strong differentiation when  $F_{st}$  values were between 0.15 and 0.25, moderate differentiation when  $F_{st}$  values were between 0.05 and 0.15 and little differentiation when  $F_{st}$  values were less than 0.05 (Hartl 1980; Mohammadi and Prasanna 2003). The Bayesian model-based clustering analysis was performed to infer genetic structure and to determine the optimal number of genetic clusters found among upland rice

varieties using the software STRUCTURE version 2.3.4 (Pritchard et al. 2000). The number of cluster ( $K$ ) was set from 1 to 10 and the analysis was repeated 10 times. The burn-in period was 100,000 interactions for each group number  $K$  and 100,000 Monte Carlo Markov Chain replications. The optimum value of  $K$  value which indicates the number of genetically distinct clusters in the data was obtained by calculating the  $\Delta k$  value. The  $\Delta k$  value was calculated based on the change in the log probability of the data between successive  $K$  values (Evanno et al. 2005).

### Conclusion

From the similarity coefficient distribution, dendrogram and population structure analysis showed that upland rice varieties in Thailand showed great genetic diversity. This knowledge of genetic diversity and population structure is important in terms of agriculture as they can be potential especially for using these upland rice varieties as a germplasm for the breeding program.

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