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# Development and Validation of A High Density 2 SNP Genotyping Array for African Oil Palm

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15 Running Title: High-density Oil Palm SNP Genotyping Array

#### 16 Abstract

High-density single nucleotide polymorphism (SNP) genotyping arrays are powerful 17 tools that can measure the level of genetic polymorphism within a population. To develop this 18 whole-genome SNP array, SNP discovery was performed using deep resequencing of eight 19 20 libraries derived from 132 Elaeis guineensis and Elaeis oleifera palms belonging to 59 origins, resulting in the discovery of >3 million putative SNPs. After SNP filtering, the Illumina OP200K 21 custom array was built with 170,860 successful probes. Phenetic clustering analysis revealed that 22 the array could distinguish between palms of different origins in a way consistent with pedigree 23 records. Genome-wide linkage disequilibrium (LD) declined more slowly for the commercial 24 populations (ranging from 120Kb at  $r^2=0.43$  to 146Kb at  $r^2=0.50$ ) when compared with the semi-25 wild populations (19.5Kb at  $r^2=0.22$ ). Genetic fixation mapping comparing the semi-wild and 26 27 commercial population identified 321 selective sweeps. A genome-wide association study (GWAS) detected a significant peak on Chromosome 2 associated with the polygenic component 28 of the shell thickness trait (based on the trait *shell-to-fruit*; S/F %) in *tenera* palms. Testing of a 29 genomic selection model on the same trait resulted in good prediction accuracy (r=0.65) with 30 42% of the S/F % variation explained. The first high-density SNP genotyping array for oil palm 31 has been developed and shown to be robust for use in genetic studies and with potential for 32 developing early trait prediction to shorten the oil palm breeding cycle. 33

Keywords: genotypes, whole-genome SNP array, genome-wide association study, genomic
selection, genetic fixation, shell thickness, oil palm, linkage disequilibrium, phenetic clustering,
whole-genome variation, genomic prediction, genome-wide linkage disequilibrium, polygenic

37 trait

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#### 41 Introduction

Oil palm (Elaeis guineensis Jacq.) is a monocot of the Arecaceae family that originated 42 from tropical Guinea's rainforest in West Africa (Hartley, 1967c). This plant is the most efficient 43 oil crop in the world, yielding commercially an average of 4 tons oil hectare<sup>-1</sup> year<sup>-1</sup>, equivalent 44 to ten times the average oil yield hectare<sup>-1</sup> year<sup>-1</sup> of other oil crops; palm oil currently supplies 45 30% of the global edible oil demand (US Department of Agriculture, 2015). In order to address 46 47 issues of climate change and land and labor shortage, the most sustainable strategy to meet future demand is to increase yields on the same planting area through oil palm breeding. For this 48 purpose, oil palm breeding will need to improve multiple traits affected by biotic, abiotic, 49 management and processing regimes simultaneously (Davidson, 1991; Meunier et al., 1979; Soh, 50 1981). 51

Genetic diversity is always the key resource for improvement of any agronomic trait. 52 Comprehensive germplasm prospections of E. guineensis have been carried out in Africa since 53 the first commercial cultivation of the crop in the early 20<sup>th</sup> century (Rajanaidu, 1986b). 54 Prospection has been extended to the African oil palm's close species relative, E. oleifera H.B.K. 55 Cortes, present in Central and South America (Rajanaidu, 1986a). The South American oil palm 56 is superior in terms of commercially desirable oil quality (high level of unsaturated oils), has low 57 height increment and carries wilt resistance (Hardon et al., 1985; Hartley, 1988; Mohd et al., 58 2002). However, the South American palm yields poorly, producing on average 0.5 tons oil 59 hectare<sup>-1</sup> year<sup>-1</sup> (Wahid et al., 2004). Attempts to introduce these beneficial traits from 60 prospective germplasm into the commercial E. guineensis through interspecific crossing 61 programs have met with varying degrees of success. Molecular markers, such as restriction 62 fragment length polymorphism (RFLP), amplified fragment length polymorphism (AFLP) and 63 random amplified polymorphic DNA (RAPD) were first used to assess genetic diversity of the 64 collected and available germplasm (Barcelos et al., 2002; Shah et al., 1994). Attention then 65 shifted to multi-allelic marker systems, particularly simple sequence repeats (SSRs), which 66 detect a high degree of allelic variation even among closely related individuals. Hence, SSR 67 markers for *E. guineensis* have been derived from genomic libraries and expressed sequence tag 68 (EST) databases (Billotte et al., 2001; Ting et al., 2010). The aim of the molecular marker 69 70 development was to identify new allelic variation in unexploited origins that could be used to

widen the genetic base of breeding programs (reducing potential inbreeding depression and 71 introducing novel trait alleles) through introgression. Genetic improvement in oil palm through 72 conventional breeding, however, is slow due to long selection cycles (typically 12 years) and the 73 lack of genetic homozygosity in current advanced parental breeding materials (Mayes et al., 74 2000). To shorten the selection cycle, molecular markers can be deployed to identify quantitative 75 trait loci (QTL) underlying traits of interests and to perform genome-wide screening of oil palm 76 (Wong and Bernardo, 2008). The individual palms that possess desired allelic combinations of 77 QTL can be identified in immature plants, even at the nursery stage, rather than relying on 7-10 78 79 years of phenotypic evaluation in the field, also facilitating rapid backcross introgression of novel traits into elite material. 80

Yield traits (e.g. oil yield and bunch production) in oil palm are mostly quantitatively 81 inherited and controlled by multiple genes. Controlled cross-based linkage analysis and GWAS 82 are the main methods that have been exploited to discover and locate genes/QTL for these 83 complex traits. Multi-allelic markers such as RFLP and SSR used so far normally only provide 84 low to medium mapping and genome coverage resolutions. In contrast, SNPs have a high level 85 of abundance and distribution uniformity throughout a genome and are now widely deployed for 86 developing high-density genome-wide scans in humans, and have been used more recently in 87 plants (Huang et al., 2010; Wang et al., 1998). The publication of the 1.535 Gb oil palm genome 88 and the independent assembly of Sime Darby's oil palm genome, together with the cost 89 reduction in high-throughput sequencing provide an opportunity for large scale and high density 90 SNP detection in oil palm (Singh et al., 2013b). 91

Whole-genome SNP arrays have been developed for many species including rice, corn, cattle, horse and more recently salmon (Chen et al., 2014; Ganal et al., 2011; Houston et al., 2014; McCue et al., 2012; Yu et al., 2014). To date, the only SNP genotyping array available for oil palm is a 4.5K custom Illumina SNP array, which has been proven to be useful in construction and comparison of linkage maps (Ting et al., 2014). However, for genome-wide scans, including GWAS and LD analysis, a higher density genotyping array is required.

98 This manuscript summarizes the overall properties of the first whole-genome SNP array 99 for oil palm. In addition, the analyses of 312 *tenera* palms from different backgrounds using the 100 genotyping array demonstrate a wide range of applications, including phenetic analysis, LD decay analysis, selective sweep mapping and GWAS. Oil palm breeders and researchers can use
 this genetic information as the basis for crop improvement programs, through applied marker
 assisted selection and/or direct genomic selection approaches.

- 104
- 105 **Results**

#### 106 Genome resequencing, SNP discovery and array design

107 The pooled resequencing generated approximately 73.76 Gb of data – approximately a 108 40x coverage of the oil palm genome. In total, 1,015,758,056 reads were generated. After 109 filtering, 870,235,918 high quality reads were identified for mapping against the published oil 110 palm genome and Sime Darby's in-house scaffolds independently. The sequencing data 111 produced had a mean sequence quality Phred score of 39 (Figure S1). There were no over-112 represented sequences and the average GC content was within the expected range of between 40-113 45%.

More than seven million SNPs were called from the resequencing data. The SNPs which 114 were unique for the *E. oleifera* genome (1,085,204 SNPs), those with mapping coverage less 115 than x17 or more than x53 (669,950 SNPs), genotyping quality (GQ) scores of less than 8 (5,089 116 117 SNPs) and minor allele frequency (MAF) of < 0.05 (71,503 SNPs) were removed as part of the quality control procedure. Based on the technical requirements of Illumina, 5,174,660 SNPs that 118 have secondary SNPs located within 60bp upstream or downstream were removed by filtering. In 119 addition, other criteria for the removal of candidates included SNPs that were indels (56,357 120 SNPs), non-biallelic (408 SNPs) and SNP types that required two bead types on the array to be 121 detected, i.e. A/T and C/G types (88,348 SNPs). Ambiguous bases within the 121bp probe region 122 were also identified and removed (10,992) (Table S1). From the 593,888 high quality SNPs 123 identified in silico, 200,000 SNPs were finally selected according to LD with a  $r^2$  cut-off set at 124 0.3 and uniform spacing across the assembled genome to populate the OP200K array. 125

126

#### 127 SNP genotyping

Out of the 200,000 designed SNPs, 170,860 markers passed bead representation and decoding quality metrics. By testing the array on 312 palms representing some of the 59 diverse oil palm origins, 161,073 SNPs generated reproducible scores and resulted in the basal data set used for the subsequent analysis. An example of the genotyping result is provided in Figure S2. Overall, only 0.65% of the entire genotyping data set was missing data.

A total of 70,804 genotyped SNPs could be located on the published oil palm physical 133 map. The density of these SNPs on the chromosomes is illustrated in Figure 1. Using the 134 published physical map as a reference, each chromosome has an average of 4,425 SNP loci. 135 Chromosome 10 has the highest marker density (1 SNP every 8,344bp), whereas Chromosome 6 136 has the lowest (1 SNP every 10,254bp). Overall, the density of the array is 1 SNP per 11Kb of 137 the oil palm physical map, inclusive of chromosomes and scaffolds. This resolution is 138 comparable with other published large-scale whole genome SNP genotyping arrays for plant 139 species (Chen et al., 2014; Ganal et al., 2011). 140

141

Figure 1. SNP distribution and density per chromosome in the oil palm genome assembly. The SNP distribution per chromosome is represented in the histogram. The mean genomic distance between markers on every chromosome is indicated as a line plot. The average SNP density of the OP200K array is 1 SNP per 11Kb, based on the oil palm physical map.

146

#### 147 **Phenetic analyses**

148 To reveal the genetic relationship between all 312 tenera palms, a full neighbor-joining dendrogram was built for the 312 individuals genotyped using the full 161,073 SNP dataset 149 (Figure S3). For illustration purposes, a representative dendrogram was also built using only 45 150 individuals (Figure 2). The bootstrap values from the representative dendrogram ranged from 80 151 -100, indicating good reproducibility. Both results identified five main clusters from the major 152 nodes, I. Ulu Remis (UR) x AVROS; II. Johore Labis (JL) x AVROS; III. Nigerian x AVROS; 153 IV. Gunung Melayu (GM) x Dumpy AVROS (DA); and V. JL x DA. This finding was consistent 154 155 with the different dura (maternal parent) and pisifera (paternal parent) used in breeding and with pedigree records. The commercial planting materials (I and II) were grouped separately from thesemi-wild (III) and breeding populations (IV and V).

158

Figure 2. Genetic relationships between representative *tenera* palms derived from different
backgrounds. A neighbor-joining dendrogram was constructed for 45 representative *tenera*palms based on pairwise genetic distances. Five main clusters were identified, including I. UR x
AVROS (red), II. JL x AVROS (purple), III. Nigerian x AVROS (green), IV. GM x DA (yellow) and V.
JL x DA (blue).

164

#### 165 Genome-wide LD decay

The average decay of LD with physical distance between SNPs was 146Kb ( $r^2 = 0.50$ ) and 120Kb ( $r^2 = 0.43$ ) for the UR x AVROS cluster and the JL x AVROS cluster. Average LD decayed faster at 87.7Kb ( $r^2 = 0.22$ ) and 54.7Kb ( $r^2 = 0.21$ ) for the JL x DA cluster and the GM x DA cluster, followed by 19.5Kb ( $r^2 = 0.22$ ) for the Nigerian x AVROS cluster (Figure 3). By using the minimal LD decay of 19.5Kb as a base line reference and compared with Chromosome 6 with the lowest marker density (1 SNP every 10,254bp), this array provides sufficient genomic resolution to detect association signals in any oil palm population.

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Figure 3. Decay of LD in the five *tenera* clusters from different backgrounds. Each cluster has a different LD decay pattern, expressed in  $r^2$ , which is probably due to different selection and breeding histories.

177

#### 178 Mapping of selective sweeps

179 A total of 321 selective sweeps were detected in the genome of the commercial 180 populations after comparison with the semi-wild population, with significant regions identified on every chromosome in the oil palm genome. 38% of the selection events were observed onChromosome 1, 2 and 3 (Figure 4).

183

Figure 4. Genome-wide mapping of selective sweeps. A total of 321 selective sweeps were identified in commercial populations by comparing them against the semi-wild Nigerian x AVROS population. Selective sweeps were defined as the top 1% of the score windows. The selective sweeps were highlighted in black.

In this study, 679 genes were present in the detected selective sweeps (Table S2). These genes could be annotated into 1097 Gene Ontology (GO) classes and 120 Enzyme Classes (EC). Statistically, 162 GOs and ECs were observed to be significantly over-represented (at Chi-square p<0.05) in the selective sweep regions for the commercial populations, as compared to the semiwild population (Table S3).

193

#### 194 GWAS & Genomic selection

To evaluate the usefulness of this array in GWAS, we selected the polygenic S/F % of 195 tenera palms as the trait of interest. While it has been shown that fruit type identity is controlled 196 by a single gene (Singh et al., 2013a), once this is taken into account there is still significant 197 198 variation remaining in breeding and commercial populations for S/F %. Here, all palms evaluated were *tenera*, which effectively removes the contribution of the SHELL identity gene, 199 per se. In this analysis, population structure was controlled using a kinship matrix while 200 201 environmental factors were initially tested as if unimportant. Genomic inflation was calculated to be 1.0, indicating that inflation was successfully controlled by this approach. From the GWAS 202 203 result, we observed a number of significantly associated SNPs across a genomic region 1,378,847bp in length, located on Chromosome 2 from position 2,526,459bp to 3,905,306bp 204 based on the genome-wide significance cut-off,  $-\log_{10} p$ -value = 5.0 (Figure 5). Some of the 205 206 genes that fall in this region include La-related protein 6B (LARP6B), SHELL/Seedstick (STK), Cytochrome c oxidase (COX), Transcription factor CPC (CPC) and Inositol-tetrakisphosphate 207 208 1-kinase (ITPK1). Two singleton markers were determined to be significantly associated at 209 position 5,708,516bp and 13,250,110bp of Chromosome 2. In addition, a few other signals were

located at Chromosome 5 and 10 (Table 1). A full GWAS SNP list is presented in Table S4.

211

Figure 5. Genome-wide association study (GWAS) for polygenic shell thickness (S/F %) in *tenera* palms. A Manhattan plot of the compressed Mixed Linear Model (MLM) for 312 *tenera* palms. Negative  $\log_{10}$  transformed *p*-values from a genome-wide scan are plotted against position on each of the 16 chromosomes. The horizontal line indicates the genome-wide significance cut-off,  $-\log_{10}(p-value) = 5.0$ .

217

Table 1. SNPs in	the main S/F	%-associated peaks
	Table 1. SNPs in	Table 1. SNPs in the main S/F

SNP ID	Chromosome	Position (bp)	-log <sub>10</sub> (p- value)	SNP effect	R-sq* (%)	Gene	
SD_SNP_000044630	2	2,526,459	5.73	0.82	3.27		
SD_SNP_000044629	2	2,528,936	5.25	3.54	3.05	LARP6B,	
SD_SNP_000054651	2	2,531,607	5.30	-3.71	2.92	STK,	
SD_SNP_000042701	2	3,516,726	5.88	-2.01	15.10	COX, CPC.	
SD_SNP_000022202	2	3,905,306	6.19	0.05	1.04	ITPK1	
SD_SNP_000008871	2	5,708,516	7.07	-1.46	2.30	GONST1	
SD_SNP_000017557	2	13,250,11 0	5.83	1.31	2.09	GBF4	
SD_SNP_000022069	5	16,977,01 6	5.04	1.58	14.16		
SD_SNP_000034242	5	17,308,44 2	5.20	0.14	15.64	FAD3C	
SD_SNP_000026705	10	25,688,31 4	5.07	1.47	18.04	NUD17	

219 \*R-sq represents the amount of variance explained

A 5-fold cross-validation, as described in the Methods section, was carried out. The correlation between Genomic Estimated Breeding Value (GEBV) and realized S/F % trait acquired was 0.65 (Figure 6) based on the criteria given in Methods section. The prediction model built suggested that 42% of the trait variation could be explained in the validation population.

225

Figure 6. Representative regression plot of GEBV and true shell thickness for the validation population set. The GEBV was obtained by Bayes A in the final model. The correlation coefficient between GEBV and true trait was 0.65 with 42% S/F variation explained.

229

#### 230 Discussion

The OP200K reported here is the first large scale oil palm SNP array. Since oil palm is a 231 232 naturally outcrossing and monoecious species, each oil palm genome can be highly heterozygous (Hartley, 1967a). The SNP array is therefore designed to be applicable to a wide range of oil 233 palm origins. As such, SNPs were discovered by resequencing palms of diverse origins including 234 all fruit forms (dura, tenera and pisifera) and the major breeding origins available at Sime Darby 235 Plantations. The different fruit forms are distinguished by the thickness of the shell, together with 236 237 the presence/absence of a fiber ring: *dura* (thick-shelled; without fiber ring), *tenera* (thin-shelled; with fiber ring), and *pisifera* (shell-less; female-sterile; with fiber ring). Shell thickness is 238 inversely correlated with the mesocarp percentage in each fruitlet, and therefore also to 239 240 commercial crude palm oil yield; because tenera, a hybrid between dura and pisifera, has the highest mesocarp oil yield, it is the fruit form exploited commercially. 241

The analyzed OP200K array data clearly revealed the genetic stratification of the 312 242 tenera palms from a number of important origins in a way concordant with pedigree records. The 243 five main clusters identified are I (UR x AVROS); II (JL x AVROS); III (Nigerian x AVROS); 244 IV (GM x DA); and V (JL x DA). Clusters I and II include the commercial populations, whereas 245 cluster III represents the semi-wild population. Cluster IV and V represent other breeding 246 populations at Sime Darby Plantations. In Southeast Asia, oil palm breeding improvement is 247 almost exclusively focused on the Deli *dura* origin on the maternal side, which is believed to be 248 249 derived from the four palms planted at the Bogor Botanic Garden, Indonesia, in 1848. The subsequent distribution to several breeding programs stressing different selection criteria over a 250 number of generations has led to differentiation between subpopulations, leading to pools of 251 252 'dissimilar' germplasm which have been termed 'breeding populations of restricted origins (BPROs)' (Rosenquist, 1986). The Deli dura BPROs, including UR, JL and GM, were selected 253

for different yield-related traits. The UR origin was chosen for high bunch number and high sex ratio (defined as the ratio of female inflorescences to total inflorescences) in Marihat Baris, Sumatra (Hartley, 1967b). Instead of bunch number, Socfindo in Sumatra selected the JL origin for bigger bunches (high bunch weight) and thinner shell. Another further interested origin, GM was planted by a Swiss company in the year 1919 and selections were made for low height increment and high oil/mesocarp (Corley and Tinker, 2003).

260 An immediate 30% increment of oil yield per hectare was gained after the introduction of the Deli dura x BM119 AVROS pisifera program to the Southeast Asia region (Corley and 261 Tinker, 2003; Hardon et al., 1987). The AVROS *pisifera* paternal parent line confers superiority 262 263 in uniformity of growth, general combining ability, precocity and high mesocarp oil content in the commercial Deli x AVROS (tenera) hybrid progeny. However, the narrow genetic base of 264 both Deli and AVROS - theoretically descended from four and two palms, respectively - may 265 hinder future breeding progress (Rosenquist, 1990). Consistent with this, it was observed that the 266 LD decayed slowest at 120Kb and 146Kb in both commercial JL x AVROS and UR x AVROS 267 clusters, respectively, compared with 19.5Kb in the semi-wild Nigerian x AVROS cluster, 268 indicating the lower genetic diversity/narrower genetic base and potentially the effects of higher 269 selection pressure in these materials. 270

Alternative genetic resources, such as semi-wild Nigerian dura and the DA pisifera 271 272 breeding materials are being exploited by the breeders to broaden the genetic variability. Direct improvement in oil yield per palm is the highest priority for oil palm breeding. However, due to 273 an acute labor shortage for fresh fruit bunch (FFB) collection, palm structural traits such as 274 275 height are also of potential interest. One approach to mitigate this problem has led to a shift in the oil palm breeder's attention to DA, GM and JL populations for development of materials 276 with lower annual height increment, to facilitate harvesting. The DA originated from the well-277 known Dumpy E206 self-pollinated cross, the parental palm that had an unusually large girth and 278 low height increment (Jagoe, 1952). The DA pisifera lines have been crossed with the short-279 280 stemmed GM dura and JL dura to reduce trunk height increment in their resulting tenera progeny (Corley and Tinker, 2003). The introduction of new genetic material into elite breeding 281 programs has indeed introduced more recombination into these tenera clusters, breaking the 282 283 long-range LDs. Hence, the LD in both the breeding populations (at 87.7Kb for JL x DA cluster; at 54.7Kb for UR x DA cluster) and the semi-wild Nigerian x AVROS cluster (at 19.5 Kb) decayed at least two-fold faster when compared to the commercial populations. Using the most rapid rate of LD as the reference (i.e. 19.5Kb), the OP200K array provides sufficient mapping resolution for whole genome linkage analysis and GWAS in any of the materials examined to date.

The long-range LD observed in the commercial populations can potentially be explained 289 further by breeding selection. The selected alleles will rise in frequency towards fixation in a 290 population and nearby linked alleles on the chromosome will be co-selected through linkage drag 291 with the selected locus to high frequency. This process is termed a 'selective sweep'. A total of 292 321 potential selective sweeps were observed in the oil palm genome based on the available 293 physical map. In these selective sweeps, the over-represented gene classes (based on Chi-square 294 295 tests) can be classified into different groups, including floral and fruit development (GO:0009733 response to auxin, GO:0048574 long-day photoperiodism, flowering, 296 GO:0048575, short-day photoperiodism, flowering, GO:0010483 pollen tube reception and 297 GO:0010183 pollen tube guidance) and lipid biosynthesis (GO:0019216 regulation of lipid 298 metabolic process, GO:0000038 very long-chain fatty acid metabolic process, GO:0042304 299 regulation of fatty acid biosynthetic process, GO:0006636 unsaturated fatty acid biosynthetic 300 process, GO:0045300 acyl-[acyl-carrier-protein] desaturase activity, EC:2.3.1.41 beta-ketoacyl-301 [acyl-carrier-protein] synthase I). The selection of these gene classes might be associated with 302 the main breeding objective in oil palm: maximizing oil yield per unit area for plantation 303 304 profitability (Corley and Tinker, 2003). As a direct comparison, in another oil crop, soybean, the main gene classes over-represented in the cultivated population were flowering time/floral 305 development, fatty acid biosynthesis and plant structure (height and pubescence) (Zhou et al., 306 2015). In rice, however, the main gene classes were flowering time/floral development, plant 307 growth and morphology (Xu et al., 2012). In all three species, we observed flowering or floral 308 309 development to be a key trait of importance for breeding programs. For the oil species, as expected, the fatty acid related classes were crucial. In the longest selective sweep of 310 Chromosome 2, we have identified the MADS-box transcription factor 16 gene, which is known 311 to be responsible for floral identity control and development (Moon et al., 1999). In oil palm, 312 female inflorescence development determines bunch yield components (Breure and Menendez, 313 1990). Even though allelic frequency can reflect artificial selection pressure on the genes, it is 314

also possible that some of these genes might be inherited through linkage drag with the oil yield-315 related genes selected under breeding. Also, given that the population base of the modern elite 316 breeding materials is narrow, it is possible that there might be a number of false positives. The 317 true causality can only be concluded through functional analysis, which is laborious. To narrow 318 the scope of functional studies, GWAS can be used in conjunction with selective sweep analysis 319 as additional evidence to identify QTL. Knowing that fruit and floral development is one of the 320 key gene classes found in the selective sweeps, identification of the important QTL in one of the 321 fruit components, S/F %, was selected to illustrate the applicability of this array in GWAS. 322

In oil palm, the fruit forms are determined as thick-shelled dura (Shell<sup>+</sup> Shell<sup>+</sup>), shell-less 323 324 pisifera (Shell Shell) and their hybrid, thin-shelled tenera (Shell Shell), in a co-dominant Mendelian inheritance pattern, first reported in the 1940s (Beirnaert and Vanderweyen, 1941). 325 The gene responsible for fruit form, determined by the presence or absence of shell and fiber 326 ring, has been reported to be SHELL, a gene homologous to the Arabidopsis SEEDSTICK (STK) 327 gene (Singh et al., 2013a). This gene is located at position 3,058,076bp of Chromosome 2. 328 However, within the tenera fruit form, varying degrees of thickness of the shell exists. 329 Presumably controlled by minor genes, it is known that there is polygenic variation in shell 330 thickness, superimposed onto the major SHELL effect (Corley and Tinker, 2003). The tenera 331 palms with a low S/F % trait have more mesocarp per fruit, leading to more oil production. 332 Given the importance of this trait in the *tenera* population, GWAS was carried out to study key 333 loci responsible for the variation in this trait that is not explained by the SHELL gene. Due to the 334 relatively small sample size of the populations, all of the 312 palms were treated as a single 335 group before performing association mapping. The kinship and population substructure of the set 336 were controlled for by using a kinship matrix. The S/F %-associated GWAS peak at 337 Chromosome 2 position 2,526,459bp to 3,905,306bp contains the published SHELL gene. This 338 339 could argue for allelic effects of different sources of SHELL and the surrounding regions. This 340 peak overlaps with the selective sweeps detected in Chromosome 2. It should be noted, however, that two independent signals on Chromosome 2 were also found farther away at 5,708,516bp and 341 13,250,110bp, and a few other signals were identified on other chromosomes. These additional 342 signals might indicate that there are other minor factors influencing S/F % in *tenera* palms. By 343 having a larger sample set and through fine mapping, these minor factors can be identified and 344 the markers combined to define a genetic ideotype for marker assisted breeding of shell-345

thickness. The new OP200K genotyping array has also been proven to be robust in detecting
significant associations for a polygenic complex trait, which is high mesocarp oil content in oil
palm (Teh et al., 2016).

As demonstrated, the association peaks, which is representative of the QTL, can be 349 detected through GWAS. In order to predict the total genetic breeding value of the genome 350 relative to a trait, instead of using only specific previously identified loci, the genetic effects of 351 352 the entire genome were evaluated. The approach adopted is known as genomic selection, where all QTL are expected to be in LD with at least one marker locus (Goddard and Hayes, 2007). As 353 previously shown, marker density for the current array is likely to fulfill this requirement, 354 355 potentially ensuring high prediction accuracy in breeding selection (Meuwissen et al., 2001b). S/F % trait was again selected to test the applicability of genomic selection using this array. Even 356 without addressing environmental variation between palms and populations, within population 357 substructures and with a limited sample size for the training population, a prediction accuracy of 358 about 0.65 was still achieved, thereby illustrating the viability of genomic selection with this 359 array for S/F %. The method used here for genomic selection is Bayes A (Meuwissen et al., 360 2001a). There are other methods that might give higher prediction accuracy and reduce 361 processing time for larger sample sizes, which will be the focus for future studies (Resende et al., 362 2012). 363

The applications of the OP200K genotyping array in genetic diversity assessment, 364 GWAS and genomic selection have proven to be robust and consistent using existing data for 365 palms in these important origins of oil palm. This SNP array has allowed the current mapping 366 367 resolution to be improved significantly and the array has become the densest genotyping array for oil palm. Hence, the array can be referred to as the first whole genome genotyping tool for oil 368 palm breeding programs. To optimize the genotyping cost, the array can further be customized 369 for the different BPROs developed in different breeding selection programs. From our 370 experience, at this time, the array-based approach is still far cheaper than genotyping by 371 sequencing for high-density marker analysis. The marker density can be reduced according to the 372 BPRO-specific LD decay analysis as shown in this study to ensure good genome coverage with 373 there likely to be at least one representative SNP in every LD block. Overall, the new OP200K 374 375 should facilitate genome-wide genetic discovery for the oil palm research community.

376

#### 377 Methods

#### 378 Whole genome resequencing

A total of 132 palms from 59 diverse origins were selected for resequencing using Illumina GAIIx, as reported previously (Teh et al., 2016). Genomic DNA (gDNA) for each palm was isolated from 100 milligram of dried leaf tissue using the DNAeasy Plant Mini Kit (Qiagen). The resulting gDNA samples were pooled, with equal amounts of gDNA from all samples. The DNA quantity and quality were then assessed using the NanoDrop ND-1000 Spectrophotometer. A total of eight paired-end (75bp each end) libraries with insert size of 100bp were constructed from the pool according to Illumina's standard protocol for whole genome shotgun sequencing.

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#### 387 SNP discovery and array design

The raw reads acquired from resequencing were trimmed to remove poor quality base calls (Phred<30) and mapped against both the published oil palm genome and in-house scaffolds using the BWA software (sampe and aln) with default parameters (Li and Durbin, 2010). SNP calls were made using samtools programs with parameters of -d 1000, -C 50, -D -S -g -u -q 20 m 3 (Li et al., 2009). The SNPs were filtered using the vcfutils script with parameters -Q25, -w 2, -W 2, -D 1000, -d 3. From the resulting vcf files, SNPs that had a mapping coverage between 17 and 53, a GQ score of >7, and a MAF of >0.05 were kept.

Only SNPs that had no secondary SNPs 60bp upstream and downstream were used. Other 395 396 criteria for removal of candidates included SNPs that were indels, were non-biallelic and SNP types that required two bead types on the array, i.e. A/T and C/G types. Ambiguous bases within 397 the 121bp probe region were also identified and removed. The probes were then sent to Illumina 398 to score for their designability. In-house scripts were used to select 200,000 SNPs with the best 399 Illumina designability, LD with  $r^2$  cut-off was set at 0.3 and the set chosen to give a uniform 400 distribution across the entire genome, based on the published and in-house genome. The 401 402 OP200K arrays were manufactured by Illumina in a 12-sample per array format.

403

#### 404 **SNP genotyping**

The OP200K arrays were used to genotype a panel of 312 accessions from commercial oil palm populations of UR x AVROS and JL x AVROS (66 and 33 individuals), a semi-wild population of Nigerian x AVROS (101 individuals) and breeding populations of GM x DA and JL x DA (13 and 99 individuals).

For each of the accessions, gDNA was extracted using the same method as described under the resequencing step. Prior to hybridization to the bead arrays, DNA was diluted to 25ng/ul and DNA quantification was obtained with Hoechst (33258 Pentahydrate (Invitrogen), using the FLUOstar Omega (BMG Labtech)). DNA quality was assessed on a 0.8% agarose gel. The genotyping was carried out using the designed array on the Infinium iScan platform (Illumina Inc, San Diego, CA) according to the recommendations of the manufacturer.

The raw intensity SNP data was analyzed using GenomeStudio version 20011.1 by Illumina with genotyping module version 1.8.4. Using a GenCall score cutoff of 0.15, autocluster of the SNPs was done. Cluster refining of the SNP clusters was done manually by visual inspection so that identifiable and scorable clusters were generated. The SNP calls were exported into the PLINK program for MAF and call rate filtering (Purcell et al., 2007). A minimal call rate of 90% and a MAF filter of 0.01 were set as baseline cut-offs.

421

#### 422 Phenetic analysis

The SNP genotypes of 312 tenera palms were compiled. Missing SNP genotypes were 423 imputed using the mean value for each of the populations in this study. The genotypes were first 424 425 recoded into 0 (AA), 1 (AB) and 2 (BB). Pairwise genetic distance was estimated using the dist.dna function of the APE package in R (Paradis et al., 2004). The dendrogram was generated 426 427 using the njs function, which implements the neighbor-joining clustering method (Saitou and 428 Nei, 1987). The bootstrap consensus tree was inferred from 1,000 replicates and the other parameters were used as default. Illustration of the tree was done using FigTree version 1.4.2, 429 acquired from http://tree.bio.ed.ac.uk/software/figtree/. 430

431

#### 432 Genome-wide LD decay

Genome-wide LD was estimated by calculating pairwise correlation coefficient  $(r^2)$ values between all pairs of SNPs in a 1-Kb window size within each chromosome across the five clusters of oil palm. An inter-SNP distance of more than 10Mb was ignored. The LD decay rate was defined as the chromosomal distance at which the average  $r^2$  dropped to half of its maximum value. For this step, the R package of SNPRelate was used, with the LD estimation based on the composite method (Zheng et al., 2012).

439

#### 440 Mapping of selective sweeps

This step was carried out using XP-CLR software (Chen et al., 2010). The window size of 0.1 cM and a grid size of 20000 kb was set for this step. Selective sweeps identified side by side were combined. Top 1% of the XP-CLR scores was selected as the threshold to identify selective sweep (Kim et al., 2015). The result was illustrated in chromosomal format using MapChart version 2.30, with the density of the chromosomal map spaced by 0.5Mb (Voorrips, 2002).

Identification of genes that fall into these regions was determined using published 447 genome information (Singh et al., 2013b). These genes were clustered based on GO by using 448 BLAST2GO (Conesa and Gotz, 2008). The Chi-square method implemented in R was used to 449 test for the enrichment of putative artificially selected genes in all the GOs/ECs identified. With 450 a *p*-value cut-off set at 0.05, a GO/EC class was considered to be significantly enriched if the 451 number of putative selected genes belonging to that class were significantly more than the 452 expected number calculated for that particular class across the significance region from the total 453 number of genes. 454

455

456 GWAS & Genomic selection

The polygenic shell thickness of 312 tenera palms was measured according to weight 457 difference between shell and the total fruit as S/F %. The individual palms were phenotyped to 458 generate a reliable mean S/F % value for analysis as per industry standards with modifications 459 (Blaak et al., 1963; Rao et al., 1983). GWAS was then carried out using the GenABEL package 460 in R, which implements a mixed linear model to detect association signals (Aulchenko et al., 461 2007). To control for cryptic relatedness in the population, a kinship matrix was generated by the 462 ibs function in the package, which computes the identity by state for the markers data. The 463 default weight parameter for the kinship matrix construction was used. The association score was 464 calculated using the mmscore function, with all the parameters set to default. Population inflation 465 was estimated using the estlambda function implemented in the package. A threshold  $-\log_{10}(p-1)$ 466 *value*) of 5.0 was set to identify an association peak. 467

The genotype data used for GWAS was also used for genomic selection. A 5-fold cross validation was carried out to separate the data into training set and validation sets. Genomic selection was done using Bayes A under the BGLR package in R (Perez and de los Campos, 2014). The number of burn-in iterations was set to 10,000 and the Gibbs sampler iteration was set to 20,000. Accuracy was measured through the correlation between the GEBV versus the realized S/F % trait value.

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#### 475 **Data availability**

All the raw data used in this study are publicly available. The sequencing raw reads have
been deposited in SRA under Bioproject PRJNA288621 and Accession ID of SRR2132870,
SRR2132879, SRR2132880, SRR2132872, SRR2132874, SRR2132884, SRR2132881,
SRR3110655, whereas the SNPs used in the array have been deposited in dbSNP under the
handle of SDTC\_BB with NCBI submitted SNP (ss) accession numbers of 18100692401810592638.

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#### 483 Author Contributions

Conceptualization, Q.B.K., C.K.T., A.L.O. and F.T.C.; Methodology, Q.B.K., C.K.T., A.L.O.;
Software, Q.B.K., A.L.O. and H.Y.H.; Formal Analysis, Q.B.K., C.K.T. and A.L.O.,
Investigation Q.B.K., C.K.T., A.L.O., H.Y.H. and H.L.L., Resources, M.M., A.S.; Data
Curation, Q.B.K, A.L.O. and J.Z.L.; Writing – Original Draft, Q.B.K, C.K.T and A.L.O.;
Writing – Review & Editing, H.Y.H, A.S., S.M., F.T.C, H.K., M.T. and D.R.A.; Visualization,
Q.B.K., A.L.O. and H.Y.H; Supervision, H.K., M.T. and D.R.A.; Project Administration,
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