

- **Supplementary information text**
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 Plant material for the reference genome. About 25 g of young leaves were collected from a mature individual of *M. polymorpha* var. *incana* (NG4) maintained in a coldframe at the University of Hawaii Hilo. NG4 was produced through a controlled-cross between two trees of this variety occurring along Kuliouou Trail, Oahu (approximate location: 21.3160, -157.7296). Leaf material was collected in small batches, wrapped in aluminum foil, and submerged in liquid N₂ within one minute of collection. The bundled samples were then kept at -80C for 72 hours and shipped overnight on dry ice to Oxford Nanopore Technologies, New York, NY and stored again at -80C. The entire tree was covered in a black plastic bag (dark-treated) for 24 hours prior to collection. **Nanopore sequencing-based whole-genome-, RNA-, and Pore-C sequencing.** Using the Qiagen DNeasy Plant Mini Kit, DNA was extracted from 2 g of collected leaf tissue. Separately, total RNA was extracted from 1 g of collected leaf tissue using Thermofisher's PureLink RNA Mini Kit. Full-length cDNA was synthesized from 50 ng of total RNA using the Oxford Nanopore Technologies PCS109 kit, followed by 14 rounds of PCR amplification using the primer mixture from the Oxford Nanopore Technologies EXP-PCA001 kit. A detailed protocol outlining the Pore-C method can be found as supplemental text (https://doi.org/10.5281/zenodo.4264399). A sequencing library was prepared using the Oxford Nanopore Technologies standard ligation sequencing kit SQK-LSK109. Sequencing was conducted on a GridION X5 sequencer for 72 hours, and the raw data were base-called by Oxford Nanopore Technologies basecaller Guppy (available on https://community.nanoporetech.com/) ver. 3.2.8 for the genomic DNA and Pore-C DNA, and ver. 3.2.10 for the cDNA in the high-accuracy mode.

Nanopore sequence-based reference genome assembly. The FASTQ files from the whole-

- genome sequencing data were filtered for high-quality long reads. We used the program filtlong
- 52 (https://github.com/rrwick/Filtlong) with parameters --min length 10000 --min mean q 85 --
- 53 min window q 70, which selects for reads longer than 10 kbp, average Q-score greater than 8.2,

 and a minimum sliding window Q-score of 5.2. The filtered nanopore reads were then assembled with the genome assembler flye (1).

 Assembly contig scaffolding with Pore-C sequencing data. The Pore-C data analysis was conducted using the Pore-C workflow developed by Oxford Nanopore Technologies (https://github.com/nanoporetech/Pore-C-Snakemake), which uses the snakemake workflow engine (2). Briefly, the workflow first aligns the nanopore Pore-C chromosome contact sequence reads to the unscaffolded *Metrosideros* genome assembly using bwa-sw ver. 0.7.17-r1188 (3) with parameters -b 5 -q 2 -r 1 -T 15 -z 10. Compared to conventional Hi-C data, Pore-C contains an enrichment of higher order contacts (4),and to process the multi-contact nanopore reads we used Pore-C tools (https://github.com/nanoporetech/pore-c) also developed by Oxford Nanopore Technologies. The alignment BAM file was processed with Pore-C tool to filter spurious alignments, detect ligation junctions, and assign fragments that originated from the same chromosomal contacts. Pore-C tools converted the alignment BAM file to a chromosome contact pairs format (https://github.com/4dn-dcic/pairix/blob/master/pairs_format_specification.md) for compatibility with the conventional downstream chromosome contact based analysis methods. The pairs file was converted to a hic file format using the Juicer ver. 1.14.08 tools (5) to use as the input data for the Juicebox assembly tools (6). The Juicebox tools were used to scaffold the draft *Metrosideros* assembly, and we followed established guidelines (https://github.com/theaidenlab/Genome-Assembly-Cookbook) to manually construct the chromosome-scale scaffolds using the Pore-C based chromosome contact frequency information. We assigned chromosome numbers to the superscaffold through synteny with the *Eucalyptus grandis* genome assembly (7). Synteny between the *Metrosideros* superscaffold and *Eucalyptus* chromosomes was determined by aligning the assemblies to each other and visualizing the alignment through the program D-GENIES (8). **Genome annotation.** The cDNA library that was sequenced on the nanopore sequencer was used to annotate the coding sequence regions for the *M. polymorpha* genome assembly. Initially, 82 we used Pychopper ver. 2.3.1 (https://github.com/nanoporetech/pychopper) to trim primers, identify full-length cDNA sequences, and orient the sequence to the correct strand. A total of

12,298,201 (70.2%) reads were classified by Pychopper and were used for downstream analysis.

 The long reads were aligned to the reference genome using minimap2 ver. 2.17-r941 (9) with options -ax splice -uf -k14. The alignment file was then used by stringtie2 ver. 2.1.3b (10), which is optimized for *de novo* transcriptome assembly using long-read sequencing and a reference genome. We used the MAKER program (11) for gene annotation using the workflow outlined on 89 the website https://gist.github.com/darencard/bb1001ac1532dd4225b030cf0cd61ce2. The transcriptome assembly from stringtie2 was used as EST evidence in MAKER, and the protein sequences from the previous *M. polymorpha* assembly (12), *E. grandis* assembly (7), and *A. thaliana* (TAIR10) were used for a protein homology search in MAKER. After an initial round of MAKER annotation the gene models were used by SNAP (13) and Augustus (14) to create gene model training datasets specifically for our *M. polymorpha* genome assembly. The training dataset was used for a second round of MAKER gene annotation.

 We identified the repetitive regions of the *M. polymorpha* reference genome, first using Repeatmodeler ver. 1.0.10 (http://www.repeatmasker.org/RepeatModeler/) for the *de novo* identification repeat sequences in the reference genome, and then using Repeatmasker ver. 4.1.0 (http://www.repeatmasker.org/RepeatMasker/) to identify the genomic locations of the repetitive sequences in our reference genome.

 Metrosideros **population sequencing.** For the population genomic sampling, we collected young leaf tissue from 9-11 adults from each of eight taxa on the focal island of Oahu and fewer adults of two taxa on Kauai. Collected leaf tissue was kept cool and stored at -80C within 48 hours of collection. Leaf material from the three outgroup samples was silica-dried in the field and stored in a dessicator jar. DNA was extracted from both frozen and dried leaf samples using the Macherey-Nagel NucleoSpin Plant II Mini kit. We used a Tn5 transposase-based method to prepare the whole-genome sequencing library. Mosaic End adaptor A and B (Tn5ME-A:

TCGTCGGCAGCGTCAGATGTGTAT AAGAGACAG; Tn5ME-B:

GTCTCGTGGGCTCGGAGATGTGTATAAGAGACAG) was annealed with Rev (Tn5ME-

Rev: /5Phos/CTGTCTCTTATACACATCT) by mixing 10uL (100uM) of each oligonucleotide

with 80 uL of reassociation buffer (10 mM Tris pH 8.0, 50 mM NaCl, 1 mM EDTA) in BioRad

113 thermocycler with the following program: 95°C for 10 min, 90°C for 1 min, and decrease

114 temperature by 1°C/cycle for 60 cycles, held for 1 min at each temperature. Pre-charge of Tn5

with adapters was carried out in solution by mixing 22.5 uL of 100 ng/uL Tn5 (Tn5 protein was

116 produced following the protocol described by (15)), 76.5 uL reassociation buffer/glycerol (1:1),

- and 4.5 uL of equal molar of annealed adaptor 1 (A-Rev) and annealed adapter 2 (B-Rev). The
- 118 reaction was then incubated at 37°C for 30 min. The annealed adapters bind to Tn5 transposase
- to form the transposome complex.

 Genomic DNA was tagmented by mixing with 1 uL of the above assembled Tn5 transposome, 4 uL of 5 X TAPS buffer (50 mM TAPS-NaOH pH 8.5 [Alfa aesar # J63268], 25 mM MgCl2, 50% v/dimethylformamide [ThermoFisher #20673], pH 8.5 at 25°C) and water to a 123 total volume of 20 uL and incubated at 55°C for 7 min. The transposome fragments and attaches adapters to gDNA. The reaction was completed by adding 5 uL of 0.2% SDS (Promega, #V6551) to each reaction followed by incubation at 55°C for 7 min to inactivate and release the Tn5. To enrich the DNA fragments that have adapter molecules on both ends we attached an index to the library: 2µl of the stopped tagmentation, 1µl i5 index primer (1uM), 1µl i7 index 128 primer (1 μ M), 10 μ l of OneTaq HS Quick-Load 2x master mix (NEB #M0486L), and 6 μ l of 129 water were combined to make a 20-ul final reaction. The reaction was heated at 68 °C for 3min 130 and 95°C for 30 sec, then thermocycled 12 times at 95°C for 10 sec, 55°C for 30 sec, and 68°C 131 for 30 sec, followed by a final extension of 5 min at 68 °C. We then pooled the libraries together, taking 5uL from individual libraries. The pooled

 library was cleaned and size-selected using Agencourt AMPure XP beads (Beckman Coulter, #A63881) at a 0.8:1 (beads: DNA) ratio. The final library was quantified with a Qubit high- sensitivity DNA kit (Invitrogen Q32854) and examined on an Agilent 2100 Bioanalyzer high-136 sensitivity DNA chip (Agilent p/n# 2938-85004) to observe the library size distribution. The 137 sequencing library was loaded on a NovaSeq 6000 S1 flow cell and sequenced under a 2×150 -bp conformation at the Genomics Core Facility within the Lewis-Sigler Institute for Integrative Genomics at Princeton University.

 Calling genome-wide polymorphisms. Raw sequencing reads were downloaded from our previous study (16) and combined with the newly generated sequencing data from the current study. From the sequence read archive (SRA) website we downloaded FASTQs with SRR identifiers SRR8943660 to SRR8943653. The sequencing reads were adapter-trimmed and quality-controlled using BBTools (https://jgi.doe.gov/data-and-tools/bbtools/) bbduk program 146 version 37.66 with option: minlen = 25 qtrim = rl trimq = 10 ktrim = r k = 25 mink = 11 hdist = 1 147 tpe tbo.

 Sequencing reads were then aligned to the scaffolded *M. polymorpha* reference genome generated from this study using bwa-mem. PCR duplicate reads were removed using picard version 2.9.0 (http://broadinstitute.github.io/picard/). Genome-wide read coverage statistics were calculated using GATK version 3.8–0 (https://software.broadinstitute.org/gatk/).

 We used the GATK HaplotypeCaller engine to call variant sites from the BAM alignment file for each sample. The option –ERC GVCF was used to output the variants in the gVCF format, and the gVCFs of each sample were merged together to allow a multi-sample joint genotype procedure using GATK GenotypeGVCFs engine. Using standard GATK best-practice hard-filter guidelines, the VariantFiltration engine was used to filter out low-quality polymorphisms. In addition, we removed SNPs that were within 5 bp of an INDEL and

polymorphic sites that had less than 80% of individuals with a genotype call.

 Population relationship analysis. The alignment BAM files were used to analyze the population relationships between samples. We used ANGSD version 0.929 (17) and ngsTools (18) to analyze the genotype likelihoods of each sample and infer the population relationships using a probabilistic framework. We only analyzed potential variant sites where more than 80% of the individuals had a genotype, while enforcing a total sequencing coverage filter such that 165 included sites had a minimum of $1/3$ the average total sequencing depth (734 \times) and a maximum 166 of three times the average total sequencing depth $(6,613\times)$. To minimize the effect of linkage on inferences on population relationships, polymorphic sites were randomly pruned using a 10-kbp 168 sliding window with a minimum distance of 5 kbp between random sites.

 NGSadmix (19) was used to estimate the admixture proportions (K) for each individual. 170 For each of K = 3 to 15, the analysis was repeated 100 times and the run with the highest log- likelihood was chosen. Principal component analysis was also conducted using genotype likelihoods. Genotype posterior probabilities were calculated using ANGSD and used by the program ngsCovar (18) to conduct the principal component analysis. Phylogenetic relationships were investigated using the hard-called genotypes from GATK and extracting four-fold degenerate sites using a Python script (available at

https://github.com/tvkent/Degeneracy). A maximum-likelihood phylogenetic tree was

 reconstructed through RAxML ver 8.2.12 (20) using the –f a algorithm (*i.e.* the rapid bootstrap analysis and search for best-scoring maximum-likelihood tree). 100 bootstrap replicates were generated to obtain confidence in the tree topology. We also conducted Bayesian phylogenetic analysis using the SNAPP package as part of the program BEAST ver. 2.6.3 (21, 22). Because SNAPP is a resource-intensive method, for each taxon/population we randomly selected four individuals and selected for biallelic polymorphic sites. Sites were randomly pruned in 10-kbp windows to generate unlinked biallelic sites, which are required for SNAPP. Using BEAST, an MCMC analysis was run for 3,000,000 generations with sampling every 1,000 generations. At the end, 30% of the chains were discarded as burn-in.

 Investigating reticulate evolutionary history. The genotype call dataset was used to examine the reticulate evolutionary history of *Metrosideros*. We conducted the ABBA-BABA D test (23, 24) using the R package admixr (25), which is based on the ADMIXTOOLS suite (26). Variants were polarized using the high-coverage Fiji sample *M. vitiensis* as the outgroup genome. The topologies used were based on the maximum-likelihood tree, the SNAPP-based tree, and the TWISST analysis. Specifically, across the island we assumed the topology [(Hawaii, Oahu), Kauai], which was supported by PCA, maximum-likelihood tree, and SNAPP based tree. Within islands, only Oahu had topologies that differed between the maximum-likelihood and SNAPP trees. Within Oahu, we examined the TWISST result and chose the topologies that were most 196 frequent to represent relationships (*i.e.* pubescent group $= [(I, C), (F, R)]$; glabrous group $=$ $[(B,L),(M,T)].$

 The genome-wide topological relationships were examined using the method TWISST (27). The genotype call dataset was imputed and phased using the program Beagle ver. 5.0 (28, 29). The phased genotype dataset was used to conduct a sliding window-based estimation of the 201 local phylogenetic relationships. Using the raxml sliding windows.py script from the genomics_general package 203 (https://github.com/simonhmartin/genomics_general/tree/master/phylo), phylogenetic trees were reconstructed in windows of 200 polymorphic sites. The option 'complete' within the TWISST program was used to calculate the exact weighting of each local window.

 Demographic modeling. We used the methods δaδi (30), G-PhoCS (31), and MSMC (32, 33) to infer the demographic history of Hawaiian *Metrosideros*. For all analyses we used the genotype call dataset.

 For δaδi analysis, we initially randomly thinned the dataset picking a SNP every 10 kbp using PLINK ver. 2.0 (34). The site frequency spectrum was estimated using the easySFS.py (https://github.com/isaacovercast/easySFS) script while using Fiji *M. vitiensis* as the outgroup genome and polarizing the polymorphic sites. The easySFS.py script was also used to project down the sample size to maximize the number of sites analyzed. The unfolded site frequency spectrum data were used as input for δaδi, and we fit 20 demographic models (Supplemental Figure 9). We optimized the model parameter estimates using the Nelder-Mead method by randomly perturbing the parameter values for four rounds. The parameter estimates were 218 perturbed threefold, twofold, twofold, and onefold in incremental rounds. Each round the perturbation was conducted for 10, 20, 30, and 40 replicates. Demography parameters were extracted from the round with the highest likelihood. Demographic models were compared using Akaike Information Criteria (AIC) values. The δaδi analysis scripts were based on the study by (35).

 To prepare our dataset for G-PhoCS analysis we first partitioned our reference genome into 1-kbp loci and determined those that are close to neutrality. Neutral loci were determined by 225 selecting loci that were 5 kb away from a genic sequence, 500 bp away from a repetitive DNA sequence, and at least 10 kbp away from each other. Since G-PhoCS is designed to analyze the variation within a single genome, we selected a single individual with high genome coverage to 228 represent each island. Selected samples included: H207 from Hawaii Island (population G_{H1}), 229 X83 from Molokai (population G_M), O385 from Oahu (taxon M), and K283 from Kauai 230 (population G_K). The *M. vitiensis* sample from Fiji was used as the outgroup. We used G-PhoCS ver. 1.2.3 and ran every demographic model five times to check for convergence in the demographic parameter estimates. Each MCMC run had 1,000,000 iterations, and the initial 500,000 iterations were discarded as burn-in. Priors were modeled using a gamma distribution $(\alpha = 1 \text{ and } \beta = 10,000 \text{ for population size and divergence time}; \alpha = 0.002 \text{ and } \beta = 0.00001 \text{ for}$ migration rates). Different demographic models involved fitting migration bands between two terminal lineages. After the MCMC run was complete, the program Tracer version 1.6 237 (http://tree.bio.ed.ac.uk/software/tracer/) was used to estimate the 95% highest posterior density

238 for each demography parameter. The G-PhoCS-estimated divergence time τ is scaled according

239 to the mutation rate (μ) . To convert divergence time to absolute divergence time T (in years) we

used the following equation:

$$
T = \frac{\tau \times g}{\mu}
$$

where g represents the generation time.

 MSMC2 was used to estimate the past changes in effective population sizes and the divergence times between individuals. From each taxon/population we chose a single representative individual for the MSMC2 analysis as follows: O310 (taxon B), O65 (taxon C), O194 (taxon L), H271 (taxon N), O464 (taxon R), O145 (taxon T), plus the individuals used in the G-PhoCS analysis. We used the alignment BAM file for each individual and the mpileup function of samtools ver. 1.3.1 (36) to detect sites that had a minimum base score of 30, a mapping quality score of 30, and the coefficient to downgrade mapping qualities for excessive mismatches at 50. The resulting text pileup output was used by bcftools ver. 1.3.1 to call variant sites but excluding INDELs and limiting the calls to biallelic SNPs. The bamCaller.py script that was provided by the MSMC suite (https://github.com/stschiff/msmc-tools) was then used to produce the per-chromosome masks and VCF files for each individual. A genome-wide mask file was also created for each individual by using the SNPable workflow (http://lh3lh3.users.sourceforge.net/snpable.shtml) on each superscaffold/chromosome assembly and then converting it to BED format using the makeMappabilityMask.py script within the MSMC suite. Phasing was done by using the output from the beagle analysis. The input files for MSMC2 were generated using the generate_multihetsep.py script, which is also part of the MSMC suite. To estimate changes in effective population size we examined the two haplotypes of each individual, while cross-coalescence rates were estimated from four haplotypes from two individuals. The combineCrossCoal.py script from the MSMC suite was used to produce the outputs for plotting. **Simulations.** Simulations were conducted using the forward-time Wright-Fisher model-based simulator SLiM (37). All of the SLiM code used is available on GitHub

265 (https://github.com/ornobalam/metrosideros simulations). In all described simulations, the

 starting population was allowed to evolve for a burn-in period of 50,000 non-overlapping generations to generate the ancestral population.

 For observing tree topologies, the basic model consisted of an ancestral population (ABCD) with a fixed population size of N that split after 50,000 non-overlapping generations into 2 daughter populations AB and CD, each with a fixed size of N. An outgroup population O 271 splits off from ABCD in the first generation with a fixed size of N. AB and CD were then allowed to evolve for a further 15,500 and 5,000 non-overlapping generations before splitting 273 into daughter populations A and B, and C and D, respectively, each with a fixed size of N. A, B, C, and D were then allowed to evolve for 150,000 generations since the origin of the ancestral population. The split times for A, B, C, and D were chosen based on estimated split times for the four sampled glabrous *Metrosideros* taxa from Oahu, using the cross-coalescence rate estimates from MSMC2. We simulated a 1-Mbp chromosome with a recombination rate ranging between 10^{-9} and 10^{-7} , and mutation rate of 10^{-8} , both per base pair and per generation. The chromosome comprised 30% genic regions with a 1:4 ratio of neutral to deleterious mutations, and 70% non-coding regions with only neutral mutations.

281 We explored three population sizes (N) of the ancestor (ABCD): (1) a large $N = 100,000$ 282 (2) a medium N = 50,000 and (3) a small N = 10,000. Ancient migration was modeled with a rate, m1, between AB and CD, and recent migration was modeled with a rate, m2, among A, B, C, and D (see Fig. S6 for visual representation of simulation model). Migration was exclusively either ancient or recent in each simulation. For each N we simulated two different migration rates of 0.1 or 0.001 for m1 or m2. At the end of each simulation, 10 individuals were sampled and outputted from populations A, B, C, D, and O, for analysis with TWISST.

 Population genomic analysis. We used the gVCFs that were called from the previous step to create a VCF file that had genotype calls for all sites including the non-variant positions. The GATK GenotypeGVCFs engine was used with the option –includeNonVariantSites. We analyzed a population VCF that had both variant and non-variant sites in order to obtain the correct number of sites to be used as the denominator for the population genetic statistics we were calculating. We used the genomics_general package 295 (https://github.com/simonhmartin/genomics general) to calculate θ , D_{xy} , and F_{ST} in 10-kbp

windows, sliding the window by 5 kbp. For each window we imposed a quality filter only

297 analyzing sites that had a minimum quality score of 30 and a minimum depth of $5\times$. Windows that had more than 30% of the sites with a genotype call after filtering were chosen for downstream analysis.

 Evidence of admixture in localized windows (38) was investigated using the 301 abbababawindows.py script from the genomics general package to calculate the f_{dm} statistics. To 302 search for genomic outliers of differentiation, the F_{ST} values were *Z*-transformed (zF_{ST}), and 303 genomic windows with $zF_{ST} \ge 4$ were considered outliers (39).

 The strength of evidence of selective sweeps was estimated using the ω statistic (40). We used the program OmegaPlus (41), which is specifically designed to estimate the ω statistics for 306 genome-wide SNP datasets. We set the grid size of OmegaPlus so that the ω statistics would be estimated at 10-kbp windows for each super-scaffold/chromosome.

 Allele state and count per polymorphic site was determined using PLINK and used for analyzing the evolutionary origin of differentiation outliers.

Gene Ontology Enrichment. Coding sequences of each gene model were assigned a

computationally predicted function and gene ontology using the eggnog pipeline (42). We

required an ontology to have more than two gene group members for further consideration. Gene

ontology enrichment was tested through a hypergeometric test.

Fig. S2. Pairwise average F_{ST} between taxa/populations.

Fig. S3. A cloudogram generated from SNAPP-reconstructed population topology. The different colors represent different species topologies. Topologies with the highest clade credibility are shown in blue, while red represents minor trees with an alternative topology.

Fig. S4. ABBA-BABA D-test statistics for all taxon/population trio combinations.

Fig. S5. TWISST-estimated topology weightings for the 15 possible topologies for a rooted 4 taxon tree. Shown are topology weights for glabrous (top) and pubescent (bottom) *Metrosideros*

Fig. S6. Simulations examining the effects of population size (N_e) and gene flow on the topology weights. (A) The simulation scenario that is based on the history of the four glabrous taxa from Oahu. Topology

Fig. S7. TWISST topology weights involving taxa/populations G_{H1}, M, and G_K Topologies were

Fig. S8. G-PhoCS-based divergence time estimates for different migration models.

Fig. S9. G-PhoCS-based divergence time estimates assuming different mutation rates and generation times.

Fig. S10. The relative cross-coalescence rates for the four phylogenetic sister pairs. A crosscoalescence rate of 0.5 was assumed to reflect the divergence time between the paired taxa. Divergence times are indicated with dotted lines and shown in the key. Relative times were converted to absolute times assuming a mutation rate of 7e-9 mutations per base pair per

Fig. S11. The 20 different δaδi scenarios that were modeled.

Fig. S12. The number of overlapping zFST outlier positions.

Fig. S13. Relative node depth (RND) values (calculated in 10-kbp window) for differentiation (F_{ST}) outliers identified for each sister pair. Red boxes are statistics from the genomic background, and green boxes are statistics from the differentiation outlier regions. * indicate significant difference with Mann-Whitney U test after Bonferroni correction; p < 0.05.

Fig. S14. Evidence of selection (calculated in 10-kbp window) in differentiation outlier regions. Red boxes are statistics from the genomic background, and green boxes are statistics from the differentiation outlier regions. (A) Levels of polymorphism in the differentiation outlier regions. (B) Selective sweep statistics (ω_{max}) in the differentiation outlier regions. * indicate significant difference with Mann-Whitney U test after Bonferroni correction; p < 0.05.

Fig. S15. Repeat sequence density in differentiation outlier regions compared to genome-wi 10-kbp windows. Red boxes are statistics from the genomic background, and green boxes are statistics from the differentiation outlier regions.

Fig. S16. Significantly enriched gene ontology terms for genes overlapping genomic regions of differentiation outliers.

Fig. S17. Maximum-likelihood phylogenetic tree of the differentiation (F_{ST}) outlier regions identified for each sister pair in Fig. 3B. For each phylogeny the topological position of the sister pair that was examined is indicated with star and the topologically incongruent taxa from the pair is indicated with red star. Nodes with greater than 95% bootstrap support are indicated with blue circles.

Fig. S18. Levels of Dxy between a taxon of the sister pair and all other Hawaiian *Metrosideros* taxa sampled. Red boxes are statistics from the genomic background, and green boxes are statistics from the differentiation outlier regions identified in a sister pair. Significant differences are shown as $*$ for p < 0.05, $**$ for p < 0.01, and $***$ for p < 0.001 following Mann Whitney U test with Bonferroni correction.

Fig. S19. Classification of sites polymorphic within a sister pair [B-L (top), C-R (middle), M-T (bottom)] into four categories according to the distribution of allele states on other islands (see text and Fig. 6 for visual representation of the four categories). Bars show the proportion within each category of sites that are polymorphic within the differentiation outlier regions of the focal sister pair. Numbers within the bars indicate the total number of SNPs within each category from the focal sister pair. ** indicate $p < 0.01$ and *** indicate $p < 0.001$ after Fisher's exact test.

Flowcell	Number of reads	Median read length	Read length N50	Median quality score(QS)	Total bases
			Raw		
FAL75935	1,821,430	2,929	13,392	11.1	11,244,606,349
FAL76650	1,909,759	2,895	13,081	11.1	11,574,820,609
FAN08536	483,618	22,785	29,641	11.3	11,013,147,889
Total	4,214,807	3,433	18,983	11.1	33,832,574,847
			Filtered		
FAL75935	294,849	16,568	19,761	12.5	5,458,999,683
FAL76650	301,870	16,519	19,679	12.4	5,574,321,352
FAN08536	282,078	25,613	29,484	12.6	7,694,845,247
Total	878,797	19,185	23,559	12.5	18,728,166,282

Table S1. Sequencing statistics from the nanopore sequencing of the whole genome.

Sequencing statistics					
12,235,945					
12,251,510,900 bp					
847 bp					
$1,174$ bp					
11 1					

Table S3. Sequencing statistics from the nanopore sequencing of the cDNA library.

Table S4. Population sa

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