1 Invaders weather the weather: rapid adaptation to a novel environment occurs
2 despite reduced genetic diversity
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4 Daniel Selechnik [1,2*], Mark F. Richardson [3,4], Richard Shine [1], Jayna DeVore [1], Simon
5 Ducatez [1], and Lee A. Rollins [2]
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[1] School of Life and Environmental Sciences (SOLES), University of Sydney 2006
[2] Evolution & Ecology Research Centre, School of Biological, Earth and Environmental
Sciences, University of New South Wales, Sydney, NSW 2052
[3] Deakin Genomics Centre, School of Life and Environmental Sciences, Deakin University,
Locked Bag 20000, Geelong, VIC, Australia 3216
[4] Centre for Integrative Ecology, School of Life and Environmental Sciences, Deakin
University, Locked Bag 20000, Geelong, VIC, Australia 3216
[*] Correspondence: Dan Selechnik, Phone: +61 452 304 787; E-mail:
danselechnik@gmail.com

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Abstract

Invasive species often exhibit rapid evolution in their introduced ranges despite the genetic

bottlenecks that are thought to accompany the translocation of small numbers of founders;

however, some invasions may not fit this "genetic paradox." The invasive cane toad (*Rhinella*

marina) displays high phenotypic variation across its environmentally heterogeneous

introduced Australian range. Here, we used three genome-wide datasets to characterize

population structure and genetic diversity in invasive toads: RNA-Seq data generated from

spleens sampled from the toads' native range in French Guiana, the introduced population in

Hawai'i that was the source of Australian founders, and Australia; RNA-Seq data generated

from brains sampled more extensively in Hawai'i and Australia; and previously published

RADSeq data from transects across Australia. We found that toads form three genetic

clusters: (1) native range toads, (2) toads from the source population in Hawai'i and long-

established areas near introduction sites in Australia, and (3) toads from more recently

established northern Australian sites. In addition to strong divergence between native and

invasive populations, we find evidence for a reduction in genetic diversity after introduction.

However, we do not see this reduction in loci putatively under selection, suggesting that

genetic diversity may have been maintained at ecologically relevant traits, or that mutation

rates were high enough to maintain adaptive potential. Nonetheless, cane toads encounter

novel environmental challenges in Australia and appear to respond to selection across

environmental breaks; the transition between genetic clusters occurs at a point along the

invasion transect where temperature rises and rainfall decreases. We identify loci known to

be involved in resistance to heat and dehydration that show evidence of selection in

Australian toads. Despite well-known predictions regarding genetic drift and spatial sorting

during invasion, this study highlights that natural selection occurs rapidly and plays a vital

role in shaping the structure of invasive populations.

Author Summary

Despite longstanding evidence for the link between genetic diversity and population viability, the "genetic paradox" concept reflects the observation that invasive populations are successful in novel environments despite a putative reduction in genetic diversity. However, some recent studies have suggested that successful invasions may often occur due to an absence of obstacles such as genetic diversity loss or novel adaptive challenges. The recent emergence of genome-wide technologies provides us with the tools to study this question comprehensively by assessing both overall genetic diversity, and diversity of loci that underlie ecologically relevant traits. The invasive cane toad is a useful model because there is abundant phenotypic evidence of rapid adaptation during invasion. Our results suggest strong genetic divergence between native and invasive populations, and a reduction in overall genetic diversity; however, we do not see this reduction when solely assessing ecologically relevant loci. This could be for reasons that support or refute the genetic paradox. Further studies may provide perspectives from other systems, allowing us to explore how variables such as propagule size affect the fit of an invasion to the model of the paradox. Studying invasive species remains important due to their largely negative impacts on the environment and economy.

Introduction

The genetic paradox of invasion [1] describes a phenomenon that challenges widespread

evidence of the relationship between genetic diversity and adaptive potential. High genetic

diversity within a population is beneficial because it likely underlies phenotypic variation

across individuals, allowing the population to respond to selection imposed by environmental

change [2, 3]. Furthermore, a greater number of alleles confers an increased frequency of

heterozygosity, which is often associated with population fitness [2]. Small or isolated

populations with low genetic diversity have been shown to suffer declines due to inbreeding

depression and the associated reduction of individual fitness [4-6]. Conservation efforts to

salvage such populations by introducing conspecific individuals from allopatric populations

(thereby introducing new alleles into the population; "genetic rescue") have been successful,

suggesting that the maintenance of genetic diversity can be crucial for population viability [5,

6].

Despite the fact that invasive populations are thought to undergo genetic bottlenecks

(due to the translocation of a small number of founders from their native range to an

introduced range [1, 7]), invasive species are also characterized by their ability to establish

and spread in their introduced ranges. Invasion success is commonly linked to rapid

evolution, including adaptation to novel environmental conditions over short timescales [8-

11]. Additionally, some invaders exhibit novel phenotypic traits that enhance invasive

potential, such as increased growth and dispersal rates [11, 12]. There are many examples of

evolutionary change during invasion without high levels of genetic diversity [13].

Although low genetic diversity may limit the ability of an invasive population to

respond to natural selection, rapid evolution can also occur through non-adaptive processes:

(1) genetic drift may occur on range edges, reducing genetic diversity across an introduced

range [14]; (2) due to spatial sorting, the invasion front may be inhabited exclusively by the

individuals with the highest dispersal rates (even if they are not the fittest) because they have

arrived first and can only breed with each other [15], resulting in a geographic separation of

phenotypes [15, 16]; (3) admixture or hybridization may occur between individuals from

different introductions or sources [17].

It has recently been suggested that the genetic paradox of invasion may be rare; some

invasive populations do not suffer a reduction in genetic diversity during introduction, and

others do not face novel adaptive challenges in their introduced ranges [18]. Furthermore,

some invasive systems demonstrate a 'spurious' paradox due to inadequate estimation of

genetic diversity (i.e. too few markers), or to maintenance of genetic diversity only at

ecologically relevant traits, or to a reduction in genetic diversity resulting from natural

selection rather than from genetic bottlenecks [18]. These ideas can be tested with genome-

wide data [19, 20].

The Australian cane toad (*Rhinella marina*) is an invader that has exhibited rapid

evolution despite apparently low genetic diversity [21]. Cane toads were serially translocated

from South America to Martinique and Barbados, to Jamaica, to Puerto Rico, to Hawai'i

before 101 founders were introduced from the Hawai'ian island of Oahu to Queensland

(QLD), Australia in 1935 [22]. Toads have since spread westward through the Northern

Territory (NT) into Western Australia (WA; Fig 1). The native range and source area are

tropical environments with relatively high mean annual rainfall (3200 mm in French Guiana,

1200-2000 mm in Hawai'i) and high temperatures (26°C in French Guiana, 22-24°C in

Hawaii) [23]. The Australian range is heterogeneous in several environmental factors,

particularly aridity; on average, sites in QLD receive more annual rainfall than do those in the

NT and WA (2000-3000 mm in QLD, 400-1000 mm in NT and WA; Fig S1A), and have

lower annual mean temperatures (21-24°C in QLD, 24-27°C in NT and WA; Fig S1B) [24].

Heritable phenotypic differences in behavior [25], thermal performance [26], morphology

[27, 28], and immune function [29] have been documented in cane toads at different localities

across the Australian range. Nonetheless, surveys have reported low levels of genetic

diversity, based on microsatellites [30] and MHC [31], and only one mitochondrial haplotype

in the *ND3* gene of 31 individuals sequenced from Hawai'i and Australia [32].

Here, we assessed genetic diversity and population structure using single nucleotide

polymorphisms (SNPs) identified in RNA-Seq data from samples originating from French

Guiana (the "native range"), Hawai'i (the "source"), QLD (the "range core"), NT

("intermediate" areas), and WA (the "invasion front") to test whether the paradox is

supported in the cane toad invasion, and to investigate the evolutionary dynamics during

introduction. We predicted high divergence between native and invasive populations due to a

combination of genetic drift, selection, and spatial sorting, but little genetic differentiation

within invasion phases due to a putative lack of standing genetic diversity. Analyses to

identify loci with outlier F_{ST} values or associations with environmental variables allow us to

test for selection; because of the increase in aridity at intermediate areas and the invasion

front, we predicted that loci under selection would be in genes involved in thermal tolerance,

which may underlie the success that toads have had dispersing through these areas.

RNA-Seq data provide genome-wide information, but are limited to transcribed

regions of the genome [33]. RNA-Seq data may therefore contain variants that exhibit strong

signals for both demography and selection. Conversely, reduced representation sequencing

approaches such as restriction site-associated DNA sequencing (RADSeq) data provide

genome-wide information on a subset of all coding and non-coding sequences (only SNPs

near restriction enzyme cleavage sites are detected); because non-coding sequences make up

most of the genome, SNPs detected using this technique are more likely to represent neutral

variation [34]. To focus on how selection may shape genomic diversity in invasive toads, we

compared the results from our RNA-Seq data to those from a publicly available RADSeq

dataset [35] that we reanalyzed (the purpose of that study was not to analyze population

structure per se, so we modified the analysis for our questions). The RADSeq dataset was

limited to samples from QLD (the "range core" towards "intermediate" areas) and the border

of NT/WA ("intermediate" areas towards the "invasion front").

Materials and Methods

Sample collection, RNA extraction, and sequencing

In November of 2017, we collected samples from French Guiana (native range; Fig 1; Table

S1). In June of 2015, we collected samples from the Hawai'ian Islands (source; Fig 1; Table

S1). In April and May of 2013-2015, we collected samples along an invasion transect in

Australia (range core, intermediate areas, and the invasion front; Fig 1; Table S1). Upon

collection, we excised whole spleen tissue (sample sizes in Table S1A) and whole brain

tissue (sample sizes in Table S1B) from female toads immediately after euthanasia. Each

tissue sample was initially preserved in RNAlater buffer (QIAGEN, USA), initially kept at -

20℃ during fieldwork (less than one month), and then drained and transferred to a -80℃

freezer for long-term storage.

We carried out RNA extractions using the RNeasy Lipid Tissue Mini Kit (QIAGEN, USA)

following the manufacturer's instructions, with an additional genomic DNA removal step

using on-column RNase-free DNase treatment (QIAGEN, USA). We quantified the total

RNA extracted using a Qubit RNA HS assay on a Qubit 3.0 fluorometer (Life Technologies,

USA). Extracts were then stored at -80∘C until sequencing. Sequencing was conducted

commercially at Macrogen (Macrogen Inc., ROK). mRNA libraries were constructed using

the TruSeq mRNA v2 sample kit (Illumina Inc., USA), which included a 300bp selection

step. We sequenced 46 spleen samples and 72 brain samples in multiple batches (not pooled)

across 3 lanes of Illumina HiSeq 2500. Capture of mRNA was performed using the oligo dT

method, and size selection parameter choices were made according to the HiSeq2500

manufacturer's protocol. Each individual sample was ligated with a unique barcode. Overall,

this generated 709 million paired-end 2 x 125-bp reads from spleen data and 1.74 billion

paired-end 2 x 125-bp reads from brain data. Raw sequence reads are available as FASTQ

files in the NCBI short read archive (SRA) under the BioProject Accession PRJNA510261

(spleen data from French Guiana and Hawai'i), PRJNA395127 (spleen data from Australia),

and PRJNA479937 (all brain data).

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Data pre-processing and alignment

First, we examined per base raw sequence read quality (Phred scores) and GC content, and

checked for the presence of adapter sequences for each sample using FastQC v0.11.5 [36].

We then processed raw reads (FASTQ format) from each sample with Trimmomatic v0.35

[37], using the following parameters: ILLUMINACLIP:TruSeq3-PE.fa:2:30:10:4

SLIDINGWINDOW:5:30 AVGQUAL:30 MINLEN:36. This removed any adaptor

sequences, trimmed any set of five contiguous bases with an average Phred score below 30,

and removed any read with an average Phred score below 30 or sequence length below 36 bp.

As a reference, we used the annotated *R. marina* transcriptome [38], which was

constructed from brain, spleen, muscle, liver, ovary, testes, and tadpole tissue. We conducted

per sample alignments of reads (FASTQ files) to the reference using STAR v2.5.0a [39] in

basic two-pass mode with default parameters, a runRNGseed of 777, and specifying binary

alignment map (BAM) alignment outputs. As STAR-generated BAM files lack read groups

(identifiers for reads that specify the individual that they come from and the platform that was

used to sequence them), we added them to our BAM files using the

AddOrReplaceReadGroups tool in Picard Tools [40]. To avoid making incorrect variant calls,

we removed duplicate reads (reads that map to the same position of the same gene in the

same individual) using the MarkDuplicates tool in Picard Tools. To split reads containing an

N (region of the reference that is skipped in a read) into individual exon segments, we used

the SplitNCigarReads tool in the Genome Analysis Toolkit (GATK) v3.8.0 [41].

Variant calling and filtering

To call SNPs and insertion-deletions (indels), we used the HaplotypeCaller tool in GATK

[41] on our alignment (BAM) files. This tool works by identifying regions of the reference

that show evidence of variation (called 'active regions'). Variants required a minimum phred-

scaled confidence of 20 to be called (marked as passing the quality filters) and emitted

(reported in the output), using the stand_call_conf and stand_emit_conf options.

To ensure that SNPs were called as accurately as possible, we performed this in

'GVCF' mode (which generates one intermediate genomic variant call format, or 'gVCF' file

per individual sample). By using this mode, we avoided missing SNPs at loci that match the

reference in some but not all individuals. We then used the GenotypeGVCFs tool to merge

the gVCF files, re-calculate genotype likelihoods at each SNP locus across all individuals,

and re-genotype and re-annotate all SNP loci. The results were written to one merged VCF

file. Although we initially genotyped all spleens and brains together, we discovered during

downstream analyses that there was an effect of tissue type on population assignment – even

from the same individuals, and using only SNPs from transcripts expressed in both tissues,

spleen and brain samples were assigned to separate populations). Thus, we genotyped spleens

and brains separately, resulting in two merged VCF files, and subsequently kept them

separate for all downstream analyses. We retained both datasets because each provides a

unique benefit: the spleen dataset includes native range samples, but the brain dataset has

more extensive sampling of the invasive populations.

Rather than following a random distribution, some SNPs are clustered. Clustered

SNPs may not be independent, and are thus filtered. We used the VariantFiltration tool to

identify and filter 'clusters' (sets of 3 SNPs that appear within a window of 35 bases) in each

of our merged VCF files. We also used this tool to filter variants with QualByDepth (QD;

variant confidence divided by the unfiltered depth of non-reference samples) less than 2.0,

depth of coverage (DP) less than 20.0, and allele frequency less than 0.05. We then subset our

VCF files to include only the variants that passed the filters we set in the VariantFiltration

step using the SelectVariants tool. This resulted in 803,489 SNPs from spleen data and

818,536 SNPs from brain data. We used bcftools [42] to further filter the VCF files, only

retaining biallelic SNPs. We examined the results of filtering for minimum minor allele

frequency (min MAF) thresholds of 0.01 or 0.05, and several missing data tolerance (MDT)

thresholds (the maximum percentage of individuals in the dataset in which a genotype for a

locus can be absent without that locus being filtered out). Our population structure results

were consistent across min MAF and MDT thresholds, so we ultimately chose to filter our

data at min MAF = 0.05 and MDT = 0% (no missing data tolerated) because some

downstream analyses cannot handle missing data. These filtering steps reduced the number of

SNPs to 65,195 in spleen data and 35,842 in brain data (Table S1A-B).

Inference of population structure

We used PLINK [43] to convert our VCF files to the Browser Extensible Data (BED) format,

which is readable to fastStructure. We then used fastStructure [44] to infer population

structure using a variational Bayesian framework for calculating posterior distributions, and

to identify the number of genetic clusters in our dataset (K) using heuristic scores [44]. We

ran the structure.py ten times each (K= 1 to 10). We then took the resulting meanQ files from

fastStructure and plotted them using the pophelper package [45] in R [46].

In addition to using fastStructure for population assignment, we also performed a

Redundancy Analysis (RDA) using the vegan package [47] in R, which fits genetic and

environmental data to a multivariate linear regression, and then performs a principal

component analysis (PCA) on the fitted values. RDA visualizes both population structure and

the effects that environmental variables may have in shaping it. To do this, we downloaded

climatic data on French Guiana, Hawai'i, and Australia from the Bioclim database [48] using

the raster package [23]. Because these areas vary in aridity, we downloaded data on rainfall

and temperature; these data are averages of annual statistics over the period of 1970 to 2000.

Specifically, we downloaded: annual mean temperature, maximum temperature of the

warmest month, minimum temperature of the coldest month, annual precipitation,

precipitation of the wettest quarter, and precipitation of the driest quarter. We then used the

vcfR package [49] to convert our VCF files to the GENLIGHT format, which is readable to

the vegan package.

Identification of candidate loci under selection

Loci that are under natural selection may have abnormally high F_{ST} values, causing them to

be outliers among all other loci in the transcriptome. Bayescan v2.1 [50] detects loci with

outlier F_{ST}s using the multinomial-Dirichlet model. We used PGDSpider [51] to convert our

VCF files to the BAYESCAN format, and then ran Bayescan to detect loci with outlier

pairwise F_{ST} values between our three genetic clusters in spleen data (native range toads

versus source and core toads versus intermediate and frontal toads) and our two genetic

clusters in brain data (source and core toads versus intermediate and frontal toads).

Loci under natural selection also may have allele frequencies that are associated with

environmental variables. We used the climatic data that we downloaded from the Bioclim

database [48] using the raster package [23] on rainfall during the driest quarter and maximum

temperature in the warmest month. We then used the lfmm v2.0 package [52] in R to perform

a latent factor mixed model (LFMM) to test the association between the allele frequencies of

every locus and these two environmental variables. We applied a Benjamini-Hochberg

correction to all p-values from the LFMM.

Because these scans reveal thousands of loci putatively under selection, and because

outlier tests have high false positive rates [53], we took a more conservative approach by

cross-matching our list of F_{ST} outliers with each of our lists of environmentally-associated

loci. For individual investigation, we only looked at loci with both an outlier F_{ST} value and an

association with a putatively significant environmental variable.

Evaluation of genetic differentiation and diversity

To quantify levels of genetic differentiation and diversity, we computed basic statistics in the

hierfstat [54] and diveRsity [55] packages in R [46]. We first used PGDSpider v2.1.1.3 [51]

to convert our VCF files to the FSTAT and GENEPOP formats, which are readable to

hierfstat and diveRsity, respectively. We then used these packages to calculate measures of

genetic differentiation, including global F_{ST} and pairwise F_{ST} (by genetic cluster, invasion

phase, and collection site), and genetic diversity, such as expected heterozygosity (He) and

rarefied allelic richness (AR). We also computed Shannon's Information Index (SI) with the

dartR package [56]. After calculating these measures of diversity across all loci (N=65,195

from spleen data, N=35,842 from brain data), we calculated the same measures in loci

putatively under selection: those with outlier F_{ST} values (N=648 from spleen data, N=203

from brain data) and those associated with environmental variables (N=4179 from spleen

data, N=530 from brain data). In spleen data, this allowed us to investigate the hypothesis

that genetic diversity is maintained at ecologically relevant traits even if genome-wide

diversity is lost. In brain data, this allowed us to examine the effect of natural selection on

genetic diversity within Hawai'i and Australia. We used Kruskal-Wallis tests to assess the

significance of the differences in genetic diversity.

Annotation of SNPs

To visualize the types of genomic regions in which our SNPs lie, we used SnpEff [57].

SnpEff analyzes the information in a VCF file, annotates the SNPs, and estimates their

effects. We calculated the relative proportions of each type of SNP for both tissue types using

the full dataset of SNPs, and our two selection candidate datasets (containing F_{ST} outlier loci

associated with an environmental variable). We then used the stats v3.5.0 R package to

perform z-tests on the differences in proportions of SNP types between the full and selection

candidate datasets. We expected differences in the proportions of certain types of SNPs, such

as a lower proportion of synonymous variants in the selection candidates because

synonymous mutations do not result in a codon change (and thus are unlikely to change

protein function, i.e. phenotype).

Isolation by distance

To examine the effects of geographic distance on genetic distance across the Australian

range, we performed a Mantel test using ade4 v1.7-5 [58]. Because we were focused on

isolation by distance through range expansion after introduction (to Australia), samples from

Hawai'i and French Guiana were excluded; and because we sampled brain tissue in a greater

number of sites within Australia (as opposed to spleen tissue; Table S1A-B), we only

performed the Mantel test on brain data. For the geographic distance matrix, we used the dist

function in R [46] to calculate the Euclidean distances in geographic space between

collection sites based on their coordinates. For the genetic distance matrix, we used the

collection site-based pairwise F_{ST} values generated from hierfstat. We performed four Mantel

tests, using: (1) all loci, (2) loci with outlier F_{ST} values, (3) loci with outlier F_{ST} values and

with an environmental association (temperature or rainfall), (4) loci with outlier F_{ST} values

but without an environmental association. A significant result across all loci may suggest

that the prominent driver of genetic structure is genetic drift. However, clinal variation in

allele frequencies may also result from spatial sorting (particularly in loci underlying

dispersal ability), or from selection associated with environmental clines; distinguishing these

evolutionary processes is difficult. Thus, we performed the additional tests to attempt to

separate the effects of genetic drift and spatial sorting. Loci with outlier F_{ST} values are

candidates for natural selection and spatial sorting; thus, a significant result from only these

loci would less likely be caused by genetic drift. Outlier loci can be further separated into

those with an environmental association, which are likely candidates for natural selection

driven by environmental variables, and those without an environmental association, which

are possible candidates for spatial sorting. If spatial sorting is indeed influencing genetic

differentiation, then we may expect to see different patterns between these two groups of loci

(e.g. linearity only in the outlier loci without an environmental association).

Acquisition, processing, and analysis of RADSeq dataset

In addition to our own RNA-Seq dataset, we downloaded a publicly available RADSeq

dataset [35] to evaluate population structure in Australian cane toads. Unlike RNA-Seq data,

which consists of coding sequences only [34], RADSeq data include SNPs from across the

entire genome, but only around sites that are cleaved by a selected set of restriction enzymes

[33]. This method allows us to contrast the results of assessing loci potentially under 'tighter'

selection (e.g., codon bias) with those primarily constituting neutral variation (RADSeq).

Individuals in the RADSeq dataset were collected from QLD (range core towards

intermediate areas, N=179) and the border of NT/WA (intermediate areas towards the

invasion front, N=441; Fig 1). However, samples were lacking from a large part of the range

between those regions; thus, identifying geographic points of differentiation between genetic

clusters may be unlikely using this dataset. Furthermore, although there were many collection

sites within these regions, metadata with coordinates (or even location names) were

unavailable. We downloaded raw 100-bp reads from NCBI short read archive (SRA) under

the BioProject Accession PRJNA328156. We then converted SRA files to the FASTQ format

using the fastq-dump tool in the SRA toolkit.

We used Stacks v2.0 [59] for all RADSeq data processing. First, we used the

process_radtags program to remove low quality reads from the FASTQ files. Next, we used

the denovo_map pipeline to perform a *de novo* assembly for each sample, align matching

DNA regions across samples (called 'stacks'), and call SNPs using a maximum likelihood

framework. We allowed a maximum of five base mismatches between stacks within an

individual and three base mismatches between stacks between individuals. This resulted in

499,623 SNPs. We then filtered the results using the populations program, using one random

SNP per locus and four combinations of parameter choices (min MAF=0.05, MDT = 0.5; min

MAF = 0.05, MDT = 0.2; min MAF = 0.01, MDT = 0.5; min MAF = 0.01, MDT = 0.2).

Structure results were consistent across min MAF values, but not MDT values: two distinct

populations were only detected using the more stringent MDT choice. Because population

structure in this dataset is relatively low, and because it is appropriate to use more stringent

filtering choices when analyzing datasets with low population structure (Selechnik et al.,

unpublished; Chapter 2), we ultimately chose to filter this dataset with min MAF = 0.05 and

MDT = 0.2. Furthermore, using these filtering choices produced results that were most

consistent with those reported by Trumbo et al. (2016) and with those of our RNA-Seq

dataset. Filtering reduced the number of SNPs to 8,296. The results were written to a VCF

file and a STRUCTURE file (which could be read into fastStructure).

We ran fastStructure using the same methodology as we did for our RNA-Seq dataset.

In addition to the model-based fastStructure, we also performed a discriminant analysis of

principal components (DAPC) on this dataset using adegenet [60]. DAPC is a multivariate

approach that identifies the number of genetic clusters using K-means of principal

components and a Bayesian framework. We also converted the VCF to the FSTAT and

GENEPOP formats and computed the same basic statistics for the RADSeq dataset as we did

for our RNA-Seq datasets using hierfstat and diveRsity. Because collection site coordinate

metadata were unavailable for the RADSeq dataset, we did not perform a Mantel test using

these data.

Results

Population structure across the Hawai'ian-Australian invasion

We found three genetic clusters in our spleen data (Fig 2A): French Guiana (native range)

formed its own genetic cluster, Hawai'i (source) clustered with QLD (core), and NT

(intermediate) clustered with WA (invasion front). These patterns were mirrored by our brain

data, which did not include native range individuals but had more extensive sampling in

Hawai'i and Australia than did the spleen data. In our brain dataset, population differentiation

seemed to align with environmental barriers (Fig 2B-C): Hawai'i (source) clustered with

coastal QLD (core), whereas inland QLD/NT (intermediate) clustered with WA (invasion

front). Inference of substructure within these two genetic clusters revealed that source and

core toads may further differentiate into two separate groups (Fig 2D); however, intermediate

and frontal toads remain one genetic cluster (Fig 2E).

RDA was mostly consistent with fastStructure for both spleen (Fig 3A) and brain (Fig

3B) data, except for the differentiation of source toads and core toads as two separate genetic

clusters. It also suggested that: (1) Source toads may have diverged from toads of other

clusters due to selection from rainfall during the driest quarter. (2) Core toads may have

diverged from toads of other clusters due to selection from rainfall during the wettest quarter,

mean annual rainfall, and minimum temperature during the coolest month. (3) Intermediate

and frontal toads may have diverged from toads of other clusters due to selection from

maximum temperature during the hottest month and mean annual temperature.

Tests for selection

To identify loci putatively under selection, we searched for those with outlier F_{ST} values or

associations with environmental variables. Out of 65,195 SNPs from spleen data (N=46

individuals across native and invasive populations), 648 were identified as F_{ST} outliers, with a

mean global F_{ST} of 0.89. Mean pairwise F_{ST} values across outlier loci were 0.93 between the

native and source/core populations, 0.97 between the native and intermediate/front

populations, and 0.34 between the source/core and intermediate/front populations. From the

same dataset (of 65,195 SNPs), 3088 SNPs were associated with maximum temperature

during the warmest month and 1813 were associated with rainfall during the driest quarter.

Of these loci, 772 were associated with both environmental variables. We focused on SNPs

that were detected both as F_{ST} outliers and environmental correlates: there were 64 outlier

SNPs associated with maximum temperature and 47 outlier SNPs associated with rainfall in

the driest quarter (including 18 outliers associated with both environmental variables).

Most of the 64 outlier SNPs associated with maximum temperature during the hottest

month (Table S2A) were in transcripts with functions such as metabolism, transcription

regulation, and immune function. There were also two SNPs in *HSP4*, a gene encoding heat

shock protein (HSP) 4 (70 kDa), which is involved in thermal tolerance (on exposure to heat

stress) and protein folding and unfolding [61]. Most of the 47 outlier SNPs associated with

rainfall during the driest quarter (Table S3A) were in transcripts with functions such as cell

signaling and immune function.

Out of 35,842 SNPs from brain data (N=72 individuals across invasive populations

only), 203 were identified as F_{ST} outliers, with a mean global F_{ST} of 0.34. Mean pairwise F_{ST}

was 0.50 between the source/core and intermediate/front populations. From the same dataset

(of 35,842 SNPs), 345 SNPs were associated with maximum temperature during the warmest

month and 194 were associated with rainfall during the driest quarter. Nine of these loci were

associated with both environmental variables. Again, we focused on SNPs that were detected

both as F_{ST} outliers and environmental correlates: there were 38 outlier SNPs associated with

maximum temperature and 33 outlier SNPs associated with rainfall in the driest quarter; no

outliers were associated with both environmental variables.

Most of the 38 outlier SNPs associated with maximum temperature during the hottest

month (Table S2B) were in genes involved in cell signaling, mitochondrial processes

(metabolism), or gene expression [61]. Similarly to the spleen data, there was one SNP from

brain data in *HSP4*. Another SNP was in *ARNT2*, a gene encoding aryl hydrocarbon receptor

nuclear translocator 2, a transcription factor involved in the hypothalamic pituitary adrenal

(HPA) axis and visual and renal function [61]. Of the 33 outlier SNPs from brain data

correlated with rainfall during the driest quarter (Table S3B), 19 were in *MAGED2*, a gene

encoding melanoma-associated antigen D2, which is involved in renal sodium ion absorption

[61]. Four were in *STXBP1*, a gene encoding syntaxin-binding protein 1, which is involved in

platelet aggregation and vesicle fusion [61]. The rest were in genes generally involved in

gene expression and cell signaling [61].

Evaluation of genetic differentiation, diversity, and isolation by distance

Our spleen data suggest that divergence from the native population is strong in the

source/core population (pairwise F_{ST} = 0.29), and even stronger in the intermediate/frontal

population (pairwise F_{ST} = 0.32). Differentiation between the source/core and

intermediate/frontal populations is much lower (pairwise F_{ST} = 0.08). Overall divergence

among natives and invaders is moderately high (global F_{ST} = 0.18)

These trends are consistent with our brain data, in which genetic differentiation

between the source/core and intermediate/frontal populations is also low (pairwise F_{ST} =

0.07). In terms of substructure within the two invasive groups, differentiation between source

and core toads was even lower (pairwise F_{ST} = 0.04 in brain data). Generally, pairwise F_{ST} by

invasion phase (Table S4A) and collection site (Table S4B) from brain data revealed that

toads from areas that are more closely linked by invasion history are less strongly

differentiated. Because the brain data did not include native range samples, overall

divergence was estimated to be much lower than it was in the spleen data (global F_{ST} = 0.04).

Diversity statistics from spleen data suggest that the native range population is only

slightly more genetically diverse than the source/core population (native: He = 0.27, AR =

1.74, SI = 0.40; source/core: He = 0.26, AR = 1.74, SI = 0.40; p = 5E-4 for He). There is a

greater drop in diversity from the native range population to the intermediate and frontal

population (intermediate/front: He = 0.23, AR = 1.68, SI = 0.36; $p < $ 2E-16 for all measures).

However, we also estimated these measures by collection site (rather than by genetic cluster)

to detect changes in genetic diversity at a finer scale; these calculations reveal a more obvious

reduction in genetic diversity from the native population to all invasive collection sites (Table

1).

 Table 1. Expected heterozygosity (He), allelic richness (AR), and Shannon's Information Index (SI) estimated using SNPs from spleen RNA-Seq data on native and invasive populations of cane toads (*Rhinella marina*) at collection sites across French Guiana, Hawai'i, and Australia.

Collection site	All loci (He, AR, SI)	Outlier F_{ST} loci (He, AR, SI)	Environmentally associated loci (He, AR, SI)
French Guiana (native)	0.27, 1.66, 0.40	0.03, 1.25, 0.06	0.17, 1.53, 0.25
Kapolei Park (source)	0.23, 1.60, 0.34	0.07, 1.20, 0.10	0.28, 1.77, 0.41
Haiku Gardens (source)	0.23, 1.60, 0.34	0.07, 1.22, 0.11	0.28, 1.75, 0.41
Gordonvale (core)	0.24, 1.63, 0.36	0.07, 1.21, 0.10	0.30, 1.84, 0.45
Daintree (core)	0.24, 1.62, 0.35	0.08, 1.24, 0.12	0.29, 1.79, 0.44
Cape Crawford (intermediate)	0.22, 1.59, 0.32	0.02, 1.06, 0.03	0.22, 1.59, 0.32
Timber Creek (intermediate)	0.21, 1.58, 0.32	0.02, 1.05, 0.03	0.22, 1.60, 0.32
Halls Creek (front)	0.21, 1.57, 0.32	0.02, 1.05, 0.03	0.19, 1.58, 0.30
Durack River (front)	0.22, 1.58, 0.32	0.03, 1.11, 0.05	0.20, 1.60, 0.31

Despite a slight reduction in genetic diversity from the native to invasive populations genome-wide, our spleen data suggest that diversity was not lost during introduction in subsets of loci putatively under selection. Among loci with outlier F_{ST} values, there is no reduction in genetic diversity from the native population to the source/core population; rather, it is higher in the source/core population (native: He = 0.03, AR = 1.25, SI = 0.06; source/core: He = 0.09, AR = 1.37, SI = 0.14; $p < 2E\text{-}16$ for all measures). However, diversity in these loci is reduced in the intermediate/front population (intermediate/front: He = 0.02, AR = 1.14, SI = 0.04). These trends are also shown in environmentally-associated loci (native: He = 0.17, AR = 1.53, SI = 0.25; source/core: He = 0.37, AR = 1.99, SI = 0.55; intermediate/front: He = 0.23, AR = 1.81, SI = 0.35). Calculations of these measures by collection site further support these findings (Table 1).

Similarly to the spleen data, genome-wide diversity statistics from brain data suggest that the source/core population is more diverse than the intermediate/frontal population (source/core: He = 0.34, AR = 2.00, SI = 0.51; intermediate/front: He = 0.31, AR = 1.97, SI = 0.47; $p < 2E\text{-}16$ for comparisons of all measures). Calculations by invasion phase (Table S5) suggest a reduction in diversity from core to invasion front. This reduction is greater in loci

putatively under selection: in those with outlier F_{ST} values, the source/core population is

much more diverse than the intermediate/front population (source/core: He = 0.38, AR =

2.00, SI = 0.55; intermediate/front: He = 0.15, AR = 1.70, SI = 0.24; $p <$ 2E-16 for

comparisons of all measures). These trends are also shown in environmentally-associated loci

(source/core: He = 0.35, AR = 2.00, SI = 0.53; intermediate/front: He = 0.08, AR = 1.57, SI =

0.15; $p <$ 2E-16 for comparisons of all measures). Calculations of these measures by

collection site further support these findings (Table 2).

Table 2. Expected heterozygosity (He), allelic richness (AR), and Shannon's Information
512   Index (SI) estimated using SNPs from brain RNA-Seq data on native and invasive
513   populations of cane toads (*Rhinella marina*) at collection sites across Hawai'i and Australia.

Collection site	All loci (He, AR, SI)	Outlier F_{ST} loci (He, AR, SI)	Environmentally associated loci (He, AR, SI)
Paradise Park (source)	0.28, 1.57, 0.41	0.31, 1.76, 0.45	0.30, 1.84, 0.45
Mauna Lani (source)	0.27, 1.56, 0.40	0.24, 1.61, 0.35	0.29, 1.73, 0.42
Kapolei Park (source)	0.29, 1.57, 0.43	0.29, 1.83, 0.43	0.25, 1.73, 0.38
Haiku Gardens (source)	0.30, 1.59, 0.44	0.31, 1.75, 0.45	0.31, 1.83, 0.46
Gordonvale (core)	0.30, 1.61, 0.44	0.36, 1.82, 0.51	0.22, 1.62, 0.33
Cairns (core)	0.32, 1.60, 0.48	0.38, 1.95, 0.55	0.24, 1.84, 0.38
Daintree (core)	0.29, 1.60, 0.43	0.32, 1.90, 0.48	0.18, 1.57, 0.28
Croydon (intermediate)	0.26, 1.60, 0.37	0.18, 1.45, 0.26	0.10, 1.24, 0.14
Burketown (intermediate)	0.28, 1.57, 0.41	0.18, 1.44, 0.26	0.09, 1.31, 0.14
Cape Crawford (intermediate)	0.28, 1.58, 0.42	0.15, 1.42, 0.23	0.08, 1.32, 0.13
Mataranka (intermediate)	0.28, 1.57, 0.41	0.16, 1.46, 0.24	0.07, 1.25, 0.11
Timber Creek (intermediate)	0.27, 1.55, 0.40	0.05, 1.18, 0.09	0.06, 1.22, 0.10
Purnululu (front)	0.29, 1.55, 0.44	0.20, 1.48, 0.12	0.07, 1.37, 0.11
Halls Creek (front)	0.27, 1.56, 0.40	0.09, 1.25, 0.14	0.06, 1.21, 0.10
Durack River (front)	0.27, 1.56, 0.41	0.13, 1.37, 0.19	0.06, 1.24, 0.10

Geographic distance in collection sites and pairwise F_{ST} values by collection site (as

estimated with brain data) were significantly associated across all four of our Mantel tests

using: (1) all loci (p = 1E-3, r = 0.76; Fig 4A); (2) all loci with outlier F_{ST} values (p = 1E-3, r

= 0.75; Fig 4B); (3) only loci with outlier F_{ST} values and an environmental association (p =

1E-3, $r = 0.62$; Fig 4C); (4) only loci with outlier F_{ST} values but without an environmental

association ($p = 1E-3$, $r = 0.77$; Fig 4D).

SNP Annotations

Annotation of SNPs revealed that most of our loci were either missense, synonymous, or

found in the 3' untranslated region (UTR). A small percentage were found in the 5' UTR, and

the remaining loci involved the loss or gain of stop or start codons (Figs S2-S3). In the spleen

data, the temperature-associated outlier F_{ST} subset had a significantly higher proportion of

synonymous variants ($p = 8E-3$) than the full set of SNPs. The outlier loci within *HSP4* were

both synonymous variants. However, there were no significant differences in proportions

between the rainfall-associated outlier F_{ST} subset and the full set of SNPs.

In the brain data, there were no significant differences in proportions of SNP variant

types between the temperature-associated outlier F_{ST} subset and the full set. The locus within

HSP4 was a 3' UTR variant, and the locus within *ARNT2* was synonymous. The rainfall-

associated outlier F_{ST} subset had a significantly higher proportion of 3' UTR variants ($p =$

$5E-3$) and lower proportion of missense variants ($p = 0.04$) than did the full set. Eighteen of

the nineteen loci within *MAGED2* were 3' UTR variants (the last was missense), and two of

the four loci within *STXBP1* were missense (the other two were synonymous).

Population structure in Australian toads from the RADSeq experiment

Analysis of the RADSeq dataset through fastStructure and DAPC also showed two genetic

clusters within Australia (Fig 5A-B): QLD (core towards intermediate areas) in the first

group, and the border of NT/WA (intermediate areas towards the invasion front) in the

second group. There was little differentiation between these two groups (pairwise $F_{ST} = 0.09$)

or overall (global $F_{ST} = 0.04$). We found no evidence of significant substructure within either

544 of the two genetic clusters. The two groups were equal in allelic richness (AR = 2.00 for both

545 groups), but expected heterozygosity was higher in core toads than in intermediate/frontal

546 toads (core: He = 0.50; intermediate/front: He = 0.33), which was consistent with our RNA-

547 Seq results.

549 **Discussion**

550 We predicted that genetic diversity would be significantly lower in invasive populations than

551 in native populations due to genetic bottlenecks taking place during the serial introductions of

552 cane toads across Hawai'i and Australia. We also predicted strong divergence between native

553 range and invasive populations due to a combination of genetic drift, selection and spatial

554 sorting. In terms of population structure, we predicted low genetic structure within invasive

555 populations in Hawai'i and Australia resulting from putatively low levels of standing genetic

556 variation. Our analyses show evidence of a reduction in genetic diversity from native to

557 invasive populations. We found that the toads form three genetic clusters: (1) native range

558 toads form their own cluster, (2) Hawai'ian source toads cluster with eastern Australian range

559 core toads, and (3) all toads from further west along the recent invasion transect cluster

560 together (intermediate through to the invasion front). Differentiation is much higher between

561 the native population and the invasive populations than between the invasive populations.

562 The loci with the highest differentiation between native and invasive populations (outlier

563 pairwise F_{ST} values > 0.9 between native and invasive) have relatively low diversity across

564 all populations; this may be because these loci are homozygous in most individuals with no

565 shared alleles between populations (fixed). Associations between our genetic structure

566 patterns and allele frequencies with environmental variables suggest that natural selection is

567 occurring in the western part of the Australian range. However, the significant result of our

568 Mantel test across all loci also reveals gradual genetic drift across invasion phases; and the

significant result and high correlation coefficient of our Mantel test of F_{ST} outliers without

environmental associations may be due to spatial sorting. It seems likely that all these forces

are responsible for the observed population structure within invasive populations.

Our data suggest that cane toads may fit the model of the genetic paradox; we found

evidence of strong divergence between native and invasive cane toad populations, and of a

reduction in overall genetic diversity. Previous studies on mtDNA [32] and microsatellite

[62] data have also suggested losses in genetic diversity from the native range to Hawai'i and

Australia. Despite the reduction in genetic diversity from native to invasive populations

across the full set of loci in the spleen data, we did not find this reduction in loci putatively

under selection (loci with outlier F_{ST} values or associations with environmental variables) in

the source/core population. This may provide support for the prediction that some invasions

do not represent a genetic paradox despite overall loss of genetic diversity and the presence

of novel adaptive challenges because genetic diversity at ecologically relevant traits is

maintained by balancing selection [18]. However, diversity of this subset of loci is not just

maintained in the source/core population; it is higher than in the native population. This may

reflect admixture between toads from multiple introductions from different areas of the native

range to Puerto Rico, where Hawai'ian toads were sourced from; the original South American

sources for the Puerto Rican introductions included toads from both French Guiana and

Guyana [22]. Alternatively, this may represent a genuine paradox; *de novo* mutation rates

may have been high enough to restore or even enhance adaptive potential in the source/core

population [18]; although these toads have been in Hawai'i and Australia for less than 100

years, they first diverged from native South American toads in the early 1800s when toads

were introduced to the islands that would eventually (directly or indirectly) provide the

source populations for the Puerto Rican introduction (i.e. Martinique and Barbados). From

the source/core population to the intermediate frontal population, however, diversity of loci

putatively under selection is highly reduced; we believe this is because of directional

selection, as outlined below.

Our two invasive genetic clusters (source/core and intermediate/front) diverge at the

transition from coastal to inland areas within Australia, in an area where temperatures

become higher and rainfall becomes scarcer, particularly during the dry season [24]. This

correlation suggests that these environmental variables may drive population structure and

that toads may be adapting to local conditions. Furthermore, RDA implicates selection from

temperature and rainfall in driving divergence between invasive populations. Heterogeneous

climates across introduced or expanding ranges have previously been linked to genetic

differentiation, possibly due to climatically-imposed selection [10, 63].

Diversity estimates from our brain data suggest a reduction in diversity from the

source/core population to the intermediate/frontal population in the full set of loci, but a

much greater reduction in diversity in the subset of loci putatively under selection. The

source (Hawai'i) and core (QLD) areas represent earlier phases of the invasion and

experience similar environmental conditions to those of the native range, whereas

intermediate (NT) and invasion front (WA) areas are more arid. Thus, the greater depletion in

diversity at loci putatively under selection across the Australian invasion may be due to the

effects of directional selection from the harsher climate at more recently colonized areas of

the Australian range [18, 64].

Among our loci under selection associated with temperature are a few within a gene

encoding protein HSP 4 (70 kDa). HSPs protect protein folding during increased

temperatures and provide cells with enhanced thermal tolerance [65], and expression of HSP

genes has been shown to underlie adaptive responses to environmental stress [66]. Changes in

HSP levels in response to thermal environment manipulation vary between native and

invasive cane toads, as well as between populations of toads within Australia [67]. In our

study, the SNPs within *HSP4* included two synonymous variants (spleen data) and a 3' UTR

variant (brain data); 3' UTR variants are known to alter levels of protein functioning through

differential expression of genes [68]. In the case of heat shock proteins, adaptation may not

necessarily mean a change in function, but rather a change in the expression levels of proteins

involved in the function, which could be caused by a 3' UTR variant. Another notable protein

was aryl hydrocarbon receptor nuclear translocator 2 (from the *ARNT2* gene), a transcription

factor that specifically acts on genes involved in the HPA axis and visual and renal function.

The HPA axis is stimulated by high temperatures [69], and the secretion of corticosterone can

be modulated adaptively in response to thermal change [70]. Elevation of corticosterone

secretion in toads increases the rate of desiccation, thus making it maladaptive in arid

environments [71]. Corticosterone levels in Brazilian and Australian toads in response to

manipulation of thermal environment display similar patterns to those of HSPs [67].

Selection on loci potentially affecting corticosterone secretion by affecting the HPA is thus in

line with these results. However, the SNP in *ARNT2* is synonymous; thus, this SNP might not

change protein function.

More than half (19 of 33) of the loci under selection associated with rainfall (in brain

data) were found in *MAGED2* (encodes melanoma-associated antigen D2). This protein

regulates the expression and localization of two co-transporters that facilitate renal sodium

ion absorption, which prevents excess loss of water and solutes via reabsorption through the

kidneys (to be returned to the extracellular fluid and circulatory system) [72]. Similarly to the

SNP within *HSP4*, 18 of 19 SNPs within *MAGED2* were 3' UTR variants, which can affect

gene expression [68]. The high number of SNPs found in the 3' UTR of *MAGED2* is likely

responsible for the difference in proportions of variant types between the full dataset and the

rainfall-associated outlier dataset. Four more SNPs were found in *STXBP1* (encodes syntaxin-

binding protein 1, which is involved in vesicle fusion and blood clotting). Dehydration is

known to increase blood clotting rate [73], and excessive blood clotting can be harmful to the

heart, brain, and limbs [74]. Two of the four SNPs in *STXBP1* were missense variants,

possibly producing change in protein function. Evolved changes in the rates of renal sodium

absorption and blood clotting may allow toads from intermediate and invasion front areas to

survive in drier environments. The remaining SNPs associated with rainfall or temperature

were in genes generally involved in gene expression or signal transduction; these genes may

facilitate expression of (or signaling to) the genes we have discussed, or may have

independent functions in cellular maintenance.

Invasive species have been shown to rapidly adapt phenotypically to heterogeneous

climatic conditions, allowing them to expand their ranges [75]. Some invaders tolerate higher

temperatures [76] and water loss [77] better than do related indigenous taxa [78]. There is

evidence of this in cane toads: under dehydrating conditions, wild-caught individuals from

Australia exhibit better locomotor performance than do conspecifics from Hawai'i and the

native range [79], and within Australia, toads from semi-arid areas (i.e. NT) exhibit better

locomotor performance than do conspecifics from wetter areas (i.e. coastal QLD) [79, 80].

This trait is heritable; captive-bred toads with parental origins from hotter areas (northwestern

Australia) outperform those with parental origins from cooler areas (northeastern Australia)

at high (but not low) temperatures in a common-garden setting [26]. Coupled with the

previously shown heritable phenotypic patterns, our results on genetic structure suggest that

intermediate and frontal toads in Australia may be evolving enhanced thermal tolerance,

thereby facilitating their continued westward range expansion [81]. However, experiments

pairing genetic and phenotypic data from the same individuals are required to test this

hypothesis.

Although our patterns of genetic structure (cluster separation along a distinct climatic

gradient) and the identities of loci putatively under selection suggest that invasive toads are

responding to localized natural selection, our data also support the presence of genetic drift.

The significant result of our Mantel test using all loci suggests that isolation by distance is

occurring in the Australian toad population. Slight reductions in diversity (SI, AR, and He)

along the invasion trajectory found here are expected and have been seen in other invasions

[14]. However, the results of our additional Mantel tests suggest that another evolutionary

force may also be at work: spatial sorting is the separation of individuals within a range-

expanding population along phases of their expansion based on their dispersive capabilities

[15], and may lead to clinal variation in dispersal-related allele frequencies along the range.

Genetic differentiation in our loci with outlier F_{ST} values (which are candidates for natural

selection and spatial sorting) is significantly associated with geographic distance.

Furthermore, separating F_{ST} outlier loci with an environmental association (likely candidates

for natural selection) from those without (possible candidates for spatial sorting) yields

different patterns. Although the Mantel tests on both datasets are significant, the test on the

environmentally associated outliers unsurprisingly shows two discrete groups of pairwise F_{ST}

values – low pairwise F_{ST} values at low geographic distances and high pairwise F_{ST} values at

high geographic distances (Fig 4C). Interestingly, the test on outliers without an

environmental association shows a clinal increase in pairwise F_{ST} values with geographic

distance (like those on the full dataset and the full outlier dataset), which may suggest that

some of these loci are influenced by spatial sorting (Fig 4D).

Our results from the RADSeq dataset [35] were similar to those from our RNA-Seq

dataset; both showed two genetic clusters within Australia. However, unlike the RNA-Seq

results (which show a divide occurring in inland QLD, where temperature increases and

rainfall decreases), the RADSeq results cluster all QLD toads in the first group, and NT/WA

toads in the second group. In the RADSeq dataset, the spatial location of the transition

between the two groups is difficult to pinpoint due to the absence of intermediate sampling

sites. The difference in group divisions between the two datasets may be because selection

from environmental variables causes differentiation between toads from coastal and inland

QLD (as seen in the RNA-Seq data), whereas demographic processes such as gene flow may

lessen differentiation between them (as seen in the RADSeq data). Alternatively, toads from

coastal and inland QLD may cluster together in the RADSeq data because the sampling sites

in that study were south of those in our study, and the environmental conditions of coastal

and inland sites are more similar in that area [48]. Estimates of diversity from the RADSeq

dataset were slightly higher than those from the RNA-Seq dataset; this may be because

neutral variation is expected to be higher than adaptive variation due to the effects of

directional and purifying selection [64].

In conclusion, it appears that serial introductions have led to a reduction in genetic

diversity of invasive cane toads. Nonetheless, cane toads do indeed face novel adaptive

challenges in their introduced ranges, and are able to respond genetically to natural selection,

particularly in the harsh environmental conditions of northwestern Australia. The ability of

toads to adapt to these conditions may reflect the maintenance of diversity at ecologically

relevant loci, or it may reflect sufficiently high mutation rates to bolster adaptive potential;

thus, it remains unclear whether the toad invasion represents a true genetic paradox. Natural

selection, genetic drift, and spatial sorting are the most plausible mechanisms by which

genetic differentiation may occur, but require further investigation to tease apart.

Determining the association between phenotypic traits and the occurrence of certain SNPs

(such as the candidates that we have identified) at the individual level may be especially

informative. Additionally, genetic changes may not be the only mechanisms by which

adaptation occurs; studies on plasticity and epigenetic changes may also be useful for

uncovering the mysteries of rapid evolution.

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Figures:

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Fig 1. Map of two invaded territories (Australia and Hawai'i) and the native range (French

Guiana) of the cane toad (*Rhinella marina*). The shaded region represents the toad's current

Australian range, which is continuing to expand. Black diamonds in French Guiana, Hawai'i,

and Australia indicate collection sites for our RNA-Seq experiment. White circles in

Australia indicate collection sites for the RADSeq experiment from which we downloaded

data.

Fig 2. (A) Population structure of cane toads from French Guiana, Hawai'i, and Australia, as

inferred by SNPs from spleen RNA-Seq data. (B) Rainfall during the driest quarter in each

location across collection sites for invasive cane toads in French Guiana, Hawaii, and

Australia. A similar environmental barrier occurs in temperature (Fig S1). (C) Population

structure of invasive cane toads from Hawai'i and Australia, as inferred by SNPs from brain

RNA-Seq data. (D) Substructure within toads from Hawai'i (HI) and Queensland (QLD) as

inferred by SNPs from brain RNA-Seq data. (E) Substructure within toads from Northern

Territory (NT) and Western Australia (WA) as inferred by SNPs from brain RNA-Seq data.

Fig 3. (A) Relationship between environmental variables and population structure of cane

toads from French Guiana, Hawai'i, and Australia, as inferred by Redundancy Analysis

(RDA) on SNPs from spleen RNA-Seq data. Environmental data include: annual mean

temperature (Mean Temp), maximum temperature of the warmest month (Max Temp),

minimum temperature of the coldest month (Min Temp), annual precipitation (Mean Rain),

precipitation of the wettest quarter (Wet Rain), and precipitation of the driest quarter (Dry

Rain). (B) Relationship between environmental variables and population structure of invasive

cane toads from Hawai'i and Australia, as inferred by RDA on SNPs from brain RNA-Seq

data.

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Fig 4. Relationship between geographic distance (km) and genetic distance (pairwise F_{ST})

between cane toads collected across their invasive Australian range (Fig 1) as inferred by

brain RNA-Seq data. (A) represents the full dataset of loci (p = 1E-3, r = 0.76). (B) represents

all F_{ST} outliers (p = 1E-3, r = 0.75). (C) represents only F_{ST} outliers with an association with

either rainfall or temperature (p = 1E-3, r = 0.62). (D) represents only F_{ST} outliers without an

association with either rainfall or temperature (p = 1E-3, r = 0.77).

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1007 Fig 5. (A) Population structure of cane toads across their invasive range in Australia, as

1008 inferred from running fastStructure on SNPs from RADSeq data. Because coordinate

1009 metadata for collection sites were unavailable, the samples could not be arranged from east to

1010 west along the Australian range. (B) Population structure of cane toads across their invasive

1011 range in Australia, as inferred from DAPC on SNPs from RADSeq data.