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4 5	Supplementary Materials for
6 7	Concerted evolution reveals co-adapted amino acid substitutions in frogs that prey on toxic toads.
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31 Materials and Methods

32 <u>Sample collection and data sources</u>

We sampled tissues from 16 anuran species. Five Leptodactvlus species, as well as 33 34 Engystomops pustulosus, Lithodytes lineatus, and Rhinella marina, were collected from different 35 geographic locations in Colombia (Table S1) and stored in RNAlater (Invitrogen) at -80°C until 36 used. Atissue sample of the toad, Atelopus zeteki, was donated by the Smithsonian's National 37 Zoo and came from a necropsied captive animal. The outgroup species, *Megophrys nasuta*, 38 Kaloula pulchra, Rana sphenocephala, Rana catesbeiana, Dendrobates auratus, 39 Melanophryniscus stelzneri, and Duttaphrynus melanostictus were obtained from the pet trade 40 under IACUC Protocol No. 2057-16. Live animals were euthanized under the supervision of a 41 research veterinarian at Princeton University. To capture all three paralogs of ATP1A, we 42 collected tissue samples from brain, skeletal muscle, and stomach – each of which highly 43 expresses at least one of the three paralogs (15). 44 45 RNA-seq based gene discovery of ATP1A paralogs 46 Full-length coding sequences of ATP1A1, ATP1A2 and ATP1A3 were reconstructed for 47 several species using RNA-seq based gene discovery. Total RNA was extracted from multiple 48 tissues of 16 anuran species (Table S2) using TRIzol Reagents (Ambion, Life technologies) 49 following the manufacturer's protocol. RNA-seq libraries were prepared with TruSeq RNA 50 Library Prep Kit v2 (Illumina) and sequenced on Illumina HiSeq2500 (Genomics Core Facility, 51 Princeton, NJ, USA) with either PE 75bp or SE 140bp (Table S2). Reads were trimmed and de 52 novo assembled with Trinity v2.2.0 (40). ATP1A1 of Xenopus laevis (GenBank 53 NM 001090595) was initially used to BLAST against the assembled transcripts of L. latrans to

54	recover ATP1A1S and ATP1A1R, which were later used as queries to reconstruct ATP1A1
55	genes from other species. ATP1A paralogs for the rest of the species used in this study were
56	mined from publicly available data (Table S2) following the same pipeline.
57	
58	Targeted sequencing of protein-coding regions of ATP1A1 paralogs
59	Total RNA was extracted from L. fuscus, L. insularum, and L. colombiensis as described
60	above and reverse-transcribed to single-strand cDNA using SuperScript III Reverse
61	Transcriptase (Invitrogen). ATP1A1 was amplified using Phusion High-Fidelity DNA
62	polymerase (Invitrogen) using forward primer: 5'-ATAAGTATGAGCCCGCAGCC-3' and
63	reverse primer: 5'-CCAGGGCTGCGTCTGATTATG-3'. PCR products were cleaned with
64	QIAquick PCR Purification Kit (Qiagen) and A-tailed with Taq Polymerase (NEB) before
65	cloning into a pTOPO-TA vector (Invitrogen). The presence of the insert in the plasmid was
66	confirmed by colony-PCR. Illumina-ready sequencing libraries of isolated plasmids were
67	prepared with Tn5 transposase, charged with Illumina-ready indexed barcodes (22), and
68	sequenced on Illumina MiSeq (Genomics Core Facility, Princeton, NJ, USA). De novo assembly
69	of the cloned PCR products was performed with Velvet v1.2.10 (41) and Oases v0.2.8 (42).
70	ATP1A1 paralogs were reconstructed by aligning with previously obtained ATP1A1 sequences
71	of L. latrans and L. pentadactylus.
72	
73	De novo genome assembly of Leptodactylus fuscus

High-molecular-weight genomic DNA was isolated from a single *Leptodactylus fuscus*individual (Table S1, JSM 205) following standard protocols. This was used to prepare a 10Xgenomics Chromium library and sequenced on Illumina HiSeq X sequencer. Barcodes were

77	removed using the Long Ranger basic v2.2.2 (https://support.10xgenomics.com/genome-
78	exome/software/downloads/latest). Trimmed reads were used for k-mer estimation in Jellyfish
79	(33, v2.2.7). The k-mer (k=21) frequency distribution was processed in GenomeScope (43) to
80	estimate the genome size, heterozygosity, and percentage of repeat content. The linked-reads
81	were assembled using the Supernova assembler (44) with "-accept-extreme-coverage" specified
82	because the coverage was lower than recommended. The assembled genome is 2.42 Gb (16,530
83	scaffolds ≥ 10 kb, scaffold N50 = 363 kb, Table S3) and was outputted in the psuedohap2
84	format (accession No. TBD). The completeness of the genome assembly was assessed using
85	Benchmarking Universal Single-Copy Orthologs (BUSCOs, v4.0.5 (45)), and 72.6% of the
86	BUSCO Tetrapoda gene annotations (version odb10) were identified (Table S3).
87	
88	Targeted long-read sequencing of intronic sequences of ATP1A1
89	Intron annotations were determined by blasting (blast-2.26) the protein-coding
90	sequences of ATP1A1 S and ATP1A1 R against the L. fuscus genome assembly (Figure S4). For
91	the other four <i>Leptodactylus</i> species (<i>L. pentadactylus</i> , <i>L. latrans</i> , <i>L. insularum</i> , and <i>L.</i>
91 92	
	the other four Leptodactylus species (L. pentadactylus, L. latrans, L. insularum, and L.
92	the other four <i>Leptodactylus</i> species (<i>L. pentadactylus</i> , <i>L. latrans</i> , <i>L. insularum</i> , and <i>L. colombiensis</i>) and two outgroup species (<i>Engystomops pustulosus</i> and <i>Lithodytes lineatus</i>),
92 93	the other four <i>Leptodactylus</i> species (<i>L. pentadactylus</i> , <i>L. latrans</i> , <i>L. insularum</i> , and <i>L. colombiensis</i>) and two outgroup species (<i>Engystomops pustulosus</i> and <i>Lithodytes lineatus</i>), introns were obtained via targeted long-read sequencing using Oxford Nanopore MinION.
92 93 94	the other four <i>Leptodactylus</i> species (<i>L. pentadactylus</i> , <i>L. latrans</i> , <i>L. insularum</i> , and <i>L. colombiensis</i>) and two outgroup species (<i>Engystomops pustulosus</i> and <i>Lithodytes lineatus</i>), introns were obtained via targeted long-read sequencing using Oxford Nanopore MinION. Genomic DNA was extracted with Agencourt DNAdvance Kit (Beckman Coulter, France) and
92 93 94 95	the other four <i>Leptodactylus</i> species (<i>L. pentadactylus</i> , <i>L. latrans</i> , <i>L. insularum</i> , and <i>L. colombiensis</i>) and two outgroup species (<i>Engystomops pustulosus</i> and <i>Lithodytes lineatus</i>), introns were obtained via targeted long-read sequencing using Oxford Nanopore MinION. Genomic DNA was extracted with Agencourt DNAdvance Kit (Beckman Coulter, France) and ATP1A1 was amplified using LongAmp <i>Taq</i> PCR kit (NEB) using customized species-specific
92 93 94 95 96	the other four <i>Leptodactylus</i> species (<i>L. pentadactylus</i> , <i>L. latrans</i> , <i>L. insularum</i> , and <i>L. colombiensis</i>) and two outgroup species (<i>Engystomops pustulosus</i> and <i>Lithodytes lineatus</i>), introns were obtained via targeted long-read sequencing using Oxford Nanopore MinION. Genomic DNA was extracted with Agencourt DNAdvance Kit (Beckman Coulter, France) and ATP1A1 was amplified using LongAmp <i>Taq</i> PCR kit (NEB) using customized species-specific barcoded primers (Table S5). PCR products were gel confirmed and isolated using QIAquick

100	read length distribution matched that shown on the gel image of the amplicons. Base-calling
101	from raw trace data was performed using Albacore v2.3.4 (Oxford Nanopore Technologies) and
102	sequences were demultiplexed using LAST v980 (46). Reads that mapped to more than one
103	barcode were discarded. Reads were assigned to each species based on barcodes using seqtk
104	(47). Only reads of the expected length \pm 200 nt were used for downstream analyses. For
105	Leptodactylus species with two ATP1A1 paralogs, reads were further split by perfectly matching
106	the 111-122 region of the two copies, which exhibit 22-25% difference in nucleotide sequences.
107	Assembly was carried out using Canu v1.8 (48) and 1000 reads (i.e. 1000x coverage) were
108	randomly selected for better performance. Reconstructed sequences were identical when
109	different sets of 1000 reads were used. Filtered reads were mapped back to the reconstructed
110	reference with minimap2 (49) and polished with racon v1.3.3 (50). Short-read sequencing data
111	were generated using Tn5 transposase-based Illumina sequencing (as described above) to further
112	correct and polish the sequences. Final sequences were aligned using MUSCLE (51)
113	implemented in SeaView (52). The boundaries between introns and exons were manually
114	adjusted to start with GT and end with AG. Sequences are available at Genbank MT422192 -
115	422203 (Table S2).

117 Estimation of genealogical relationships

A time-tree of anuran species in Figure 1A was derived from Feng *et al.* (*38*). Amino acid substitutions at sites that are implicated in cardenolide sensitivity (*22*) are shown. The nucleotide tree and protein tree (Figure. 1B, C) of *Leptodactylus* and outgroup species were built with the exons and introns and protein sequences (Table S2), respectively. Phylogenies for ATP1A1 were estimated using a maximum likelihood method implemented in SeaView and visualized inEvolView (53)

124 We constructed a multi-locus species tree for three *Leptodactylus* species (*L. fuscus*, *L.* 125 *pentadactylus*, *L. latrans*) and two outgroups (*Engystomops pustulosus* and *Lithodytes lineatus*) 126 with high-confidence split time estimates specifically for use in the analyses described in 127 sections "Theoretical single-site model for the probability of maintaining an adapted 128 substitution" and "Simulations of ATP1A1 gene family evolution". Protein-coding genes were 129 predicted from *de novo* transcriptome assemblies for each species using Augustus (v3.2.2) (54) 130 and queried against the Tetrapoda ortholog database (odb10, https://www.orthodb.org) using 131 BLAST (tblastn). A concatenated multi-alignment of mRNA sequences was created for 813 132 orthologous proteins longer than 100 amino acids that were shared among all five species. The 133 best-fit nucleotide substitution model for each protein (i.e. each initial partition) was first 134 determined using the "ModelFinder" function of IQ-TREE 2 (v.2.0.4; 55) (command line: 135 iqtree2 -s concat 813 mafft.fasta -spp partition.txt -m MFP -nt AUTO -safe --prefix 136 concat 813 partition MFP). Proteins with the same inferred mutation model were subsequently merged into the same partition (using "-m TESTMERGE") prior to phylogenetic inference 137 138 (command line: iqtree2 -s concat 813 mafft.fasta -spp partition MFP best scheme.nex -m 139 TESTMERGE -nt AUTO --prefix concat 813 partition MFP merged). 140 141 Maximum likelihood analysis of site-wise support for alternative tree topologies 142 We used site-wise likelihoods to evaluate the relative level of statistical support for two

143 alternative tree topologies relating to the origin of R/S ATP1A1 paralogs: Model 1 ("Non-

144 Concerted") posits a single ancient origin of a R/S duplication with no concerted evolution:

145	$((Lfus_S,(Lpen_S,(Lins_S,Llat_S,Lcol_S))),(Lfus_R,(Lpen_R,(Lins_R,Llat_R,Lcol_R)))).$
146	Model 2 ("Concerted") is the expected topology under concerted evolution: ((Lfus_S, Lfus_R),
147	((Lpen_S, Lpen_R), ((Lins_S, Lins_R), (Llat_S, Llat_R), (Lcol_S, Lcol_R)))). We note that the
148	speciation events are assumed to follow the order inferred in the section "Estimation of
149	genealogical relationships". For each nucleotide states (e.g. AAAATTTTTT, in the order of
150	Lfus_S, Lfus_R, Lpen_S, LpenR, Llat_S, LlatR, LcolS, LcolR, LinsS, LinsR), likelihoods for the
151	two topologies were calculated using <i>PAML 4.8 baseml</i> (56). We consider $ \Delta \log - likelihood \ge 2$,
152	as significant support for one topology over the other. 4-, 2-, 0-fold degenerate sites were
153	classified using MEGA 7 (57) and all variants at these sites were categorized as either
154	synonymous or nonsynonymous. We used Fisher's Exact Test to test the hypothesis that the ratio
155	of synonymous and nonsynonymous variants is independent of support for one of the topologies
156	over the other (Table S4).
157	We further tested whether synonymous variants supporting alternative tree topologies (as
158	outlined above) are equally distant from R/S distinguishing substitutions: We computed the
159	distance of each variant from the nearest R/S distinguishing substitution, and compared the
160	median distance of synonymous variants with $ \Delta log-likelihood \ge 2$ support for the "Non-
161	Concerted" genealogy to a random sample of synonymous variants supporting multiple origins.
162	
163	Theoretical single-site model for the probability of maintaining an adapted substitution
164	Below, we describe the model and parameters used to compute the probability of
165	maintaining a diverged substitution in two gene copies.
166	Model. We consider a single biallelic amino acid site in tandemly duplicated genes,
167	evolving for t years. The two gene copies are initially fixed for the two distinct alleles. The site

168 experiences mutation at rate 2μ (or 4μ for both copies) where μ is the per-nucleotide mutation 169 rate, assuming for simplicity that all sites are biallelic, all mutations in the first two positions of 170 the codon are nonsynonymous and all mutations at the third position are synonymous. The site 171 also experiences non-allelic gene conversion at rate 4c (for both copies) and is under purifying 172 selection with fitness cost s > 0, such that having two distinct alleles at the two copies confers a 173 fitness of 1 and having the same allele confers to fitness (1 - s).

174 *De novo* mutations (through point mutation or gene conversion) from the initial distinct-175 allele haplotype to a same-allele haplotype can occur in all haplotypes in the population. In a

176 diploid population of size N, de novo same-allele haplotypes arise at rate

177
$$P(de \ novo \ same - allele \ haplotype) = \ 2N \cdot 4 \cdot (\mu + c).$$

178 The probability of fixation is bounded by the neutral case of s = 0, such that

179
$$P(same - allele haplotype fixes) < \frac{1}{2N}$$

180

If

$$181 8N \cdot (\mu + c) \ll 1$$

182 and

$$\frac{1}{2N} \ll 1,$$

184 then the overall per-year rate of fixation for deleterious haplotypes, α , can be approximated by 185 the product of these two,

186
$$\alpha = P(de \text{ novo same} - allele haplotype) \cdot P(same - allele haplotype fixes) =$$

187
$$8N(c+\mu) \cdot \frac{e^{s}-1}{e^{2Ns}-1}$$

188 where we replaced *P*(*deleterious haplotype fixes*) with Kimura's fixation probability for a

189 deleterious allele (58, 59). Assuming a vanishingly small probability of back-mutations—

namely, that no fixation of a same-allele haplotype is followed by another fixation reversing the
haplotype back to the distinct alleles—the probability of maintaining the distinct-alleles

192 haplotype for *t* years is:

193
$$P(\text{maintenance of distinct alleles}) = (1 - \alpha)^t = \left(1 - 8N(c + \mu)\frac{e^s - 1}{e^{2Ns} - 1}\right)^t.$$
(1)

194 We note that if $s \ll 1$ then

195

$$e^s \approx 1 + s$$

and therefore

197
$$P(maintenance \ of \ distinct \ alleles) \approx \left(1 - 4(c + \mu)\frac{2Ns}{e^{2Ns} - 1}\right)^t, \tag{2}$$

198 giving a maintenance probability that is only dependent on the effective population size and the 199 selection coefficient through the compound population parameter 2Ns.

200 Parameters. To compute maintenance probabilities, we set the point mutation rate to its
201 estimate by Sun *et al.* (60) (also supported by earlier work from Crawford (61)) of

202 $\mu = 0.776 \cdot 10^{-9}$ mutations per bp per year. (3)

203 We wished to use the total branch length of the *Leptodactylus* phylogeny for *t*, the

204 maintenance time, to reflect the observation of trans-specific maintenance. In considering the

205 phylogenetic tree and split times here and in the evolutionary simulations of the section

206 "Simulations of ATP1A1 gene family evolution" below, we only considered a subset of three

207 Leptodactylus species—L. fuscus, L. latrans and L. pentadactylus—for which confident species

- split time estimates were available (see "Estimation of genealogical relationships" section; Fig.
- S5): a split between *L. fuscus* and the common ancestor of the two other species 29,187,798
- 210 years ago, followed by a split between *L. latrans* and *L. pentadactylus* 27,426,120 years ago.
- 211 Therefore, the total time on the species tree was set to

213 214The maintenance probabilities shown in Fig. 3A were computed using eq. (1), plugging in the parameters in eq. (3) and (4) and across a grid of (Ns) $\in [-1,1.5]$ and $c \in [0,2.5]$ values.216Simulations of ATP1A1 gene family evolution217Simulations of ATP1A1 gene family evolution218Overview. We developed evolutionary simulations with the goal of gauging the evolutionary parameters that could have produced the observed spatial divergence patterns along ATP1A1. Typically, and whenever possible, analytic likelihood or posterior probability functions are derived for such a task. Alternatively, backward-in-time simulations are used, because of their high computational efficiency. However, analytic or backward-in-time approaches were intractable for our purposes: both because we wished to account for the spatial divergence patterns and not consider sites independently—and because our model of ATP1A1 evolution in <i>Leptodactylus</i> includes complex interactions between point mutation, NAGC, and selection that violate typical assumptions of analytic / backward in time sequence evolution models. We therefore developed a forward-in-time simulation of R and S. The simulations take a set of parameters θ as input (see section "Fitness model and other parameterization" below), start with two ancestral sequences and end with an output of contemporary R and S sequences in multiple <i>Leptodactylus</i> species, which we later compare to the observed data (see section "Inference of evolutionary parameters using Approximate Bayesian Computation").223Fitness model and other parameterization. At the heart of our simulation, we consider the possible fixation of new haplotypes in <i>Leptodactylus</i> lineages. These fixations follow random occurrence of de novo point mutations or NAGC in one of the haplotypes in the population; but the p	212	$t = 2 \cdot 29,187,798 + 27,426,120 = 85,801,716$ years. (4)
216217Simulations of ATP1A1 gene family evolution218Overview. We developed evolutionary simulations with the goal of gauging the219evolutionary parameters that could have produced the observed spatial divergence patterns along210ATP1A1. Typically, and whenever possible, analytic likelihood or posterior probability211functions are derived for such a task. Alternatively, backward-in-time simulations are used,222because of their high computational efficiency. However, analytic or backward-in-time223approaches were intractable for our purposes: both because we wished to account for the spatial224divergence patterns and not consider sites independently—and because our model of ATP1A1225evolution in <i>Leptodactylus</i> includes complex interactions between point mutation, NAGC, and226selection that violate typical assumptions of analytic / backward in time sequence evolution227models. We therefore developed a forward-in-time simulation of R and S. The simulations take228a set of parameters θ as input (see section "Fitness model and other parameterization" below),229start with two ancestral sequences and end with an output of contemporary R and S sequences in230multiple <i>Leptodactylus</i> species, which we later compare to the observed data (see section231"Inference of evolutionary parameters using Approximate Bayesian Computation").232Fitness model and other parameterization. At the heart of our simulation, we consider233the possible fixation of new haplotypes in <i>Leptodactylus</i> lincages. These fixations follow random233occurren	213 214	The maintenance probabilities shown in Fig. 3A were computed using eq. (1), plugging in the
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	235	the probability of fixation on the lineage will depend on the selection acting on the novel variant.

The ancestral haplotype with which the simulation begins is assumed to underlie the
optimal function of R, S and interactions between them, and thus to be of optimal fitness.
Therefore, the absolute fitness *f* of a haplotype *X* at any point of the simulation depends on its

239 divergence from the ancestral haplotype with which the simulation begins, as follows:

240

$$f(X) = s_1 X_1 + s_2 X_2 + s_y Y + s_z Z + s_{12} X_1 X_2 + s_{1y} X_1 Y + s_{2y} X_2 Y,$$

241 where $X_1 \in \{0,1,2\}$ is the number of residue differences between X and the ancestral haplotype at 242 position 111 of the amino acid sequences of both R and S; $X_2 \in \{0,1,2\}$ is the number of residue 243 differences between X and the ancestral haplotype at position 122; $Y \in \{0, 1, ..., 20\}$ is the 244 number of residue differences between X and the ancestral haplotype at the other 10 R/S245 distinguishing substitutions (referring to the substitutions strongly distinguishing R and S in the 246 observed sequences); and Z is the number of total residue differences between X and the ancestral haplotype in the rest of the amino acid sequence. $\{s_1, s_2, s_y, s_z, s_{12}, s_{1y}, s_{2y}\}$ represent 247 248 selection coefficients and are fixed parameters that are taken as input of the simulation. 249 Other parameters taken as input by our simulation (see pseudocode below) include: 250 - N, the population size of each extant *Leptodactylus* lineage 251 - μ , the per haplotype, per nucleotide per year mutation rate. 252 - *l*, the mean NAGC tract length in base pairs. We model the tract length as Geometrically 253 distributed (39, 62). 254 -c, the NAGC per nucleotide per year rate. Note that this is the rate in which a site is included in 255 a NAGC tract, not the rate at which NAGC events initiate at the site. 256 - A rooted species tree, consisting of a bifurcating topology and branch lengths (split times) in 257 years.

258 Simulation pseudocode:

259	1.	Initialize time t to the TMRCA of all species.
260	2.	While $t < today$,
261		2.1. Advance t by t_w , the waiting time for the next mutational event, where
262		$t_w \sim Exp((2N \ haplotypes) \cdot (extant \ species) \cdot$
263		(2 paralogs per species) \cdot (ATP1A1 sequence length) \cdot
264		(rate per nucleotide $c + \mu$))
265		2.2 If $t > time$ for lineage split that had not yet occurred,
266		2.2.1 bifurcate lineage: copy R and S sequences of ancestral lineage into
267		an identical copy and label each of the two sets as one of the lineages.
268		2.3 Draw $U_{event} \sim U(0,1)$. if $U_{event} < \frac{\mu}{\mu+c}$ then the de novo mutational event is
269		a point mutation, else, it is a NAGC event.
270		2.4 Draw (uniformly) an extant species in which the event occurred.
271		2.5 Draw (uniformly) a paralog (R or S) in which the mutation occurred or
272		served as the template for NAGC.
273		2.6 Draw (uniformly) a random nucleotide position where the mutational
274		event occurred.
275		2.7 If the de novo event is a NAGC event,
276		2.5.1 Draw a tract length $L \sim Geo(l)$. Expand tract around initiation site,
277		with a uniform fraction extending to the left and right of the site.
278		2.8 Translate the derived, de novo haplotype and the ancestral haplotype to
279		amino acid sequences and calculate their fitness; calculate the resulting
280		relative fitness of the derived haplotype.

281 2.9 Calculate
$$p_{fix}$$
, the fixation probability (see below) for a haplotype at

282 frequency
$$\frac{1}{2N}$$
 conferring relative fitness as calculated in 2.8.

283
$$2.10 Draw U_{fix} \sim U(0,1). If U_{fix} < p_{fix}$$

- 284 2.10.1 Fix: Replace ancestral haplotype in the species with the de novo
 - haplotype.
- In step 2.9, we consider a *de novo* haplotype arising in the population (namely, at
- 287 frequency $\frac{1}{2N}$) with relative fitness 1 + s to have probability
- 288 frequency $\frac{1}{2N}$) with relative fitness 1 + s to have probability

289
$$p_{fix} = \begin{cases} \frac{e^{s} - 1}{e^{2Ns} - 1} & \text{if } s < 0 \text{ (deleterious)} \\ \frac{1}{2N} & \text{if } s = 0 \text{ (neutral)} \\ \frac{1 - e^{-s}}{1 - e^{-2Ns}} & \text{if } s > 0 \text{ (advantageous)} \end{cases}$$

290 of fixing in the population, following Kimura (58).

291 of fixing in the population, following Kimura (58).

292

285

293 Inference of evolutionary parameters using Approximate Bayesian Computation

Overview. We used an Approximate Bayesian Computation (ABC) approach to estimate evolutionary parameters, including gene conversion rates and the strength of purifying selection acting at different sites in ATP1A1. In each iteration *j*, we sampled a set of parameters θ_j from a predefined prior distribution. We approximated the posterior distribution of θ_j by the empirical distribution given by a subset of this sample that generates divergence patterns that we inferred as closest to the true data. To infer the "distance" of simulated data from the observed data, we ran forward-in-time evolutionary simulations of ATP1A1 sequence evolution and quantified the 301 similarity of the simulated divergence patterns to the observed divergence patterns. Simulations 302 all begin with the same ancestral R and S genes in a common ancestor, and end with six evolved 303 (simulated) contemporary sequences, corresponding to R and S in three *Leptodacylus* species. 304 From the divergence patterns between these six simulated sequences, we computed $d(\Theta_j)$, the

305 distance between the simulated and the observed (real sequence data) ATP1A1 divergence

306 patterns.

307 Parameter set and prior distribution. Our evolutionary simulations take as input a set
308 of parameters as defined in the section "Simulations of ATP1A1 gene family evolution",

309
$$\Theta = \{\mu, c, l, N, s_1, s_2, s_z, s_y, s_{12}, s_{1y}, s_{2y}\}$$

310 The prior distributions of single parameters are mutually independent. Namely, the prior 311 distribution on Θ was set as

312 $\pi(\Theta) = \pi_c(c)\pi_{\tilde{s}}(\tilde{s})\pi_{s_z}(s_z),$

where π_K is the marginal prior distribution of *K*, and $\tilde{s} \coloneqq s_1 = s_2 = s_y$ such that all 12 sites distinguishing R and S in the observed data are under the same selective constraint, but it is free to differ from the selective constraint on other amino acids. The marginal priors on the gene conversion rate *c* and selection coefficients \tilde{s} , s_z were set as

317
$$\log_{10}(\frac{c}{\mu}) \sim U(0, 2.5),$$

318
$$\log_{10}(N\tilde{s}) \sim U(-1,1)$$

319 and

320 $\log_{10}(Ns_y) \sim U(-1,1).$

321 The other parameters were assumed fixed: we set the mutation rate to be $\mu = 0.776 \cdot 10^{-9}$

322 mutations per bp per year and the diploid population size (in each extant species at a given time

323 in the simulation) to be N=10 (2N=20) as in the section "Theoretical single-site model for the

probability of maintaining an adapted substitution". This small population size was chosen to allow for computational efficiency, because the simulation run time scaled linearly with N, and our inference became computationally infeasible with substantially larger population sizes. The mean tract length for gene conversion events was set to l = 100bp. We have found that there is very limited resolution given by our inference scheme on selection interaction terms, s_{12} , s_{1y} and s_{2y} when we allowed them to vary. We therefore set these fitness interaction terms to zero.

Measuring similarity to observed divergence patterns. Given y, a set of R and S 330 331 nucleotide sequences in three species, we computed two summaries of the divergence at each nucleotide site i: $d_o(y_i)$, the sum of pairwise Hamming distances between R sequences in a pair 332 333 of species (each $\in \{0,1\}$ since only one site is considered) plus the sum of pairwise Hamming distances between R sequences; and $d_p(y_i)$, the sum—across the three species—of Hamming 334 distances between paralogous R and S sequences. Let y^{obs} be the six observed sequences and 335 336 y^{Θ_j} be the sequences output at the end of simulation run j. We measured the divergence between 337 the simulated and observed data at site I as

338
$$d_i(\Theta_j) = d_i(y^{obs}, y^{\Theta_j}) = d_o\left(y_i^{obs}, y_i^{\Theta_j}\right) + d_p\left(y_i^{obs}, y_i^{\Theta_j}\right).$$

This per-site distance was computed for all positions *I*, namely nucleotide sites without missing data or insertions/deletions in any of the six observed sequences. Finally, the distance between simulation *j* and the observed data is given by

342
$$d(\Theta_j) = \sum_{\text{sites i}} w_i d_i (y^{obs}, y^{\Theta_j})$$

343 where w_i are position-importance weights, giving extra weight for divergence patterns near R/S 344 distinguishing sites. These weights were set as

345
$$w_i = 1 + \sum_{k=1}^{12} 10 \cdot e^{-|i-i_k|},$$

346 where $\{i_k\}$ is the set of $12 \cdot 3$ positions coding for one of the 12 R/S distinguishing substitution 347 sites.

348 Analysis. We ran 23,323 simulations with Θ sampled from its prior distribution. We kept 349 ~1% of these parameter sets—234 sets which produced simulations with the lowest $d(\cdot)$ values, 350 and considered them as samples from the approximate posterior distribution. We then used the functions kde3d (for the approximate posterior distribution of c, s_z and \tilde{s}) and kde2d (for the 351 352 marginal approximate posterior distribution of c and \tilde{s}) from the R packages misc3d (63) and 353 MASS (64) to estimate the posterior with a spline fit using over 200 bins per dimension, in the 354 range set by our prior distribution on each parameter, and with otherwise default settings of 355 kde3d and kde2d. The approximate posterior mode was $(c = 18\mu, 2N\tilde{s} = 6, 2Ns_z = 1),$ 356 357 and the marginal posterior mode on the first two parameters was 358 $(c = 9\mu, 2N\tilde{s} = 7).$ 359 The (single dimension) marginal credible interval mentioned in the main text are high 360 posterior density credible intervals. 361 362 Construction of expression vectors 363 Na⁺,K⁺-ATPase is a multi-subunit protein that requires co-expression of the alpha (ATP1A) 364 and beta subunits (ATP1B) in cell lines. An RNA-seq analysis of Leptodactylus brain, stomach, 365 and muscle tissues revealed that ATP1B1, one of four paralogous copies of ATP1B, is the most 366 ubiquitously expressed. cDNA was reverse transcribed from Leptodactylus latrans stomach 367 mRNA using the Superscript III Reverse Transcriptase kit (InvitrogenTM). The ATP1B1 gene

- 368 was amplified from cDNA with the primers,
- 369 5'ATCCTCGAGATGGCCAGAGACAAAACCAAGGA 3' and 5'
- 370 TGTGGTACCTCAGCTACTCTTAATCTCCAACTTTA 3', which added a XhoI site at the 5'
- and a KpnI site at the 3' end. ATP1B1 amplicons were inserted into pFastBac Dual
- 372 expression vectors (Life Technologies) at the p10 promoter with XhoI and KpnI (FastDigest;
- 373 Thermo ScientificTM), and then control sequenced. The vector insert sequence was an identical
- match to the *L. latrans* β 1-subunit transcript generated in this study. ATP1A1S was amplified
- 375 from cDNA with the primers 5' TAATACTAGTATGGGATACGGGGCCGGACGTGAT 3'
- 376 and 5' ACTGCGGCCGCTTAATAATAGGTTTCTTTCTCCA 3' and ATP1A1R was amplified
- from a previously constructed vector containing a truncated copy of the gene with the overhangprimers 5'
- 379 TAATACTAGTATGGGATACGGGGCCGGACGTGATGAGTATGAGCCCGCAGCCACTT
- 380 CTGAACATGGCGGCAAGAAGAAGGCAAAGGGAAGGATAAGGAT 3' and 5'
- 381 ACTGCGGCCGCTTAATAATAGGTTTCTTCTCCACCCAGCCGCCAGGGCTGCGTCTG
- 382 ATTATCAGTTTTCGGATTTCATCATATATGAAGATGAGCAGAGAGTAGGGGAAGGC
- 383 ACAGAACCACCATGTTGGTTTCAGTGGGTACATGCGGAGTGCCACATCCATGCCTGG
- 384 G 3'. Both pairs of primers added a SpeI site at the 5' end and a NotI site at the 3' ends. All gene
- 385 amplifications were performed using a high-fidelity proofreading polymerase (Phusion High-
- ³⁸⁶ Fidelity DNA Polymerase; Thermo ScientificTM). ATP1A1S and ATP1A1R amplicons were
- 387 inserted at the P_{PH} promoter of pFastBac Dual expression vectors already containing ATP1B1
- 388 with SpeI and NotI (FastDigest; Thermo ScientificTM), and then control sequenced. The
- 389 ATP1A1S sequence was an identical match to the *L. latrans* sensitive α1-subunit transcripts and
- the ATP1A1R sequence was an identical match to *L. latrans* resistant α1-subunit transcripts

391	generated from this study. <i>Escherichia coli</i> DH5 α cells (Invitrogen TM) were transformed with the
392	two resulting expression vectors (pFastBac Dual + ATP1B1 + ATP1A1S and pFastBac Dual +
393	ATP1B1 + ATP1A1R). These completed vectors were then used to introduce the amino acid
394	codons of interest by site-directed mutagenesis (QuikChange II XL Kit; Agilent Technologies,
395	La Jolla, CA, USA) according to the manufacturer's protocol. One ATP1A1S gene construct was
396	synthesized by Invitrogen TM GeneArt (S+12R). All resulting vectors had the α 1-subunit gene
397	under the control of the P _{PH} promoter and the β 1-subunit gene under the p10 promoter (Table
398	S6).
399	
400	Generation of recombinant viruses and transfection into Sf9 cells
401	Escherichia coli DH10bac cells harboring the baculovirus genome (bacmid) and a
402	transposition helper vector (Life Technologies) were transformed according to the
403	manufacturer's protocol with expression vectors containing the different gene constructs.
404	Recombinant bacmids were selected through PCR screening, grown, and isolated (65).
405	Subsequently, Sf9 cells (4 x 10 ⁵ cells*ml) in 2 ml of Insect-Xpress medium (Lonza,
406	Walkersville, MD, USA) were transfected with recombinant bacmids using Cellfectin reagent
407	(Life Technologies). After a three-day incubation period, recombinant baculoviruses were
408	isolated (P1) and used to infect fresh Sf9 cells (1.2 x 10 ⁶ cells*ml) in 10 ml of Insect-Xpress
409	medium (Lonza, Walkersville, MD, USA) with 15 mg/ml gentamycin (Roth, Karlsruhe,
410	Germany) at a multiplicity of infection of 0.1. Five days after infection, the amplified viruses
411	were harvested (P2 stock).
412	

414 <u>Preparation of Sf9 cell membranes</u>

415 For production of recombinant Na⁺,K⁺-ATPase, Sf9 cells were infected with the P2 viral stock at a multiplicity of infection of 1e3. The cells (1.6 x 10⁶ cells*ml) were grown in 50 ml of 416 417 Insect-Xpress medium (Lonza, Walkersville, MD, USA) with 15 mg/ml gentamycin (Roth, 418 Karlsruhe, Germany) at 27°C in 500 ml flasks (33). After 3 days, Sf9 cells were harvested by 419 centrifugation at 20,000 x g for 10 min. The cells were stored at -80 °C, and then resuspended at 420 0 °C in 15 ml of homogenization buffer (0.25 M sucrose, 2 mM EDTA, and 25 mM 421 HEPES/Tris; pH 7.0). The resuspended cells were sonicated at 60 W (Bandelin Electronic 422 Company, Berlin, Germany) for three 45 s intervals at 0 °C. The cell suspension was then 423 subjected to centrifugation for 30 min at 10,000 x g (J2-21 centrifuge, Beckmann-Coulter, 424 Krefeld, Germany). The supernatant was collected and further centrifuged for 60 min at 100,000 425 x g at 4 °C (Ultra- Centrifuge L-80, Beckmann-Coulter) to pellet the cell membranes. The 426 pelleted membranes were washed once and resuspended in ROTIPURAN® p.a., ACS water 427 (Roth) and stored at -20 °C. Protein concentrations were determined by Bradford assays using 428 bovine serum albumin as a standard. Six biological replicates were produced for each Na⁺,K⁺-429 ATPase construct.

430

431 Verification by SDS-PAGE and western blotting

For each biological replicate, 50 ug of protein were solubilized in 4x SDS-polyacrylamide
gel electrophoresis sample buffer and separated on SDS gels containing 10% acrylamide.

434 Subsequently, they were blotted on nitrocellulose membrane (HP42.1, Roth). To block non-

435 specific binding sites after blotting, the membrane was incubated with 5% dried milk in TBS-

436 Tween 20 for 1 h. After blocking, the membranes were incubated overnight at 4 °C with the

437 primary monoclonal antibody a5 (Developmental Studies Hybridoma Bank, University of Iowa, 438 Iowa City, IA, USA). Since only membrane proteins were isolated from transfected cells, 439 detection of the α subunit also indicates the presence of the β subunit. The primary antibody was 440 detected using a goat-anti-mouse secondary antibody conjugated with horseradish peroxidase 441 (Dianova, Hamburg, Germany). The staining of the precipitated polypeptide-antibody complexes 442 was performed by addition of 60 mg 4-chloro-1 naphtol (Sigma-Aldrich, Taufkirchen, Germany) 443 in 20 ml ice-cold methanol to 100 ml phosphate buffered saline (PBS) containing 60 ul 30% 444 H₂O₂. See Fig. S6.

445

446 <u>Ouabain inhibition assay (measurement of CG resistance)</u>

447 To determine the sensitivity of each Na⁺,K⁺-ATPase construct against the water-soluble 448 cardiac glycoside, ouabain (Acros Organics), 100 ug of each protein was pipetted into each well 449 in a nine-well row on a 96-well microplate (Fisherbrand) containing stabilizing buffers (see 450 buffer formulas in (66). Each well in the nine-well row was exposed to exponentially decreasing 451 concentrations (10⁻³ M, 10⁻⁴ M, 10⁻⁵ M, 10⁻⁶ M, 10⁻⁷ M, 10⁻⁸ M, dissolved in distilled H₂O) of 452 ouabain, distilled water only (experimental control), and a combination of an inhibition buffer lacking KCl and 10⁻² M ouabain to measure background ATPase activity (see (66)). The proteins 453 454 were incubated at 37°C and 200 rpms for 10 minutes on a microplate shaker (Quantifoil 455 Instruments, Jena, Germany). Next, ATP (Sigma Aldrich) was added to each well and the 456 proteins were incubated again at 37°C and 200 rpms for 20 minutes. The activity of Na⁺,K⁺-457 ATPases following ouabain exposure was determined by quantification of inorganic phosphate 458 (Pi) released from enzymatically hydrolyzed ATP. Reaction Pi levels were measured according 459 to the procedure described by Taussky and Shorr (67) (see (66)). All assays were run in duplicate and the average of the two technical replicates was used for subsequent statistical analyses.
Absorbance for each well was measured at 650 nm with a plate absorbance reader (BioRad
Model 680 spectrophotometer and software package).

463

464 <u>ATP hydrolysis assay (measurement of ATPase activity as a proxy for protein activity)</u>

465 To determine the functional efficiency of different Na^+, K^+ -ATPase constructs, we

466 calculated the amount of Pi hydrolyzed from ATP per mg of protein per minute. The

467 measurements were obtained from the same assay as described above. In brief, absorbance from

the experimental control reactions, in which 100 ug of protein was incubated without any

469 inhibiting factors (i.e., ouabain or buffer excluding KCl), were measured and translated to mM Pi

470 from a standard curve that was run in parallel (1.2 mM Pi, 1 mM Pi, 0.8 mM Pi, 0.6 mM Pi, 0.4

471 mM Pi, 0.2 mM Pi, 0 mM Pi).

472

473 <u>Statistical analyses of biochemical assay results</u>

474 Background phosphate absorbance levels from reactions with inhibiting factors were used 475 to calibrate phosphate absorbance in wells measuring ouabain inhibition and in the control wells 476 (66). For ouabain sensitivity measurements, calibrated absorbance values were converted to 477 percentage non-inhibited Na⁺,K⁺-ATPases activity based on measurements from the control 478 wells (66). These data were plotted and $\log IC_{50}$ values were obtained for each biological 479 replicate from nonlinear fitting using a four-parameter logistic curve, with the top asymptote set 480 to 100 and the bottom asymptote set to zero (Fig. S7). Curve fitting was performed with the 481 nlsLM function of the minipack.lm library in R. For comparisons of recombinant protein ATPase 482 activity, the calculated Pi concentrations of 100 ug of protein assayed in the absence of ouabain

483 were converted to nmol Pi/mg protein/min. We used ANOVA to test for effects of substitutions 484 on ouabain resistance (log IC₅₀) and enzyme activity (Table S8; Levene's Test for Homogeneity 485 of Variance for IC₅₀: $F_{7,40}$ =0.68 p=0.69 and enzyme activity: $F_{7,40}$ =0.31 p=0.94). We used linear 486 regression to estimate effect sizes associated with substitutions and pairwise t-tests to identify 487 significant differences between substitution combinations (Table S8). All statistical analyses 488 were implemented in R.

489 Table S1. Collection information for samples of leptodactylid frogs used in this study.

490 ANDES-A refers to the Amphibian collection of the *Museo de Historia Natural C. J. Marinkelle*

491 of the Universidad de los Andes, Bogotá, Colombia. Collector acronyms are Andrew J. Crawford

492 (AJC), Juan Salvador Mendoza (JSM), Juan Manuel Padial (JMP). All collecting sites are

493 located in Colombia (CO). Samples without museum voucher IDs are in the process of being

494 accessioned into the ANDES-A collection.

495

Species	Museum ID	Field ID	Data type	Locality	Latitude, longitude
Engystomops pustulosus		AJC 3734	RNA-seq	Mariquita, Tolima, CO.	05.2635, -074.891
Engystomops pustulosus		JSM 228	intron	Zambrano, Bolívar, CO	09.75, -074.8333
Lithodytes lineatus		AJC 6408	RNA-seq	pending	
Lithodytes lineatus	ANDES-A 2536	AJC 2406	intron	Trubon, Río Vaupés, Vaupés, CO.	01.21, -070.62
Leptodactylus fuscus	ANDES-A 3141	AJC 5344	plasmid	Neiva, Huila, CO.	02.8796, -075.2757
Leptodactylus fuscus		JSM 205	genome	Garzón, Huila, CO.	02.2058, -075.6440
Leptodactylus pentadactylus	ANDES-A 2327	AJC 4761	RNA-seq	Leticia, Amazonas, CO.	-03.865, -070.2061
Leptodactylus pentadactylus	ANDES-A 949	JMP 2179	intron	Leticia, Amazonas, CO.	-04.10592, -069.25
Leptodactylus latrans		AJC 3653	RNA-seq	Puerto Carreño, Vichada, CO.	06.10, -067.483
Leptodactylus latrans	ANDES-A 1148	AJC 3430	intron	Orocué, Casanare, CO.	04.9093, -071.4286
Leptodactylus insularum	ANDES-A 3146	AJC 5345		Neiva, Huila, CO.	02.8441, -075.3328
Leptodactylus insularum		AJC 3752	CDS	Montería, Córdoba, CO.	08.7917, -075.8629
Leptodactylus insularum		JSM 261	intron	Barú, Bolívar, CO.	10.1458, -075.6792
Leptodactylus colombiensis		AJC 5510		Santa María, Boyacá, CO.	04.8499, -073.2653
Leptodactylus colombiensis	ANDES-A 3066	AJC 3755	CDS	Nilo, Cundinamarca, CO.	04.3584, -074.5649
Leptodactylus colombiensis		AJC 4301	intron	San Martín, Meta, CO.	03.6969, -073.6986

497 Table S2. Sources of ATP1A1 sequences included in the phylogenetic analysis (Fig. 1). New

498 data generated by this study are indicated by blue text (RNA-seq datasets: PRJNA627222,

499 genome assembly: PRJNA631731).

Species	Data type and format	Accession
Atelopus zeteki	RNA-seq, PE 140 bp	skin: SRR11583991
Bombina maxima	RNA-seq, PE 90 bp	skin: SRR566619
Bufotes viridis	RNA-seq, PE 100 bp	SRR2163277
Craugastor fitzingeri	RNA-seq, SE 100 bp	skin: SRR1560905
Cyclorana alboguttata	RNA-seq, SE 105 bp	muscle: SRR619475
Dendrobates auratus	RNA-seq, PE 150 bp	brain: SRR11583990
		muscle: SRR11583979
		stomach: SRR11583968
	de novo assembly	CDS: MT813444
Duttaphrynus melanostictus	RNA-seq, PE 150 bp	brain: SRR11583966
	* *	muscle: SRR11583965
		stomach: SRR11583964
	<i>de novo</i> assembly	CDS: MT813445
Engystomops pustulosus	RNA-seq, PE 140 bp	brain: SRR11583963
		stomach: SRR11583962
	de novo assembly	CDS: MT396181
	long-read sequencing	partial gene: MT422192
Fejervarya cancrivora	RNA-seq, PE 100 bp	SRR1554290
Homo sapiens	NCBI reference sequence	NM 001160233.1
Kaloula pulchra	RNA-seq, PE 150 bp	muscle: SRR11583961
		stomach: SRR11583989
	de novo assembly	CDS: MT813446
Leptobrachium boringii	RNA-seq, PE 100 bp	SRR4436787
Leptodactylus colombiensis	cloning, plasmid sequencing	CDS: MT396187 (ATP1A1S)
		MT396188 (ATP1A1R)
	long-read sequencing	partial gene: MT422198 (ATP1A1S)
		MT422199 (ATP1A1R)
Leptodactylus fuscus	cloning, plasmid sequencing	CDS: MT396183 (ATP1A1S)
		MT396184 (ATP1A1R)
	single-molecule genomic	de novo assembly: TBD
	single-molecule genomic sequencing	
		<i>de novo</i> assembly: TBD partial gene: MT422194 (ATP1A1S) MT422195 (ATP1A1R)
Leptodactylus insularum		<i>de novo</i> assembly: TBD partial gene: MT422194 (ATP1A1S)
Leptodactylus insularum	sequencing	<i>de novo</i> assembly: TBD partial gene: MT422194 (ATP1A1S) MT422195 (ATP1A1R)
Leptodactylus insularum	sequencing	de novo assembly: TBD partial gene: MT422194 (ATP1A1S) MT422195 (ATP1A1R) CDS: MT396191 (ATP1A1S) MT396192 (ATP1A1R) partial gene: MT422202 (ATP1A1S)
	sequencing cloning, plasmid sequencing	de novo assembly: TBD partial gene: MT422194 (ATP1A1S) MT422195 (ATP1A1R) CDS: MT396191 (ATP1A1S) MT396192 (ATP1A1R) partial gene: MT422202 (ATP1A1S)
	sequencing cloning, plasmid sequencing	de novo assembly: TBD partial gene: MT422194 (ATP1A1S) MT422195 (ATP1A1R) CDS: MT396191 (ATP1A1S) MT396192 (ATP1A1R) partial gene: MT422202 (ATP1A1S)
	sequencing cloning, plasmid sequencing long-read sequencing	de novo assembly: TBD partial gene: MT422194 (ATP1A1S) MT422195 (ATP1A1R) CDS: MT396191 (ATP1A1S) MT396192 (ATP1A1R) partial gene: MT422202 (ATP1A1S) MT422203 (ATP1A1R) brain: SRR11583988 stomach: SRR11583987
	sequencing cloning, plasmid sequencing long-read sequencing	de novo assembly: TBD partial gene: MT422194 (ATP1A1S) MT422195 (ATP1A1R) CDS: MT396191 (ATP1A1S) MT396192 (ATP1A1R) partial gene: MT422202 (ATP1A1S) MT422203 (ATP1A1R) brain: SRR11583988
	sequencing cloning, plasmid sequencing long-read sequencing RNA-seq, SE 140 bp de novo assembly	de novo assembly: TBD partial gene: MT422194 (ATP1A1S) MT422195 (ATP1A1R) CDS: MT396191 (ATP1A1S) MT396192 (ATP1A1R) partial gene: MT422202 (ATP1A1S) MT422203 (ATP1A1R) brain: SRR11583988 stomach: SRR11583987 CDS: MT396189 (ATP1A1S) MT396190 (ATP1A1R)
	sequencing cloning, plasmid sequencing long-read sequencing RNA-seq, SE 140 bp	de novo assembly: TBD partial gene: MT422194 (ATP1A1S) MT422195 (ATP1A1R) CDS: MT396191 (ATP1A1S) MT396192 (ATP1A1R) partial gene: MT422202 (ATP1A1S) MT422203 (ATP1A1R) brain: SRR11583988 stomach: SRR11583987 CDS: MT396189 (ATP1A1S)
Leptodactylus latrans	sequencing cloning, plasmid sequencing long-read sequencing RNA-seq, SE 140 bp de novo assembly	de novo assembly: TBD partial gene: MT422194 (ATP1A1S) MT422195 (ATP1A1R) CDS: MT396191 (ATP1A1S) MT396192 (ATP1A1R) partial gene: MT422202 (ATP1A1S) MT422203 (ATP1A1R) brain: SRR11583988 stomach: SRR11583987 CDS: MT396189 (ATP1A1S) MT396190 (ATP1A1R)
Leptodactylus insularum Leptodactylus latrans Leptodactylus pentadactylus	sequencing cloning, plasmid sequencing long-read sequencing RNA-seq, SE 140 bp de novo assembly	de novo assembly: TBD partial gene: MT422194 (ATP1A1S) MT422195 (ATP1A1R) CDS: MT396191 (ATP1A1S) MT396192 (ATP1A1R) partial gene: MT422202 (ATP1A1S) MT422203 (ATP1A1R) brain: SRR11583988 stomach: SRR11583987 CDS: MT396189 (ATP1A1S) MT396190 (ATP1A1R) partial gene: MT422200 (ATP1A1S)

	de novo assembly	CDS: MT396185 (ATP1A1S)
		MT396186 (ATP1A1R)
	long-read sequencing	partial gene: MT422196 (ATP1A1S)
		MT4221957 (ATP1A1R)
Limnodynastes peronii	RNA-seq, PE 100 bp	SRR8712702
Lithodytes lineatus	RNA-seq, PE 75 bp	muscle: SRR11583984
		stomach: SRR11583983
	de novo assembly	CDS: MT396182
	long-read sequencing	partial gene: MT422193
Mantella betsileo	RNA-seq, PE 90 bp	skin: SRR7592160
Megophrys nasuta	RNA-seq, PE 150 bp	brain: SRR11583982
		muscle: SRR11583981
		stomach: SRR11583980
	<i>de novo</i> assembly	CDS: MT813448
Melanophryniscus stelzneri	RNA-seq, PE 150 bp	brain: SRR11583978
* *	* *	muscle: SRR11583977
		stomach: SRR11583976
	de novo assembly	CDS: MT813449
Odorrana tormota	RNA-seq, PE 150 bp	skin: SRR6896138
Oreobates cruralis	RNA-seq, PE 126 bp	intestine: SRR5507183
Oreolalax rhodostigmatus	RNA-seq, PE 150 bp	SRR6265740
Pelobates fuscus	RNA-seq, PE 90 bp	SRR5119616
Pelophylax lessonae	RNA-seq, PE 90 bp, PE	SRR1164893
Quasipaa boulengeri	RNA-seq, PE 100 bp, PE	SRR2962603
Rana catesbeiana	RNA-seq, PE 150 bp	brain: SRR11583975
		muscle: SRR11583974
		stomach: SRR11583973
	de novo assembly	CDS: MT813450
Rana sphenocephala	RNA-seq, PE 150 bp	brain: SRR11583972
~ *	- · ·	muscle: SRR11583971
		stomach: SRR11583970
	de novo assembly	CDS: MT813451
Rattus norvegicus	NCBI reference sequence	NM_012504.1
Rhinella marina	RNA-seq, PE 140 bp	brain: SRR11583969
	- · ·	skin: SRR11583967
	de novo assembly	CDS: MT813452
Xenopus tropicalis	NCBI reference sequence	NM 204076.1

502 Table S3. Summary of the *de novo* genome assembly of *Leptodactylus fuscus*

503

Sequencer	HiSeq X
Assembly software	Supernova 2.1.1
Number of reads	775.95 million
Read format	Paired end 150 bp
Effective read depth coverage	48.35
Estimated genome size	2.42 Gb
Weighted mean molecule size	29.36 kb
Number of scaffolds ≥ 10 kb (long scaffolds)	16,530
N50 contig size	19.69 kb
N50 scaffold size	362.61 kb
Assembly size (only scaffolds >= 10 kb)	1.26 Gb
BUSCO version	4.0.5
Lineage dataset	Tetrapoda_odb10
Input genome format	supernove pseudohap2_2
Total groups searched	5310
Complete BUSCOs	3182 (60.0%)
Complete and single-copy BUSCOs	3041 (57.3%)
Complete and duplicated BUSCOs	141 (2.7%)
Fragmented BUSCOs	669 (12.6%)

1459 (27.4%)

504

Missing BUSCOs

Table S4. Fisher's exact test (FET) for the likelihood of site-wise support of "Non-Concerted" and "Concerted" topologies.

507

	# sites	Non-Concerted topology	Concerted topology	FET p-value		
Nonsynonymous	28	17	11			
Synonymous	164	9	153	- 3.2e-8		

Table S5. List of primers used for intron sequencing

Species	Primer
Engystomops	N-terminal
pustulosus	Forward: EP_wwBC6_1F
-	GATGTAGAGGGTACGGTTTGAGGCACATGGCGGCAAGAAGAA
	Reverse: EP_wwBC6_11R
	GATGTAGAGGGTACGGTTTGAGGCGTGGAGCATCGGTCCAGGA
	C-terminal
	Forward: EP_wwBC7_11F
	GGCTCCATAGGAACTCACGCTACTGATCCTGGACCGATGCTCCA
	Reverse: EP_wwBC7_19R
	GGCTCCATAGGAACTCACGCTACTTGACAATGCTGACGAAGAAGGC
Lithodytes	Forward: Lep_wwBC3_1F
lineatus	TACATGCTCCTGTTGTTAGGGAGGACATGGCGGCAAGAAGAA
	Reverse: Lep_wwBC3_21R
	TACATGCTCCTGTTGTTAGGGAGGAGGAGGCACAGAACCACCATGT
Leptodactylus	Forward: Lep_wwBC5_1F
pentadactylus	ACAGCATCAATGTTTGGCTAGTTGACATGGCGGCAAGAAGAA
	Reverse: Lep_wwBC5_21R
	ACAGCATCAATGTTTGGCTAGTTGAGGCACAGAACCACCATGT
Leptodactylus	Forward: Lep_wwBC2_1F
latrans	AGGTGATCCCAACAAGCGTAAGTAACATGGCGGCAAGAAGAA
	Reverse: Lep_wwBC2_21R
	AGGTGATCCCAACAAGCGTAAGTAAGGCACAGAACCACCATGT
Leptodactylus	Forward: Lep_wwBC1_1F
insularum	AACGGAGGAGTTAGTTGGATGATCACATGGCGGCAAGAAGAA
	Reverse: Lep_wwBC1_21R
	AACGGAGGAGTTAGTTGGATGATCAGGCACAGAACCACCATGT
Leptodactylus	Forward: Lep_wwBC8_23F
colombiensis	AGAGGGTACTATGTGCCTCAGCACAAGTATGAGCCCGCAGCCACTTC
	Reverse: Lep_wwBC8_3044R
	AGAGGGTACTATGTGCCTCAGCACCCAGGGCTGCGTCTGATGATTAA

514 Table S6. List of engineered ATP1A1 constructs used to test functional effects of amino

515 acid substitutions in *Leptodactylus*.

516

Construct Name	Engineered Substitution(s)	Description
S	-	Sensitive (S) paralog of L. latrans ATP1A1
S+Q111R	Q111R	Q111R on the S paralog background
S+N122D	N122D	N122D on the S paralog background
S+Q111R+N122D	Q111R + N122D	Q111R and N122D on the S paralog background
S+10subs	A112T, E116D, I135V, L180Q, I199L, I279V, S403C, L536M, Q701L, I788M	All substitutions strongly distinguishing R and S paralogs, except Q111R and N122D, on the S paralog background
R-Q111R-N122D	R111Q, D122N	Reversions R111Q and D122N on the R paralog background
S+12subs	Q111R, A112T, E116D, N122D, I135V, L180Q, I199L, I279V, S403C, L536M, Q701L, I788M	Twelve substitutions strongly distinguishing R and S paralogs on the S paralog background
R	-	Resistant (R) paralog of L. latrans ATP1A1

517 Note: R and S paralogs of *L. latrans* differ by the 12 substitutions that are the focus of this study

and by 9 additional amino-acid substitutions and a two-amino acid insertion-deletion difference.

519 Our experiments revealed that these 10 *latrans*-specific substitutions do not contribute detectably

520 to S vs. R differences in CG resistance of enzyme function (using all 10 as one co-variate,

ANOVA p>0.5. Following convention, positions of substitutions are standardized relative to the sheep (*Ovis aries*) sequence NM 001009360 - 5 AA from 5' end.

524 Table S7. Summary of the ouabain sensitivity and catalytic properties of Na⁺,K⁺-ATPase

525 for each ATP1A1 construct. The values represent the mean and standard deviation (SD)

526 ouabain sensitivity $(log_{10}IC_{50})$ of ATPase activity of six biological replicates. ATP1B1 of

527 Leptodactylus latrans was co-expressed with ATP1A1.

528

ATP1A1 Constructs	Ouabain sensitivity (mol/L) Mean(log10 IC ₅₀) ± SD	ATPase activity nmol Pi/(mg protein*min) ± SD
S	-5.63 ± 0.59	16.30 ± 5.01
S+Q111R	-4.89 ± 0.85	10.68 ± 3.71
S+N122D	-5.06 ± 0.66	7.16 ± 3.62
S+Q111R+N122D	-3.62 ± 0.28	8.29 ± 3.86
S+10subs	-5.82 ± 0.47	16.37 ± 3.05
R-Q111R-N122D	-5.60 ± 0.33	17.41 ± 2.87
S+12subs	-3.23 ± 0.75	12.17 ± 2.43
R	-3.25 ± 0.77	14.09 ± 2.77

529

530

532 Table S8. Statistical analysis of ouabain sensitivity and ATPase activity. Significant p values

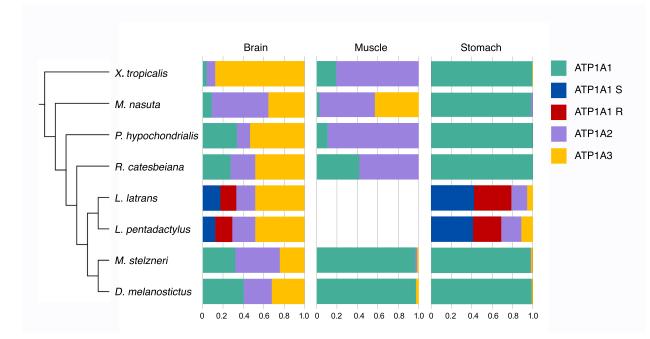
533 are highlighted in bold.

(Explanatory Variables) ANOVA	Ouaba log10(in sensit IC ₅₀)	ivity			ATPase activity nmol Pi/(mg protein*min)							
	df	MS	F	p value	df	MS	F	p value					
Q111R N122D	1, 42 1, 42	42.8 11.8	107.8 27.72	2.7e-13 2.3e-6	1, 42 1, 42	83.0 101.4	6.9 7.98	0.015 7.2e-3					
10subs R-S background Q111R:N122D	1, 42 1, 42 1, 42	0.59 0.04	1.6 1.9	0.22 0.74	1, 42 1, 42 1, 42 1, 42	101.4 228.1 11.5 7.6	7.98 17.96 0.34 5.64	1.2e-3 1.2e-4 0.34 0.022					

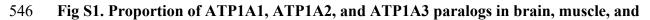
(Explanatory Variables) Linear regression	Ouaba log10(in sensit IC ₅₀)	ivity			ATPase activity nmol Pi/(mg protein*min)							
	Est	SE	t	p value	Est	SE	t	p value					
Intercept	-6.03	0.17	-36.3	<2e-16	14.3	1.19	12.1	3e-15					
Q111R	1.32	0.21	6.27	2.7e-13	-4.39	1.88	-2.34	0.024					
N122D	1.14	0.21	5.45	2.3e-6	-6.81	1.88	-3.63	7.7e-4					
10subs	0.26	0.22	1.19	0.24	1.94	1.46	17.96	0.18					
R-S background	-0.08	0.26	-0.34	0.74	1.39	1.46	0.34	0.35					
Q111R:N122D	-	-	-	-	6.90	2.91	5.64	0.022					

537 Note: "R-S background" in the ANOVA refers to 9 additional amino acid substitutions and a two

amino acid insertion-deletion difference that distinguishes the R and S constructs (derived from
 Leptodactylus latrans).







547 stomach of seven anuran species. RNA-seq reads for eight species were mapped to species-

- 548 specific copies of ATP1A1, ATP1A2, and ATP1A3 using bwa (see Methods). Uniquely mapped
- 549 reads were counted for each paralog and estimated as a proportion of the sum of the reads for all

three ATP1A paralogs. X. tropicalis: Xenopus tropicalis; M. nasuta: Megophrys nasuta; P.

- 551 *hypochondrialis: Phyllomedusa hypochondrialis; R. catesbeiana: Rana catesbeiana; L. latrans:*
- 552 Leptodactylus latrans; L. pentadactylus: Leptodactylus pentadactylus; M. stelzneri:
- 553 *Melanophryniscus stelzneri*; *D. melanostictus: Duttaphrynus melanostictus.*

						AT	Ρ1	A1							A٦	'P1	LA	2							Α	ΤР	1A	3		
		1	1	1	1	1	1	1	1 1	ι 1	1	1	1	1	1	1	1	1	1	1	1	1	1	1	1	1	1	1 1	1	. 1
		0	1	1	1	1	1	1	1 1	L 2	2	0	1	1	1	1	1	1	1	2	2	0	1	1	1	1	1	1 1	1 2	2
		8	1	2	4	5	6	5 7	7 9	9 0	2	8	1	2	4	5	6	7	9	0	2	8	1	2	4	5	6	7 9) C) 2
Family	Species	Y	Q	A	Т	E	E	E	Ē	2 N	I N	Y	Q	A	Μ	Ε	D	Ε	Q	Ν	N	Y	Q	A	т	Е	D	D S	5 0	δN
Human	Homo sapiens		•																S				•		•		•		• •	
Brown rat	Rattus norvegicus	•	R	S	·	•	•		۰F	.	D	•	L	·	·	·	·	·	S	·	·	·	·	·	·	·	·	• •	• •	
Bombinatoridae	Bombina maxima	•	·	·	·	•	D)	•	• •	•																			
Pipidae	Xenopus tropicalis	•	Т	·	•		•		• 1	г·	•	•	·	I	·	·	·	·	Т	·	·									•
Pelobatidae	Pelobates fuscus	•	•	·	•		•		•	• •	•	•	·	Т	·	•	·	·	Т	·	•	•	•	•	•	·	•	۰A	ł .	
Megophryidae	Megophrys nasuta	•	·	·	•		•		۰N	١·		•	·	·	·	·	·	·	·	·		•	·	·	·	·	·	۰A	<i>۱</i> ،	
Megophryidae	Oreolalax rhodostigmatus	•	·	·	•		•		•			•	·	Т	·	·	·	·	Т	·		•	·	·						
Megophryidae	Leptobrachium boringii		•		•				•														•	•	•		•	· A	<i>۱</i> ۰	
Microhylidae	Kaloula pulchra		•		•			- [с -					Т			·		L	·										
Mantellidae	Mantella betsileo																							L	Μ			ΕI		1.
Dicroglossidae	Quasipaa boulengeri						D)																						
Dicroglossidae	Fejervarya cancrivora							- [<u>с</u>					T					T											
Ranidae	Pelophylax lessonae													Т					Т									· A	A M	1.
Ranidae	Odorrana tormota													T					T						М			· A	A M	1.
Ranidae	Rana sphenocephala			T	Ν	۱.	D)	•					T					T				L						. N	1.
Ranidae	Rana catesbeiana													T					T				L						. N	1.
Myobatrachidae	Limnodynastes peronii													T					T									· 4	A M	1.
Hylidae	Cyclorana alboguttata													T					T											
Hylidae	Phyllomedusa hypochondrialis													T					T									· 4	۰ <i>μ</i>	
, Craugastoridae	Craugastor fitzingeri													T					T											
Strabomantidae	Oreobates cruralis													T					T											
Leptodactylidae	Engystomops pustulosus													T					T				•					· A	4.	
Leptodactylidae	Lithodytes lineatus								<u>с</u>					T					T											
	Leptodactylus latrans S																													
Leptodactylidae	Leptodactylus latrans R		R	т		۰.	C	5			D		•	I	·	•	·	·	I	·	•	•	·	·	•	·	•	• 4	۰ ۱	•
Dendrobatidae	Dendrobates auratus													I					T									• 1	г·	
Bufonidae	Melanophryniscus stelzneri	н	L	V			D)	• •	۰ ۱				T					М								•	N A		1.
Bufonidae	Atelopus zeteki		R	к	S	C	L		• [<u>.</u>		1										1								
Bufonidae	Duttaphrynus melanostictus		R	к	S	C	L		• [<u>,</u>			т	V	i.			D	т				L					Е·	P	
Bufonidae	Bufotes viridis		R	к	S	C	L		• [<u>.</u>			т	V	÷.			D	т				R	к	S	D	L	ΕC	1 C	1 - 1
Bufonidae	Rhinella marina		R	к	S	C	L		• [<u>.</u>		1											L					Е·	P	

556 Fig S2. Variation among sites implicated in CG-resistance for ATP1A paralogs of various

557 species. Sequences of ATP1A2 and ATP1A3 were reconstructed using the same method as

558 ATP1A1 described in Materials and Methods. Consensus sequences of anuran species were

generated in MEGA 7.0 and used as reference for each paralog. Only sites implicated in CG-

560 resistance are shown. Following convention, positions of substitutions, shown at the top, are

aligned relative to the sheep (Ovis aries) sequence NM 001009360 subtracting 5 AA from 5'end

562 (e.g., the first position is 108). A dot indicates identity with the reference sequence. ATP1A1S

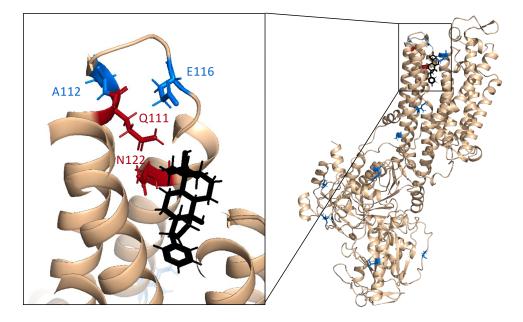
and ATP1A1R of *Leptodactylus latrans* are indicated in blue and red, respectively. Bufonid

564 (toad) species, the prey species that produce CG toxins, are highlighted in purple. Blank: missing

565 data. We failed to identify an ortholog of ATP1A4 in any of the available anuran genome

assemblies, including our assembly of *Leptodactylus fuscus*.

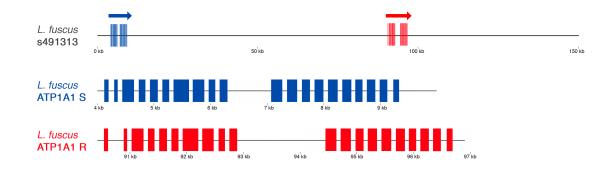
- 567
- 568



573 Fig S3. Positions of 12 R copy-specific amino acid substitutions on the crystal structure of

pig ATP1A1 bound to the cardiac glycoside bufalin. Red residues correspond to key CG

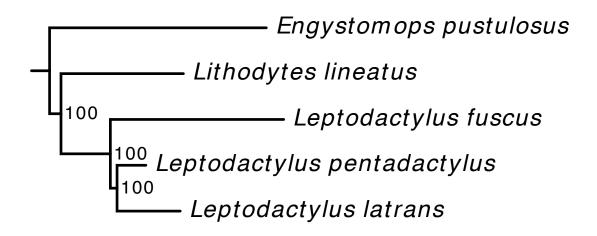
- 575 resistance-conferring sites 111 and 122; blue residues correspond to 10 additional residues
- 576 strongly associated with variants at sites 111 and 122. (Sus scrofa, PDB 4RES.)



579 Fig S4. Annotation of ATP1A1S and ATP1A1R paralogs in the Leptodactylus fuscus de

- 580 *novo* genome assembly (Table S3). ATP1A1S (blue) and ATP1A1R (red) occur in tandem on
- 581 scaffold s491313 (Genbank Acc#) ~80 kilobases apart. The boundary between exons and introns
- 582 were determined by blast and manual correction (*i.e.*, ensuring that each intron started with GT
- and ended with AG). The gene structure figures were plotted with ggbio in R.

584





587 Figure S5. A phylogenetic tree of *Leptodactylus* and outgroup species. The tree was

- 588 constructed using an alignment of 813 orthologous mRNA sequences under the best partition
- 589 model with iq-tree 2.0 (see Methods). Branch lengths: (*Engystomops pustulosus*:0.699436,
- 590 (*Lithodytes lineatus*:0.395135, (*Leptodactylus fuscus*:0.559378, (*Leptodactylus pentadactylus*:
- 591 0.092965, *Leptodactylus latrans*: 0.203845)100:0.0216082)100:0.159296)100:0.0368124).

kDa		1A	1B	1C	1D	1E	1F	2A	2B	2C	2D	2E	2F		3A	3B	3C	3D	3E	3F
100	$\alpha \rightarrow $	-				-			-	-				-	-	-	-	-	-	-
nmol Pi/m	ng protein	13.24	13.77	12.56	16.12	25.80	13.73	14.11	6.60	7.00	13.00	12.71	6.14		12.09	10.72	3.65	4.23	5.11	9.23
kDa		4A	4B	4C	4D	4E	4F	5A	5B	5C	5D	5E	5F		6A	6B	6C	6D	6E	6F
100	$\alpha \rightarrow$	-	-	-				-	-	-	-	-			=	-			-	
nmol Pi/m	ng protein	14.05	9.24	6.02	9.56	2.58	9.40	13.89	17.73	18.78	16.84	14.60	10.43		22.17	16.84	18.03	14.55	15.44	14.80
kDa		7A	7B	7C	7D	7E	7F	8A	8B	8C	8D	8E	8F		1 – a 2 – a	S S + C	2111R			
100	$\alpha \rightarrow \phi$	-	-	-	-	-	-	-		-	-	-	-			S + N				
nmol Pi/m	g protein	14.17	10.26	10.20	14.20	11.99	15.43	10.55	12.46	15.79	18.49	13.17	13.65		5 – a 6 – a	S + C S +1(R - Q S + 1 R)R 111R			



593 Fig S6. Western blot analysis of Na⁺,K⁺-ATPase with engineered ATP1A1 (α) subunits

produced in this study. The 110 kDa ATP1A1 protein is stained with the α5 monoclonal

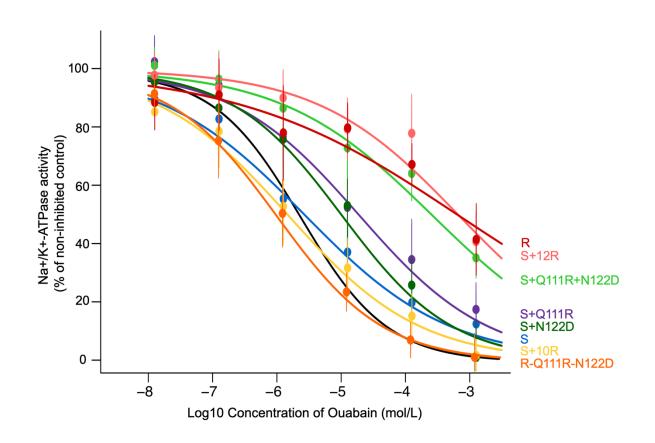
antibody followed by a horseradish peroxidase conjugated goat antimouse antibody. Samples

596 represent six biological replicates of eight different recombinant Na⁺,K⁺-ATPase (Table S5)

597 produced through cell culture. All western blots are aligned to the same protein ladder level.

598 ATPase activity levels (nmol Pi/mg protein) of each biological replicate is indicated under its

- 599 respective band.
- 600
- 601



604 Fig S7. Cardiac glycoside (ouabain) inhibition curves for six each engineered Leptodactylus 605 Na⁺,K⁺-ATPase produced in this study. Points and error bars represent the mean \pm SEM (n=6 606 biological replicates) percentage of protein activity relative to controls measured in the absence 607 of ouabain and excluding the activity of background ATPases. The black inhibition curve was 608 measured from commercially procured porcine cerebral cortex (CAS 9000-83-3, Sigma-Aldrich, 609 Inc.) and represents a standard bench-mark reference for cardiac glycoside sensitivity (Log10 610 IC₅₀= -5.61). The ATP1B1 of *Leptodactylus latrans* was co-expressed with each engineered 611 version of ATP1A1 (Table S5).

612