Transcriptomic Evidence for Reproductive Suppression in Male *Peromyscus eremicus* (Cactus Mouse)

Subjected to Acute Dehydration

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Abstract

Understanding how organisms adapt to extreme environments is an outstanding question facing
evolutionary biologists. Research related to a specific example of adaptation, mammals in desert
environments, has focused on survival, while questions related to the reproductive effects of dehydration
have been largely ignored. Here, we explore the reproductive consequences of acute dehydration by
utilizing RNAseq data in the desert-specialized rodent, *Peromyscus eremicus*. Nine genes were
consistently differentially expressed between hydrated and dehydrated mice, a low number which aligns
with current perceptions of this species' extreme desert specialization. However, these differentially
expressed genes include Insulin-like 3 (Insl3), a regulator of male fertility, as well as Slc45a3 and Slc38a5,
both of which interact with genes with important roles in reproductive function. Together, these findings
suggest that acute dehydration is linked to reproductive mitigation, a result which is unexpected in an
animal capable of surviving and successfully reproducing without available external water sources.

Introduction

Populations native to particular environments are believed to possess physiological, behavioral and ecological adaptations that enable their long term persistence. For decades, evolutionary biologists have successfully described examples where natural selection has resulted in the exquisite match between organism and environment (*e.g.* Salinity adaptations in three-spine sticklebacks: Hohenlohe et al., 2010; Jones et al. 2012; high-altitude adaptations for hemoglobin: Storz et al., 2010, Lorenzo et al., 2015; and *Peromyscus* adaptations for multiple environments: Hoekstra et al., 2006; Bedford & Hoekstra, 2015; Munshi-South & Richardson, 2016). The match between organism and environment has largely been studied in the context of one component of fitness – survival – while the other component, reproductive success, has received negligible attention in the context of environmental adaptations (but see Kordonowy & MacManes, 2016). However, this frequently neglected aspect of selection is critical to long term persistence in a given environment. While substantial research has been done on the effects of various types of stress on reproduction (*e.g.* Wingfield & Sapolsky, 2003; Ahmed et al., 2015; Nargund, 2015; Wingfield, 2013), the impacts of dehydration stress on reproduction in extreme environmental specialists have not been studied. Habitat specialists must possess phenotypes enabling survival and successful reproduction; therefore, cases where environmental selective pressures result in reduced reproductive success (*e.g.* Martin & Wiebe, 2004; Bolger, Patten & Bostock, 2005; Evans et al., 2010; Wingfield, Kelley & Angelier, 2011), but not survival, demand attention. Species occupying extreme environments are likely more vulnerable to the bifurcation of these two components of fitness. Moreover, long-term events like global climate change are predicted to increase climate variability and may enhance the challenges faced by species living on the fringes of habitable environments (Martin & Wiebe, 2004; Somero, 2010; Wingfield, Kelley & Angelier, 2011; Wingfield, 2013; Asres & Amha, 2014).

Deserts present extraordinary environmental impediments for habitation, including extreme heat, aridity, and solar radiation. Examples of well-described desert mammal behavioral adaptations are seasonal torpor (reviewed in Kalabukhov 1960; Geiser, 2010), nocturnality (*e.g.* Stephens & Tello, 2009; Fuller et al., 2014) and burrowing (reviewed in Vorhies, 1945; Kelt, 2011) to avoid high temperatures and sun exposure. Desert mammals also exhibit a wide range of morphological adaptations, including large ears for effective heat dissipation (*e.g.* Schmidt-Nieslen, 1964; Hill & Veghte, 1976), metabolic water production (*e.g.* MacMillen & Hinds, 1983; reviewed in Walsberg 2000), and renal adaptations to minimize water-loss (*e.g.* Schmidt-Nielsen et al., 1948; Dantzler, 1982; Diaz, Ojeda & Rezenda, 2006).

Although desert rodents must possess adaptations conferring survival *and* reproductive benefits,
researchers have focused exclusively on their adaptations for survival. For example, renal adaptations in
species of Kangaroo rats (*Dipodomys* species) have been described and explored for over 60 years
(Schmidt-Nielsen et al., 1948; Schmidt-Nielsen and Schmidt-Nielsen, 1952; Marra et al., 2012; Urity et
al., 2012). While early research determined the renal physiology for Kangaroo rats (Schmidt-Nielsen et
al., 1948; Schmidt-Nielsen and Schmidt-Nielsen, 1952; Vimtrup and Schmidt-Nielsen), recent research
has focused on the genetic underpinnings of this phenotype (Marra et al., 2012; Urity et al., 2012; Marra,
Romero & DeWoody, 2014; Marra et al., 2014), which is indicative of a larger methodological shift in
the approach for examining adaptation.

Research in another desert-adapted rodent, *Peromyscus eremicus* (cactus mouse), has followed a
somewhat different trajectory; however, it too has only pursued survival oriented mechanisms (but see
Kordonowy and MacManes, 2016). The ecology, physiology and behaviors of the cactus mouse in
comparison with other *Peromyscus* species were summarized in 1968 (King, ed.), and the relationships
between basal metabolic rate, body mass, and evaporative water loss were reviewed several decades later
(MacMillen and Garland, 1989). Known desert adaptations for cactus mouse include nocturnality and
torpor (reviewed in Veal and Caire, 1979); however, the cactus mouse does not possess the same elaborate
kidney structures responsible for renal adaptations in kangaroo rats (Dewey, Elias & Appel, 1966;
MacManes 2016, *unpublished data*). The physiological renal adaptations in *P. eremicus* have not been
described in detail, despite considerable explorations of other aspects of this species' biology (reviewed
in Veal and Caire, 1979). In order to initially characterize renal function of the cactus mouse, water
consumption measurements and electrophysical dehydration effects for this species have also recently
been documented (Kordonowy et al., 2016). Because the renal mechanisms for mitigating renal water-
loss in *P. eremicus* have not been determined, a comparative genetic approach may be instrumental for
characterizing this species' adaptive kidney phenotype. To this end, MacManes and Eisen (2014)
conducted a comparative analysis to find genes expressed in the kidney tissue of cactus mouse that were
under positive selection relative to other mammals. Furthermore, the transcriptomic resources available
for this species extend considerably beyond renal tissue; transcripts from cactus mouse (as well as
numerous other *Peromyscus* species) have been heavily utilized to pursue questions related to multiple
aspects of evolutionary biology (reviewed in Bedford and Hoekstra, 2015; Munshi-South and Richardson,
2016). Current investigations into cactus mouse desert-adaptive physiology include transcriptomic

analyses; however, we extend this genetic approach by shifting the focus from adaptions for survival to include adaptations for reproductive success.

In nature, wild cactus mice are subjected to both acute and chronic dehydration, and understanding the reproductive effects of dehydration stress is an initial step for fully characterizing the suite of phenotypes enabling successful reproduction. Given that this species has evolved in southwestern United States deserts, we predicted that neither acute nor chronic water stress, while physiologically demanding, would be associated with reproductive suppression. To test these predictions, we leveraged previous research that characterized the transcriptome of male *P. eremicus* reproductive tissues from functional and comparative perspectives (Kordonowy and MacManes, 2016). We extend upon this work by performing an RNAseq experiment to identify differentially expressed genes in testes between male *P. eremicus* subjected to acute dehydration versus control (fully hydrated) animals in order to determine the impacts, if any, on male fertility. We pursue this line of research on the effects of dehydration on reproduction in cactus mouse in order to begin to address the need for studies focusing on adaptation related to reproductive success in animals living in extreme, and changing, environments.

Methods

Treatment Groups, Sample Preparation and mRNA Sequencing

The *cactus mice* used for this study include only captive born individuals purchased from the *Peromyscus* Genetic Stock Center (Columbia, South Carolina). The animals, originally collected from a hot-desert location in Arizona, have been housed for several generations, at the University of New Hampshire in conditions that mimic temperature and humidity levels in southwestern US deserts, as described previously (Kordonowy & MacManes, 2016). Males and females are housed together, which provides olfactory cues to support reproductive maturation. Males do not undergo seasonal testicular atrophy, as indicated by successful reproduction throughout the year. The individuals used in this study were deemed reproductively mature once they became scrotal.

Males that were provided with water *ad libidum* had free access to water prior to euthanasia, and these individuals are labeled as WET mice in our analyses. Mice that were water deprived, which we refer to as DRY mice, were weighed and then water deprived for ~72 hours directly prior to euthanasia. All

mice were weighed prior to sacrifice, and DRY mice were evaluated for weight loss during dehydration.
Individuals in the study were collected between September 2014 – April 2016.

Cactus mice were sacrificed via isoflurane overdose and decapitation in accordance with
University of New Hampshire Animal Care and Use Committee guidelines (protocol number 130902) and
guidelines established by the American Society of Mammalogists (Sikes et al., 2016). Trunk blood
samples were collected following decapitation for serum electrolyte analyses with an Abaxis Vetscan VS2
using critical care cartridges (Abaxis). The complete methodology and results of the electrolyte study, as
well as the reported measures of water consumption and weight loss due to dehydration are described fully
elsewhere (Kordonowy et al., 2016). Rather, this study focused on differential gene expression between
the testes of 11 WET and 11 DRY mice. Testes were harvested within ten minutes of euthanasia, placed
in RNAlater (Ambion Life Technologies), flash-frozen in liquid nitrogen, and stored at -80° degree
Celsius. A TRIzol, chloroform protocol was implemented for RNA extraction (Ambion Life
Technologies). Finally, the quantity and quality of the RNA product was evaluated with both a Qubit 2.0
Fluorometer (Invitrogen) and a Tapestation 2200 (Agilent Technologies, Palo Alto, USA).

Libraries were made with a TruSeq Stranded mRNA Sample Prep LT Kit (Illumina), and the
quality and quantity of the resultant sequencing libraries were confirmed with the Qubit and Tapestation.
Each sample was ligated with a unique adapter for identification in multiplex single lane sequencing. We
submitted the multiplexed samples of the libraries for processing on lanes at the New York Genome Center
Sequencing Facility (NY, New York). Paired end sequencing reads of length 125bp were generated on an
Illumina 2500 platform. Reads were parsed by individual samples according to their unique hexamer IDs
in preparation for analysis.

Assembly of Testes Transcriptome

We assembled a testes transcriptome from a single reproductively mature male using the *de novo*
transcriptome protocol described previously (MacManes, 2016). The testes transcripts were assembled
with alternative methodologies utilizing several optimization procedures to produce a high-quality
transcriptome; however, the permutations of this assembly process are described extensively elsewhere
(MacManes, 2016; Kordonowy and MacManes, 2016). The testes transcriptome we selected was
constructed as described below. The raw reads were error corrected using Rcorrector version 1.0.1 (Song
& Florea, 2015), then subjected to quality trimming (using a threshold of PHRED <2, as per MacManes
2014) and adapter removal using Skewer version 0.1.127 (Jiang et al, 2014). These reads were then

assembled in the *de novo* transcriptome assembler BinPacker version 1.0 (Liu et al., 2016). We also
reduced sequence redundancy to improve the assembly using the sequence clustering software CD-HIT-
EST version 4.6 (Li & Godzik, 2006; Fu et al., 2012). We further optimized the assembly with Transrate
version 1.0.1 (Smith-Unna et al., 2015) by retaining only highly supported contigs. We then evaluated the
assembly's structural integrity with Transrate and assessed completeness using the vertebrata database in
BUSCO version 1.1b1 (Simão et al., 2015). We quasimapped the raw reads to the assembly with Salmon
version 0.7.2 (Patro, Duggal & Kingsford, 2015) to confirm that mapping rates were high. Finally, the
assembly was also annotated in dammit version 0.3.2, which finds open reading frames with TransDecoder
and uses five databases (Rfam, Pfam, OrthoDB, BUSCO, and Uniref90) to thoroughly annotate transcripts
(https://github.com/camillescott/dammit).

Differential Gene and Transcript Expression Analyses

Several recent studies have critically evaluated alternative methodologies for differential transcript
and gene expression to determine the relative merits of these approaches (Gierlinski et al., 2015; Schurch
et al., 2016; Soneson, Love & Robinson, 2016; Froussios et al., 2016). Soneson and colleagues (2016)
demonstrated that differential gene expression (DGE) analyses produce more accurate results than
differential transcript expression (DTE) analyses. Furthermore, the differential gene expression approach
is more appropriate than differential transcript expression for the scope of our research question, which is
true of many evolutionary genomic studies (Soneson et al., 2016). However, because both DTE and DGE
approaches are widespread in current literature, we deemed it important to confirm that these
methodologies yielded concordant results in the current study.

We utilized edgeR (Robinson, McCarthy & Smith, 2010; McCarthy, Chen & Smith, 2012) as our
primary statistical software because Schurch and colleagues (2016) rigorously tested various packages for
analyzing DGE, and edgeR performed optimally within our sample size range. While edgeR is a widely
used statistical package for evaluating differential expression, we also confirmed our results with another
popular package, DESeq2 (Love, Huber & Anders, 2014), in order to validate our findings.

We performed differential expression analyses with three alternative methodologies. Two analyses
were conducted in R version 3.3.1 (R Core Team, 2016) using edgeR version 3.16.1, a Bioconductor
package (release 3.4) that evaluates statistical differences in count data between treatment groups
(Robinson, McCarthy & Smith, 2010; McCarthy, Chen & Smith, 2012). Our first method utilized
txtimport, an R package developed by Soneson and colleagues (2016), which incorporates transcriptome

mapping-rate estimates with a gene count matrix to enable downstream DGE analysis. The authors assert

that such transcriptome mapping can generate more accurate estimates of DGE than traditional pipelines

(Soneson et al, 2016). While our first methodology evaluated differential gene expression, our second

analysis used the transcriptome mapped read sets to perform differential transcript expression and identify

the corresponding gene matches. The purpose of this second analysis was to evaluate whether the

transcript expression results coincided with the gene expression results produced by the same program,

edgeR. Finally, our third methodology determined differential gene expression with tximport in

conjunction with DESeq2 version 1.14.0 (Love, Huber & Anders, 2014), a Bioconductor package (release

3.4) which also evaluates statistical differences in expression. We performed this alternative DGE analysis

with DESeq2 in order to corroborate our DGE results from edgeR. Thus, the results for all three

differential expression analyses were evaluated to determine the coincidence among the genes identified

as significantly different between the WET and DRY groups. These alternative differential expression

methods are described in detail below.

We quasimapped each of the 11 WET and 11 DRY sample read sets to the testes transcriptome

with Salmon version 0.7.2 to generate transcript count data. To perform the gene-level analysis in edgeR,

we constructed a gene ID to transcript ID mapping file, which was generated by a BLASTn (Altschul et

al., 1990; Madden, 2002) search for matches in the *Mus musculus* transcriptome (ensembl.org) version

7/11/16 release-85. We then imported the Salmon-generated count data and the gene ID to transcript ID

mapping file into R using the tximport package (Soneson et al. 2016) to convert the transcript count data

into gene counts. This gene count data was imported into edgeR for differential gene expression analysis

(Robinson, McCarthy & Smith, 2010; McCarthy, Chen & Smith, 2012). We applied TMM normalization

to the data, calculated common and tagwise dispersions, and performed exact tests (p < 0.05) adjusting

for multiple comparisons with the Benjamini-Hochburg correction (Benjamini & Hochburg, 1995) to find

differentially expressed genes, which we identified in Ensembl (ensemble.org).

Next, we performed a transcript-level analysis using edgeR. To accomplish this, the Salmon-

generated count data was imported into R and analyzed as was described above for the gene-level analysis

in edgeR. After determining which transcript IDs were differentially expressed, we identified the

corresponding genes using the gene ID to transcript ID matrix described previously. The significantly

expressed transcripts without corresponding gene matches were selected for an additional BLASTn search

in the NCBI non-redundant nucleotide database (http://blast.ncbi.nlm.nih.gov/Blast.cgi). However, these

results were not subjected to any additional analyses, because these matches were not consistent across all
three differential expression analyses. This list of BLASTn search matches is provided in supplementary
materials (DTEno-matchBLASTnSequences.md).

The third analysis used DESeq2 to conduct an additional gene-level test, using the same methods
as described for the previous gene-level analysis, with the exception that data were imported into an
alternative software package. We determined the significantly differentially expressed genes ($p < 0.05$)
based on normalized counts and using the Benjamini-Hochburg correction (Benjamini & Hochburg, 1995)
for multiple comparisons. We only retained genes with a $-1 < \log_2$ fold change > 1 in order to filter genes
at a conservative threshold for differential expression based on our sample size (Schurch et al., 2016).
This filtering was not necessary for either of the edgeR analyses because $\log_2$ fold changes exceeded this
threshold for the differentially expressed genes and transcripts ($-1.3 < \log_2$ fold change > 1.4, in all cases).

We also compared the $\log_2$ fold change values (of treatment differences by mapped count) for each
gene from the edgeR and DESeq2 gene-level analyses in a linear regression. This statistical test was
performed in order to evaluate the degree of concordance between the two DGE analyses. Furthermore,
we constructed a list of genes identified as differentially expressed by all three analyses, which were
further evaluated for function as well as chromosomal location. These genes were also explored in
STRING version 10.0 (string-db.org) to determine their protein-protein interactions (Snel et al., 2000;
Szklarczyk et al., 2015).

Results

Data and Code Availability

The testes transcriptome was assembled from a 45.8 million paired read data set. Additionally,
there were 9-20 million paired reads for each of the 22 testes data sets used for the differential expression
analysis **(Supplemental 1)**, yielding 304,466,486 reads total for this analysis. The raw reads are available
at the European Nucleotide Archive under study accession number PRJEB18655. All data files, including
the testes un-annotated transcriptome, the dammit annotated transcriptome, and the data generated by the
differential gene expression analysis (described below) are available on DropBox
(https://www.dropbox.com/sh/ffr9xrmjxj9md1m/AACpxjQNn-Jlf25qNdslfRSCa?dl=0). These files will

be posted to Dryad upon manuscript acceptance. All code for these analyses is posted on GitHub

(https://github.com/macmanes-lab/testesDGE).

Assembly of Testes Transcriptome

242 The performance of multiple transcriptome assemblies was evaluated thoroughly, and the selected

243 optimized testes assembly met high quality and completeness standards, and it also contains relatively few

244 contigs and has high read mapping rates (**Table 1**). Therefore, this transcriptome was used for our

245 differential expression analyses. The transcriptome was also annotated, and the complete statistics for this

246 dammit annotation are provided in **Table 2**.

Differential Gene and Transcript Expression Analyses

Salmon quasimapping rates of all read datasets to the assembly were sufficiently high (range:

81.46% - 87.02%; **Supplemental 1**), indicating the successful generation of transcript count data for our

differential expression analyses. The exact test performed for our gene-level analysis in edgeR indicated

that fifteen genes reached statistical significance (after adjusting for multiple comparisons) for DGE

between the WET and DRY treatment groups **(Figure 1)**. Specifically, seven genes were more highly

expressed in WET individuals, and eight genes were more highly expressed in DRY individuals **(Table

3).**

We also performed an alternative transcript-level analysis using the referenced transcriptome

mapped reads exclusively with edgeR. The exact test found 66 differentially expressed transcripts **(Figure

2)**, 45 of which were more highly expressed in the WET group, and 21 were more highly expressed in the

DRY group **(Table 4).** 10 of these differentially expressed transcripts were consistent with differentially

expressed genes from the edgeR DGE analysis. In addition, the significantly expressed transcripts without

an Ensembl ID match (nine WET and nine DRY) were retrieved for performing an nt all species BLASTn

search (http://blast.ncbi.nlm.nih.gov/Blast.cgi), and these results are in the supplementary materials.

The gene-level analysis conducted in DESeq2 yielded 215 significantly differentially expressed

genes **(Figure 3),** 67 of which were more highly expressed in the WET group, while 148 were highly

expressed in the DRY group. However, only 20 of these genes remained when we filtered them with a -1

$< \log_2$ fold change > 1 to retain genes with a conservative threshold difference between treatment groups.

This list of 20 genes yielded 16 genes more highly expressed in WET mice and four genes highly

expressed in DRY mice **(Table 5**). Nine of these genes overlapped with those found to be significant in
the previous two edgeR analyses.

To evaluate the correlation of $\log_2$ fold change results for each gene (Ensembl ID) from the two
DGE analyses (EdgeR and DESeq2), we performed a regression of these log values, and they were
significantly correlated (**Figure 4:** Adj-R^2 = 0.6596; $F(1,14214) = 2.754 \times 10^4$; $p < 2.2 \times 10^{-16}$). This further
demonstrates the concordance of the DGE analyses in these two software packages.

To evaluate the degree to which the three analyses produced concordant results, we generated a
list of genes which were found to be significantly differently expressed by treatment across all three
analyses **(Supplemental 2).** There were six genes that were consistently highly-expressed in the WET
group and three genes that were highly-expressed in the DRY group. The six highly-expressed WET genes
are Insulin-like 3 (Insl3), Free-fatty acid receptor 4 (Ffar4), Solute carrier family 45 member 3 (Slc45a3),
Solute carrier family 38 member 5 (Slc38a5), Integrin alpha L (Itgal), and Transferrin (Trf). The three
highly-expressed DRY genes are Ras and Rab Interactor 2 (Rin2), Insulin-like growth factory binding
protein 3 (Igfbp3), and Connective tissue growth factor (Ctgf). Because the patterns of expression of these
nine genes were corroborated by multiple methodologies, we are confident that they are differentially
expressed between our treatments. Estimates of expression for these genes generated using the gene-level
edgeR analysis are plotted in **Figure 5**.

The significantly differently expressed genes were evaluated for gene function and chromosomal
location (**Table 6**). These genes occur throughout the genome; namely, they are located on different
chromosomes. The diverse functions of each gene will be described below. In addition, we generated
STRING diagrams (string-db.org) to view the protein-protein interactions for each of these nine genes
(Snel et al., 2000; Szklarczyk et al., 2015).

Slc38a5 and Slc45a3 are among the highly expressed genes in the WET group (they have lower
expression in the DRY group); these two solute carriers are members of a large protein family that is
responsible for cross-membrane solute transport (reviewed in Hediger et al., 2004; Hediger et al., 2013;
Cesar-Razquin et al., 2015). Slc38a5 is involved sodium-dependent amino-acid transport, while Slc45a3
is purported to transport sugars (Vitavska and Wieczorek, 2013; Schiöth, et al., 2013;
http://slc.bioparadigms.org/), thereby playing an important potential role in maintaining water balance via
management of oncotic pressures. Slc38a5 (**Figure 6a**) has interactions with multiple additional solute
carriers, including Slc1a5, Slc36a2, Slc36a3, and Slc36a4. Slc38a5 also has an interaction with disintegrin

and metalloproteinase domain-containing 7 (Adam7), which is involved in sperm maturation and the

acrosome reaction (Oh et al., 2005). In contrast, Slc45a3 (**Figure 6b**) does not have known protein

interactions with other solute carriers; however, this protein does interact with steroidogenic acute

regulatory protein (StAR), which is critical in steroidogenesis (Christenson and Strauss III, 2001).

Insl3 was lower expressed in the DRY group, and this hormone regulates fertility in male and

female mammals, by preventing apoptosis of germ cells in reproductive organs of both sexes (Kawamura

et al., 2004: Bathgate et al., 2012; Bathgate et al., 2013). In male rodents, Insl3 is critical to development

by facilitating testicular descent, and it is also present in testes of adults, where it binds to relaxin family

peptide receptor 2 (RXFP2), also known as Lrg8 (Bathgate et al., 2012; Bathgate et al., 2013). Within

adult rodent testes, luteinizing hormone (LH) stimulates expression of Insl3 in Leydig cells, and Insl3

binds to Lrg8 in seminiferous tubules, which results in inhibited apoptosis of germ-line cells, thus

increasing their availability (Kawarmura et al., 2004). Therefore, our finding of lower Insl3 expression in

DRY mice indicates less attenuation of germ-line cell apoptosis, which could result in the reduction of

sperm cells and lower fertility. Protein interaction data for Insl3 (**Figure 6c**) indicate that this hormone

interacts with RXFP2 and RFXP1, as well as other proteins, including leptin (Lep), a pleiotropic hormone

involved in reproduction, immunity, and metabolism (reviewed in Friedman, 2014).

Ffar4 was also down-regulated in the DRY group. Omega-3 fatty acid receptor 1 (O3Far1) is an

alias of Ffar4, and it has roles in metabolism and inflammation (Moniri, 2016). This protein interacts with

multiple other free fatty acid receptors and G-protein coupled receptors as well as Stanniocalcin 1 (Stc1)

(**Figure 6d**). Stc1 is involved in phosphate and calcium transportation (Wagner and Dimattia, 2006);

however, this protein's functional role in mice remains enigmatic (Chang et al, 2005).

Another of the lower expressed DRY group genes is Itgal (also known as CD11a), which has

multifaceted roles in lymphocyte-mediated immune responses (Bose et al., 2014). Concordantly, the

protein interactions with Itgal (**Figure 6e**) include numerous proteins integral to immunity, such as

Intracellular adhesion molecules (specifically, ICAM1,2,4), which are expressed on the cell surface of

immune cells and endothelial cells. Itgal is a receptor for these ICAM glycoproteins, which bind during

immune system responses (reviewed in Albelda, Smith and Ward, 1994). However, an additional role of

intercellular adhesion molecules has been proposed in spermatogenesis, whereby ICAMs may be integral

to transporting non-mobile developing sperm cells through the seminiferous epithelium (Xiao, Mruk and

Cheng, 2013).

The final gene with lower expression levels in the DRY treatment is Trf, which modulates the
amount of free-iron in circulation and binds to transferrin receptors on the surface of erythrocyte
precursors to deliver iron (reviewed in Gkouvastos Papanikolaou and Pantopoulos, 2012). TRF interacts
with multiple proteins (**Figure 6f**) involved in iron transport and uptake, including STEAP family member
3 (Steap3), hephaestin (HEPH), cerulopslamin (Cp), Solute carrier protein 40 member 1 (Slc40a1), and
several H+ ATPases. Furthermore, TRF is linked to apolipoprotein A-1 (Apoa1), which interacts with
immunoglobulin in a complex named sperm activating protein (SPAP) to activate the motility of sperm
when it inhabits the female genital tract (Akerlof et al., 1991; Leijonhufvud, Akerlof and Pousette, 1997).

One of the highly expressed genes in the DRY group is Rin2, which is involved in endocytosis
(reviewed in Doherty and McMahon, 2009) and membrane trafficking through its actions as an effector
protein for the GTPases in the Rab family within the Ras superfamily (reviewed in Stenmark and
Olkkonen, 2001). Rin2 protein-protein interactions **(Figure 7a)** include Ras related protein Rab5a and
Rab5b, which are involved in vesicle transport as well as vasopressin-regulated water reabsorption. This
mechanism for water reabsorption via Aquaporin 2 (AQP2) in the kidney has been thoroughly reviewed
by Boone and Deen (2008) and Kwon and colleagues (2013).

The second gene highly expressed in the DRY group is Igfbp3, which modulates the effects of
insulin growth factors. Thus, the protein directly interacts **(Figure 7b)** with insulin growth factors 1 and
2 (Igf1, Igf2), which are responsible for increasing growth in most tissues (reviewed in le Roth 1997;
Jones and Clemmons, 2008). Ctgf was also highly expressed in the DRY group, and this protein is
responsible for increased fibrosis and extracellular matrix formation (Reviewed in Moussad and
Brigstock, 2000). The protein interactions for Ctgf (**Figure 7c**) include many transcription activators in
the Hippo signaling pathway, including multiple TEA domain transcription factors (Tead1, 2, 3 and 4),
WW domain containing transcription regulator 1 (Wwtr1), as well as Yes-associated protein 1 (Yap1),
which is responsible for both increasing apoptosis and preventing cell proliferation to mitigate tumor
growth and control organ size (Reviewed in Pan, 2010).

Discussion

This is the first study to evaluate the reproductive correlates of acute dehydration in a desert-
specialized rodent. We analyzed differential gene expression levels for testes in male *Peromyscus*

eremicus (cactus mouse) in acute dehydration (DRY) versus a control group that was fully hydrated
(WET). Our results provide evidence suggesting that reproductive function is attenuated in acutely
dehydrated mice, which is surprising, given that this is not consistent with our understanding of *P.*
eremicus as a desert specialist. While acute dehydration is less common than chronic dehydration for
desert mammals, it is a selective force they must overcome. Indeed, throughout much of the described
range of the cactus mouse, rainfall events may occur several times per year. Cactus mice, and many other
rodents, are known to rehydrate during these rainfall events (MacManes, *personal observation*). Following
rehydration, cactus mice experience acute dehydration, followed by a steady state of chronic dehydration.

In male *P. eremicus*, we found strong evidence for reproductive suppression via down-regulation
of the Insl3 gene in acutely dehydrated animals. Insl3, which is a hormonal regulator of fertility among
mammals of both sexes, is known to inhibit germ line apoptosis in the testes (Kawamura et al., 2004;
Bathgate et al., 2012; Bathgate et al., 2013). Specifically, the binding of Insl3 to Lrg8 in rodent testes
decreases the apoptosis of germ-line cells, which increases sperm availability (Kawamura et al., 2004).
Lower Insl3 expression in the testes of acutely dehydrated mice suggests that fertility may be attenuated
due to acute water deprivation. Although the current study suggests reproductive suppression in acute
dehydration, future work characterizing the functional consequences of Insl3 down-regulation is needed.
Specifically, does the number or quality of sperm decrease, and does this decrease reduce the probability
of successful fertilization? Moreover, what are the temporal dynamics of reproductive suppression?
Logically, desert species with core reproductive functions that are suppressed by dehydration seem likely
to be rapidly outcompeted. Given this assertion, research characterizing the reproductive correlates of
chronic dehydration is a logical extension of this work, although doing so here is beyond the scope of this
study.

Solute carrier proteins, specifically Slc45a3 and Slc38a5, are also downregulated in acute
dehydration. These genes are part of a large family essential for transferring solutes across membranes
(reviewed in Hediger et al., 2004; Hediger et al., 2013; Cesar-Razquin et al., 2015). Another member of
this family, Solute carrier family 2 member 9 (Slc2a9), has been found to be undergoing positive selection
in studies on kidney transcriptomes of cactus mouse (MacManes & Eisen, 2014) and of other desert
rodents (Marra, Romero & DeWoody, 2014). Our previous work with the male reproductive transcriptome
of cactus mouse found evidence for positive selection in two additional solute carrier proteins: Slc15a3
and Slc47a1 (Kordonowy and MacManes, 2016). Therefore, our current findings that two solute carrier

proteins are lower expressed in the DRY treatment group is consistent with previous research in the kidney
and male reproductive transcriptomes for this species. This leads us to further support the hypothesis
originally proposed by Marra, Romero & DeWoody (2014) that this protein family is intrinsic to
osmoregulation in desert rodents. Indeed, the findings of MacManes and Eisen (2014) and Kordonowy
and MacManes (2016) also lend support to the essential role of solute carrier proteins for maintaining
homeostasis in the desert specialized cactus mouse.

In addition to their well characterized role in the maintenance of water and electrolyte balance, the
differential expression of solute carrier proteins may have important reproductive consequences,
particularly as they relate to hormone secretion. Indeed, the interaction between Slc38a5 and Adam7 is
relevant, because Adam7 is involved in sperm maturation and the acrosome reaction (Oh et al., 2005).
Furthermore, the protein-protein interactions between Slc45a3 with STaR and between Insl3 and Lep are
of particular interest because both STaR and Lep are integral to reproduction, as well as to homeostasis
(reviewed in Christenson and Strauss III, 2001; Anuka et al., 2013; Friedman, 2014; Allison and Myers,
2014). Thus, the protein interactions with reproductive implications are not restricted to solute carrier
proteins. The protein relationships between Itgal and intercellular adhesion molecules are also noteworthy
with respect to research hypothesizing an integral role for ICAMs in spermatogenesis (Xiao, Mruk and
Cheng, 2013). Furthermore, Trf is linked to Apoa1, which is a critical component of sperm activating
protein (Akerlof et al., 1991; Leijonhufvud, Akerlof and Pousette, 1997). While the relationship between
these differentially expressed genes and the hormones involved in reproductive function are currently
poorly-characterized, our findings that genes integral to sperm development and activation interact with
genes differentially expressed in acute dehydration provide strong evidence that, contrary to our
expectations, acute dehydration is linked to reproductive suppression in the cactus mouse.

In contrast to genes that are down-regulated in dehydration, the genes that were upregulated in the
DRY group are known to be responsible for water homeostasis and cellular growth. The significance of
Rin2 is notable, because this protein is an effector for Rab5, which as a GTPase involved in vasopressin-
regulated water reabsorption, a critical homeostatic process mediated through the AQP2 water channel in
kidneys (Boone and Deen, 2008; Kwon et al., 2013). It is not surprising that genes in addition to solute
carrier proteins, which are implicated in alternative processes for water homeostasis, are differentially
expressed in response to water limitation. The other two genes that are up-regulated in the DRY treatment
are indicative of modulated growth due to water limitation. Specifically, Igfb3 interacts directly with

insulin growth factors responsible for tissue growth (le Roth 1997; Jones and Clemmons, 2008), and Ctgf

is linked with numerous transcription factors in the Hippo signaling pathway, which modulates apoptosis,

proliferation and organ size control (Pan, 2010).

Emerging from this work is a hypothesis related to the reproductive response to water stress in the

cactus mouse, and perhaps other desert animals. Specifically, we hypothesize that while acute dehydration

is related to reproductive suppression, chronic dehydration is not. Indeed, it is virtually oxymoronic to

suggest that chronic dehydration, which is the baseline condition in desert animals, has negative

consequences for reproductive success. Generating an integrative, systems-level understanding of the

response to dehydration is required for testing this hypothesis. While understanding the renal response to

dehydration is critical for making predictions about survival, understanding the reproductive correlates is

perhaps even more relevant to evolutionary fitness. Though decades of research have characterized the

former, this study, to the best of our knowledge, is the first to describe the reproductive correlates of water-

limitation. Developing this understanding, particularly in light of global climate change, where increasing

climate variability is predicted, is necessary if we are to understand how climate change may modify the

distribution of extant organisms.

432   Conclusion

The genetic mechanisms responsible for physiological adaptations for survival and reproduction

in deserts remain enigmatic. Previous research has focused exclusively on adaptations related to survival,

specifically on renal adaptations to combat extreme water-limitation. In contrast, no research related to

the reproductive effects of acute and chronic dehydration on desert-adapted rodents has been previously

conducted. To this end, we characterized the reproductive correlates of acute dehydration in a desert-

specialized rodent, the cactus mouse, using a highly replicated RNAseq experiment. In contrast to

expectations, we describe a signal of reproduction suppression in dehydrated male mice. Specifically,

dehydrated mice demonstrated significantly lower expression of Insl3, which is a regulator of fertility, as

well as Slc45a3 and Slc38a5, both of which have interactions with genes important to reproductive

function. While the low number of differentially expressed genes between acutely dehydrated and control

mice might otherwise have suggested that this species is relatively unaffected by acute water-limitation,

the diminished expression of these genes, particularly Insl3, in dehydrated mice indicates the potential for

compromised fertility. Although an experimental demonstration of a functional relationship between this

gene expression pattern and reduced fertility is still needed, our finding that acute-dehydration alters the

expression of genes critical to reproductive function is concerning, particularly with respect to global

climate change. Climate change driven increased variabilities in weather patterns may result in a greater

frequency of acute water-stress, which could result in reduced reproductive function for the cactus mouse.

In addition, because global climate change is predicted to shift habitats toward extremes in temperature,

salinity, and aridity, and to alter species ranges, an enhanced understanding of the reproductive

consequences of these changes, and of the potential for organisms to rapidly adapt, may enable us to

effectively conserve innumerable species facing dramatic habitat changes.

Table 1: Transcriptome assembly performance metrics: contig number (n), TransRate score (Score),

BUSCO indices: % single copy orthologs (% SCO), % duplicated copy orthologs (% DCO), %

fragmented (% frag), and % missing (% miss), as well as Salmon mapping rates (% map) for the

optimized testes assembly.

Assembly Description	n	Score	% SCO	% DCO	% frag	% miss	% map
BinPacker CD-hit-est TransRate Corrected	155134	0.335	77	27	5.9	16	92.14

Table 2: Dammit transcriptome assembly annotation statistics. Dammit annotation includes searches in

the program TransDecoder for open reading frames (ORFs) and searches for homologous sequences in

five databases: Rfam, Pfam-A, Uniref90, OrthoDB, and BUSCO. Percentages were calculated from the

count number of each parameter divided by the total number of contigs in the transcriptome (155,134).

The only exception to this calculation is for complete ORFs, which were calculated as a percentage of

the total ORFs (75,482). The BUSCO results for the annotated assembly are not shown here as they are

identical to those for the un-annotated assembly (Table 2).

Search Type	TransDecoder		Rfam	Pfam-A	Uniref90	OrthoDB	Dammit
Parameter	Total ORFs	Complete ORFs	ncRNAs	Protein Domains	Proteins	Orthologs	Total Annotated Contigs
Count	75,482	43,028	937	25,675	62,865	51,806	77,915
Percentage	48.7%	57.0 %	0.6 %	16.6 %	40.5 %	33.4 %	50.2 %

472 Table 3: EdgeR determined significantly differentially expressed genes by treatment group in *P.*

473 *eremicus* testes. Of the 15 DGE, seven were significantly more highly expressed in WET mice (High in

474 WET) and eight were more highly expressed in DRY mice (High in DRY).

Ensembl ID	log$_2$FC	logCPM	FDR	Gene ID	HIGH
ENSMUSG00000079019.2	-4.354	1.650	5.82E-09	Insl3	WET
ENSMUSG00000054200.6	-3.734	0.619	1.82E-06	Ffar4	WET
ENSMUSG00000026435.15	-2.448	2.447	1.13E-03	Slc45a3	WET
ENSMUSG00000025020.11	-2.231	1.770	1.13E-03	Slit1	WET
ENSMUSG00000031170.14	-2.421	2.578	1.13E-03	Slc38a5	WET
ENSMUSG00000030830.18	-2.180	1.666	3.37E-02	Itgal	WET
ENSMUSG00000032554.15	-2.066	3.287	4.85E-02	Trf	WET
ENSMUSG00000001768.15	3.086	1.006	1.46E-07	Rin2	DRY
ENSMUSG00000025479.9	2.971	3.001	7.97E-05	Cyp2e1	DRY
ENSMUSG00000020427.11	2.681	3.887	1.13E-03	Igfbp3	DRY
ENSMUSG00000019997.11	2.314	3.235	1.13E-03	Ctgf	DRY
ENSMUSG00000040170.13	1.951	0.753	1.72E-03	Fmo2	DRY
ENSMUSG00000023915.4	1.534	1.290	2.02E-02	Tnfrsf21	DRY
ENSMUSG00000052974.8	2.077	0.647	2.26E-02	Cyp2f2	DRY
ENSMUSG00000027901.12	2.492	-0.620	4.78E-02	Dennd2d	DRY

477   Table 4: EdgeR determined significantly differentially expressed transcripts by treatment group in *P.*

478   *eremicus* testes. Of the 66 total DTE, 45 were significantly more highly expressed in WET mice (High

479   in WET) and 21 were more highly expressed in DRY mice (High in DRY). BLASTn matches to

480   Ensembl IDs and corresponding Gene IDs.

HIGH: WET					
Transcript ID	**log$_2$FC**	**logCPM**	**FDR**	**Ensembl ID**	**Gene**
BINPACKER.15365.1	-3.703	0.047	5.31E-11	ENSMUSG00000054200.6	Ffar4
BINPACKER.2960.1	-4.268	1.147	2.06E-09	ENSMUSG00000079019.2	Insl3
BINPACKER.17981.2	-2.975	0.436	6.29E-08	ENSMUSG00000026435.15	Slc45a3
BINPACKER.9961.2	-2.426	1.998	7.50E-07	ENSMUSG00000031170.14	Slc38a5
BINPACKER.3452.1	-2.507	-0.140	3.56E-06	no match	-
BINPACKER.724.4	-2.162	2.667	8.32E-06	ENSMUSG00000032554.15	Trf
BINPACKER.9604.1	-2.582	0.547	7.87E-05	no match	-
BINPACKER.31087.1	-2.908	-0.858	9.74E-05	no match	-
BINPACKER.24398.1	-2.440	-0.689	9.74E-05	ENSMUSG00000036596.6	Cpz
BINPACKER.9726.1	-3.474	-0.107	2.38E-04	ENSMUSG00000026435.15	Slc45a3
BINPACKER.9218.3	-1.578	1.525	2.76E-04	ENSMUSG00000021253.6	Tgfb3
BINPACKER.18534.1	-2.332	1.346	4.85E-04	ENSMUSG00000025020.11	Slit1
BINPACKER.17022.3	-2.899	-0.561	1.00E-03	no match	-
BINPACKER.13806.1	-2.442	-0.381	1.13E-03	ENSMUSG00000025172.2	Ankrd2
BINPACKER.7740.1	-2.790	1.095	1.13E-03	ENSMUSG00000057074.6	Ces1g
BINPACKER.10034.2	-4.420	0.387	1.23E-03	ENSMUSG00000026516.8	Nvl
BINPACKER.11560.2	-1.465	2.050	1.66E-03	ENSMUSG00000021913.7	Ogdhl
BINPACKER.13701.1	-1.312	1.804	2.28E-03	ENSMUSG00000025648.17	Pfkfb4
BINPACKER.3510.3	-2.163	0.906	2.95E-03	ENSMUSG00000027822.16	Slc33a1
BINPACKER.15806.1	-1.700	1.062	3.39E-03	ENSMUSG00000015702.13	Anxa9
BINPACKER.17992.1	-2.542	0.653	3.39E-03	ENSMUSG00000030830.18	Itgal
BINPACKER.9726.2	-2.119	0.560	3.48E-03	ENSMUSG00000026435.15	Slc45a3
BINPACKER.6383.3	-2.093	1.270	4.16E-03	ENSMUSG00000002109.14	Ddb2

Transcript ID	log₂FC	logCPM	FDR	Ensembl ID	Gene
BINPACKER.20716.2	-4.204	-0.566	5.75E-03	ENSMUSG00000013846.9	St3gal1
BINPACKER.20114.1	-1.661	0.501	5.97E-03	ENSMUSG00000030972.6	Acsm5
BINPACKER.18622.1	-1.645	1.704	6.36E-03	no match	-
BINPACKER.24914.1	-2.211	-0.159	9.83E-03	ENSMUSG00000003555.7	Cyp17a1
BINPACKER.31815.1	-1.905	-0.770	9.83E-03	no match	-
BINPACKER.6740.3	-3.090	-0.434	1.04E-02	no match	-
BINPACKER.20530.1	-1.626	0.545	1.12E-02	ENSMUSG00000038463.8	Olfml2b
BINPACKER.20656.1	-1.910	-0.531	1.22E-02	ENSMUSG00000029373.7	Pf4
BINPACKER.4855.1	-1.340	4.025	1.23E-02	ENSMUSG00000059991.7	Nptx2
BINPACKER.1846.1	-3.280	-0.792	1.23E-02	no match	-
BINPACKER.6494.2	-3.363	0.029	1.26E-02	ENSMUSG00000052861.13	Dnah6
BINPACKER.1818.1	-1.713	3.289	2.03E-02	ENSMUSG00000024125.1	Sbpl
BINPACKER.10743.2	-1.915	-0.525	2.06E-02	ENSMUSG00000041607.16	Mbp
BINPACKER.13054.2	-1.147	2.697	2.06E-02	ENSMUSG00000022994.8	Adcy6
BINPACKER.6807.1	-1.330	2.106	2.13E-02	ENSMUSG00000046687.5	Gm5424
BINPACKER.14160.1	-2.051	0.603	2.86E-02	ENSMUSG00000041556.8	Fbxo2
BINPACKER.16191.1	-1.431	0.926	3.42E-02	ENSMUSG00000028654.13	Mycl
BINPACKER.10141.3	-3.283	-1.191	3.68E-02	ENSMUSG00000024132.5	Eci1
BINPACKER.23790.1	-1.756	-0.275	4.51E-02	ENSMUSG00000001119.7	Col6a1
BINPACKER.22521.1	-1.841	-0.056	4.52E-02	ENSMUSG00000054083.8	Capn12
BINPACKER.1061.6	-1.807	1.943	4.93E-02	no match	-
BINPACKER.17734.1	-1.660	2.109	4.94E-02	ENSMUSG00000049608.8	Gpr55
HIGH: DRY					
Transcript ID	**log₂FC**	**logCPM**	**FDR**	**Ensembl ID**	**Gene**
BINPACKER.21794.1	2.434	3.117	4.41E-08	ENSMUSG00000020427.11	Igfbp3
BINPACKER.28731.1	2.484	1.634	4.41E-08	no match	-
BINPACKER.5662.4	2.061	2.419	1.32E-07	ENSMUSG00000019997.11	Ctgf
BINPACKER.87639.1	2.682	0.345	1.96E-07	ENSMUSG00000001768.15	Rin2
BINPACKER.35470.1	2.367	1.786	1.89E-04	no match	-
BINPACKER.52106.1	2.096	-0.542	6.83E-04	no match	-

BINPACKER.3957.3	6.309	1.579	1.02E-03	ENSMUSG00000019988.6	Nedd1
BINPACKER.116235.1	2.212	0.301	3.94E-03	no match	-
BINPACKER.4449.4	3.428	-0.538	6.74E-03	ENSMUSG00000005150.16	Wdr83
BINPACKER.28.2	4.183	2.295	1.05E-02	ENSMUSG00000075706.10	Gpx4
BINPACKER.56553.1	1.472	0.172	1.46E-02	no match	-
BINPACKER.93518.1	1.711	-0.793	1.57E-02	no match	-
BINPACKER.11512.1	1.187	3.654	1.70E-02	ENSMUSG00000031591.14	Asah1
BINPACKER.66588.1	1.851	-0.347	1.71E-02	no match	-
BINPACKER.42718.1	1.542	0.507	2.06E-02	ENSMUSG00000030790.15	Adm
BINPACKER.49203.1	1.639	-0.035	2.44E-02	no match	-
BINPACKER.147548.1	1.744	-0.007	2.99E-02	ENSMUSG00000042757.15	Tmem108
BINPACKER.23756.2	1.265	3.468	3.01E-02	ENSMUSG00000022061.8	Nkx3-1
BINPACKER.12709.1	3.906	2.611	3.01E-02	ENSMUSG00000028639.14	Ybx1
BINPACKER.5280.2	3.874	0.257	3.76E-02	ENSMUSG00000074582.10	Arfgef2
BINPACKER.58702.1	1.780	-0.500	4.93E-02	no match	-

485   Table 5: DESeq2 determined significantly differentially expressed genes by treatment group in *P.*

486   *eremicus* testes. Of the 20 DGE with a $-1 < \log_2$ fold change > 1, 16 were significantly more highly

487   expressed in WET mice (High in WET) and four were more highly expressed in DRY mice (High in

488   DRY).

Ensembl ID	baseMean	log$_2$FC	p-adjusted	Gene ID	HIGH
ENSMUSG00000054200.6	8.77721485	-2.2659204	1.24E-27	Ffar4	WET
ENSMUSG00000026435.15	38.7630267	-2.2184407	1.16E-42	Slc45a3	WET
ENSMUSG00000079019.2	24.7158409	-1.6454793	4.55E-13	Insl3	WET
ENSMUSG00000031170.14	42.2322119	-1.6434261	6.64E-15	Slc38a5	WET
ENSMUSG00000038463.8	16.2605998	-1.4619721	3.55E-12	Olfml2b	WET
ENSMUSG00000030830.18	22.0478661	-1.4358002	3.41E-10	Itgal	WET
ENSMUSG00000032554.15	67.5197473	-1.3762549	7.26E-10	Trf	WET
ENSMUSG00000021253.6	31.2493344	-1.3551661	7.02E-14	Tgfb3	WET
ENSMUSG00000030972.6	13.8934534	-1.1709964	2.37E-07	Acsm5	WET
ENSMUSG00000059991.7	173.025492	-1.1528314	5.12E-11	Nptx2	WET
ENSMUSG00000046687.5	44.9527785	-1.0989949	8.31E-09	Gm5424	WET
ENSMUSG00000024125.1	101.5876	-1.0962074	9.77E-06	Sbpl	WET
ENSMUSG00000021913.7	46.5401886	-1.0876018	8.70E-07	Ogdhl	WET
ENSMUSG00000015702.13	27.7002506	-1.0603879	1.95E-05	Anxa9	WET
ENSMUSG00000036596.6	6.6698922	-1.0243046	9.04E-05	Cpz	WET
ENSMUSG00000025172.2	13.2622565	-1.0138171	0.00013318	Ankrd2	WET
ENSMUSG00000042757.15	14.5676529	1.00643936	0.00019556	Tmem108	DRY
ENSMUSG00000019997.11	64.49614	1.03331405	7.67E-05	Ctgf	DRY
ENSMUSG00000020427.11	92.3763518	1.56656207	4.55E-13	Igfbp3	DRY
ENSMUSG00000001768.15	12.3794312	1.72433255	8.16E-16	Rin2	DRY

491   Table 6: Functional information and chromosome (CHR) locations (*Mus musculus*) for the nine genes

492   differentially expressed across all three analyses in *P. eremicus* testes by treatment group

Gene Name	Gene ID	Gene Function	CHR	HIGH
Insulin-like 3	Insl3	testicular function and testicular development	8	WET
Free-fatty acid receptor 4	Ffar4	metabolism and inflammation	19	WET
Solute carrier family 45 member 3	Slc45a3	sugar transport	1	WET
Solute carrier family 48 member 5	Slc38a5	sodium-dependent amino acid transport	X	WET
Integrin alpha L	Itgal	lymphocyte-mediated immune responses	7	WET
Transferrin	Trf	iron transport and delivery to erythrocytes	9	WET
Ras and Rab Interactor 2	Rin2	endocytosis and membrane trafficking	2	DRY
Insulin-like growth factor binding protein 3	Igfbp3	modulates effects of insulin growth factors	11	DRY
Connective tissue growth factor	Ctgf	fibrosis and extracellular matrix formation	10	DRY

495   Figure 1: Plot of edgeR determined differentially expressed genes. The 15 significant genes are in red,

496   with positive values indicating increased expression in the DRY group, and negative values depicting

497   increased expression in the WET group.

505 Figure 2: Plot of edgeR determined differentially expressed transcripts. The 66 significant transcripts are
506 in red, with positive values indicating increased expression in the DRY group, and negative values
507 depicting increased expression in the WET group.
508
509

510
511

512   Figure 3: Plot of DESeq2 determined differentially expressed transcripts. The 215 significant transcripts

513   are in red, with positive values indicating increased expression in the DRY group, and negative values

514   depicting increased expression in the WET group.

521   Figure 4: Correlation of $\log_2$ fold change results for all Ensembl ID gene matches from DESeq2 and

522   edgeR DGE analyses (Adj-R^2 = 0.6596; F(1,14214) = 2.754x10^4; p < 2.2x10^{-16}).

Figure 5: Box plots of edgeR analyzed differences in gene expression by treatment for the nine genes

significantly differentially expressed in all three analyses. Counts per million (cpms) for both treatments

(WET and DRY) are indicated.

Figure 6: STRING diagrams of protein-protein interactions for genes significantly differentially

expressed (highly expressed) in the WET treatment group. These six genes are (a) Slc38a5, (b) Slc45a3,

(c) Insl3, (d) Ffar4 (also known as O3far1), (e) Itgal, and (f) Trf.

(a)

(b)

(c)

(d)

(e)

(f)

Figure 7: STRING diagrams of protein-protein interactions for genes significantly differentially

expressed (highly expressed) in the DRY treatment group. These three genes are (a) Rin2, (b) Igfbp3,

and (c) Ctgf.

(a)

(b)

(c)

Supplemental 1: Testes read data statistics, including sample identification (Mouse ID), number of reads

(# Reads), percent reads mapped to transcriptome (% Mapping), and treatment group (TRT). Mouse ID

335T* is the dataset which was used to assemble the testes transcriptome; therefore, these reads were

not used for the differential expression analysis.

Mouse ID	# Reads	% Mapping	TRT
335T*	*45759114*	*85.46*	*wet*
3333T	15135923	82.56	Wet
2322T	12584407	82.37	Dry
382T	14305186	83.87	Dry
381T	14178847	83.23	Wet
376T	14588175	82.56	Dry
366T	13641731	82.95	Wet
349T	17289781	85.93	Wet
209T	11724617	84.02	Dry
265T	11536510	84.17	Dry
383T	13250034	81.46	Dry
384T	12152820	82.75	Dry
102T	11131941	84.84	Wet
400T	13259393	83.98	Wet
1357T	20603232	82.32	Wet
1358T	12240814	86.58	Wet
1359T	11144962	85.54	Wet
13T	11075885	83.55	Dry
343T	9423867	83.58	Dry
344T	17146134	85.36	Wet
355T	13948415	85.21	Wet
888T	18890387	86.52	Dry
999T	15213425	87.02	Dry

566   Supplemental 2: Significantly differentially expressed genes identified in the three analyses (DGE in

567   edgeR, DTE in edgeR, and DGE in DESeq2) by treatment group in *P. eremeicus* testes. Of the 34

568   different genes which were more highly expressed in WET mice, six were significant across all three

569   analyses (Gene IDs are italicized). Of the 17 genes which were more highly expressed in DRY mice,

570   three were significant across all three analyses (Gene IDs are italicized).

HIGH: WET			
Gene ID	DGE edgeR	DTE edgeR	DGE DESeq2
Insl3	x	X	x
Ffar4	x	X	x
Slc45a3	x	X	x
Slc38a5	x	X	x
Itgal	x	X	x
Trf	x	X	x
Slit1	x	X	
Cpz		X	x
Tgfb3		X	x
Ces1g		X	
Ankrd2		X	x
Nvl		X	
Ogdhl		X	x
Pfkfb4		X	
Slc33a1		X	
Anxa9		X	x
Ddb2		X	
St3gal1		X	
Acsm5		X	x
Cyp17a1		X	
Olfml2b		X	x
Pf4		X	
Nptx2		X	x

Gene ID	DGE edgeR	DTE edgeR	DGE DESeq2
Dnah6		X	
Sbpl		X	x
Adcy6		X	
Gm5424		X	x
Mbp		X	
Fbxo2		X	
Mycl		X	
Eci1		X	
Capn12		X	
Col6a1		X	
Gpr55		X	
HIGH: DRY			
Gene ID	**DGE edgeR**	**DTE edgeR**	**DGE DESeq2**
Rin2	x	X	x
Igfbp3	x	X	x
Ctgf	x	X	x
Cyp2e1	x		
Fmo2	x		
Tnfrsf21	x		
Cyp2f2	x		
Dennd2d	x		
Nedd1		X	
Wdr83		X	
Gpx4		X	
Asah1		X	
Adm		X	
Tmem108		X	x
Nkx3-1		X	
Ybx1		X	
Arfgef2		X	

DropBox Files (will be submitted to Dryad upon acceptance):

Optimized final un-annotated transcriptome (good.BINPACKER.cdhit.fasta)

Annotated transcriptome (good.BINPACKER.cdhit.fasta.dammit.fasta)

Dammit gff3 file of annotation (good.BINPACKER.cdhit.fasta.dammit.gff3)

Salmon folder including salmon quant outputs for 22 individuals (salmon)

Salmon merged quant file (NEWmergedcounts.txt)

Gene ID by Transcript ID matrix (NEWESTfinalMUS.txt)

Transcripts without matches from edgeR DTE analysis (DTEno-matchBLASTnSequences.md)

Literature Cited:

Ahmed A, Tiwari RJ, Mishra GK, Jena B, Dar MA, Bhat AA. 2015. Effect of environmental heat stress
on reproductive performance of dairy cows- a review. International Journal of Livestock Research 5(4):
10-18. doi: 10.5455/ijlr.20150421122704.

Akerlof E, Jornvall H, Slotte H, Pousette A. 1991. Identification of apolipoprotein A1 and
immunoglobulin as components of a serum complex that mediates activation of human sperm motility.
Biochemistry 30: 8986-8990. Doi:10.1021/bi00101a011.

Albelda SM, Smith CW, Ward PA. 1994. Adhesion molecules and inflammatory injury. The FASEB
Journal 8(8): 504-512. ***Doi N/A***

Allison MB, Myers MG Jr. 2014. 20 years of leptin: connecting leptin signaling to biological function.
Journal of Endocrinology 223(1): T25-T35. Doi: 10.1530/JOE-14-0404.

Altschul SF, Gish W, Miller W, Myers EW, Lipman DJ. 1990. Basic local alignment search tool.
Journal of Molecular Biology 215(3): 403-10. doi:10.1016/S0022-2836(05)80360-2.

Anuka E, Gal M, Stocco DM, Orly J. 2013. Expression and roles of steroidogenic acute regulatory
(StAR) protein in 'non-classical', extra-adrenal and extra-gonadal cells and tissues. Molecular and
Cellular Endocrinology 371(1-2):47-61. doi: 10.1016/j.mce.2013.02.003.

Asres A, Amha N. 2014. Physiological adaptation of animals to the change of environment: a review.
Journal of Biology, Agriculture and Healthcare 4(25): 146-151.
http://www.iiste.org/Journals/index.php/JBAH/article/view/17387

Bathgate RAD, Zhang S, Hughes RA, Rosengren KJ, Wade JD. 2012. The Structural Determinants of
Insulin-Like Peptide 3 Activity. Frontiers in Endocrinology 3:11 (pp.1-10).
doi:10.3389/fendo.2012.00011.

Bathgate RAD, Halls ML, van der Westhuizen ET, Callander GE, Kocan M, Summers RJ. 2013.
Relaxin family peptides and their receptors. Physiological Reviews 93(1): 405-480. doi:
10.1152/physrev.00001.2012

Bedford NL, Hoekstra HE. 2015. The Natural History of Model Organisms: *Peromyscus* mice as a
model for studying natural variation. eLife 4: e06813. doi:10.7554/eLife.06813.

Benjamini Y, Hochberg Y. 1995. Controlling the false discovery rate: a practical and powerful approach
to multiple testing. Journal of the Royal Statistical Society: Series B (Methodological) 57(1): 289-30.
Stable URL: http://www.jstor.org/stable/2346101

Bolger DT, Patten MA, Bostock DC. 2005. Avian reproductive failure in response to an extreme
climatic event. Oecologia 142(3):398-406. Doi: 10.1007/s0042-004-1734-9.

Boone M, Deen PMT. 2008. Physiology and pathophysiology of the vasopressin-regulated renal water
reabsorption. Pflugers Archiv 456(6): 1005-1024. Doi: 10.1007/s00424-008-0498-1

Bose TO, Colpitts SL, Pham Q-M, Puddington L, Lefrancois L. 2014. CD11a is essential for normal
development of hematopoietic intermediates. Journal of Immunology 193: 2863-2872. Doi:
10.4049/jimmunol.1301820.

César-Razquin A, Snijder B, Frappier-Brinton T, Isserlin R, Gyimesi G, Bai X, Reithmeier RA,
Hepworth D, Hediger MA, Edwards AM, Superti-Furga G. 2015. A call for systematic research on
solute carriers. Cell 162 (3): 478-487. doi:10.1016/j.cell.2015.07.022.

Chang ACM, Cha J, Koentgen F, Reddel RR. 2005. The murine stanniocalcin 1 gene is not essential for
growth and development. Molecular and Cellular Biology 25(3): 10604-10610. Doi:
10.1128/MCB.25.23.10604-10610.2005.

Christenson LK, Strauss JF III. 2001. Steroidogenic acute regulatory protein: an update on its regulation
and mechanism of action. Archives of Medical Research 32(6):576-586.
http://dx.doi.org/10.1016/S0188-4409(01)00338-1

Dantzler WH. 1982. Renal Adaptations of Desert Vertebrates. BioScience 32(2):108-113. Doi:
10.2307/1308563

Dewey GC, Elias H, Appel KR. 1966. Stereology of renal corpuscles of desert and swamp deermice.
Nephron 3(6): 352-365. Doi: 10.1159/000179552.

Diaz GB, Ojeda RA, Rezende EL. 2006. Renal morphology, phylogenetic history and desert adaptation

of South American hystricognath rodents. Functional Ecology 20: 609–620. Doi: 10.1111/j.1365-

2435.2006.01144.x

Doherty GJ, McMahon HT. 2009. Mechanisms of Endocytosis. Annual Review of Biochemistry 78:

857-902. Doi: 10.1146/annurev.biochem.78.081307.110540

Evans MEK, Hearn DJ, Theiss KE, Cranston K, Holsinger KE, Donoghue MJ. 2010. Extreme

environments select for reproductive assurance: evidence from evening primroses (Oenothera). New

Phytologist 191:555-563. Doi: 10.1111/j.1469-8137.2011.03697.x.

Friedman J. 2014. 20 Years of leptin: lepttin at 20: an overview. Journal of Endocrinology 223:1 T1-T8.

Doi: 10.1530/JOE-14-0405.

Froussios K, Schurch NJ, Mackinnon K, Gierlinski M, Duc C, Simpson GG, Barton GJ. 2016. How well

do RNA-Seq differential gene expression tools perform in higher eukaryotes? Doi: 10.1101/090753.

Fu L, Niu B, Zhu Z, Wu S, Li W. 2012. CD-HIT: accelerated for clustering the next generation sequencing

data. Bioinformatics 28(23): 3150-3152. Doi: 10.1093/bioinformatics/bts565

Fuller A, Hetem RS, Maloney SK, Mitchell D. 2014. Adaptation to Heat and Water Shortage in Large, Arid-

Zone Mammals. Physiology 29(3): 159-167. Doi: 10.1152/physiol.00049.2013

Geiser F. in *Aestivation: Molecular and Physiological Aspects* (Volume 49 in the series Progress in

Molecular and Subcellular Biology, Springer, Berlin, 2010, Navas A, Carvalho C, Eduardo J.) 95-111.

Doi:10.1007/978-3-642-02421-4_5

Gierliński M, Cole C, Schofield P, Schurch NJ, Sherstnev A, Singh V, Wrobel N, Gharbi K, Simpson G,

OwenHughes T, Blaxter M, Barton GJ. 2015. Statistical models for RNA-seq data derived from a two-

condition 48-replicate experiment. Bioinformatics 31(22): 3625-3630. Doi:

10.1093/bioionformatics/btv425.

Gkouvatsos K, Papanikolaou G, Pantopoulos K. 2012. Regulation of iron transport and the role of

transferrin. Biochemica Biophysica Acta (BBA) – General Subjects 1820(3): 188-202.

Doi:10.1016/j.bhagen.2011.10.013

Hediger MA, Romero MF, Peng J-B, Rolfs A, Takanaga H, Bruford EA. 2004. The ABCs of solute

carriers: physiological, pathological and therapeutic implications of human membrane transport

proteins: introduction. Pffugers Archiv: European Journal of Physiology 447(5): 465-548.

doi:10.1007/s00424-003-1192-y.

Hediger MA, Clemencon B, Burrier RE, Bruford EA. 2013. The ABCs of membrane transporters in

health and disease (SLC series): introduction. Molecular Aspects of Medicine 34(2-3): 95-107. doi:

10.1016/j.mam.2012.12.009.

Hill RW, Veghte JH. 1976. Jackrabbit ears: surface temperatures and vascular responses. Science

194(4263):436-8. Doi: 10.1126/science.982027

Hoekstra HE, Hirschmann RJ, Bundey RA, Insel PA, Crossland JP. 2006. A single amino acid mutation

contributes to adaptive beach mouse color patterns. Science 313:101–104 DOI

10.1126/science.1126121.

Hohenlohe PA, Bassham S, Etter PD, Stiffler N, Johnson EA, Cresko WA. 2010. Population genomics

of parallel adaptation in threespine stickleback using sequenced RAD Tags. PLoS Genetics

6(2):e1000862 DOI 10.1371/journal.pgen.1000862.

Jiang H, Lei R, Ding S-W, Zhu S. 2014. Skewer: a fast and accurate adapter trimmer for next-generation

sequencing paired-end reads. BMC Bioinformatics 15:182 (pp. 1-12). doi: 10.1186/1471-2015-15-182

Jones FC, Grabherr MG, Chan YF, Russell P, Mauceli E, Johnson J, Swofford R, Pirun M, Zody MC,

White S, Birney E, Searle S, Schmutz J, Grimwood J, Dickson MC, Myers RM, Miller CT, Summers

BR, Knecht AK, Brady SD, Zhang H, Pollen AA, Howes T, Amemiya C, Broad Institute Genome

Sequencing Platform & Whole Assembly Team, Lander ES, Di Palma F, Lindblad-Toh K, Kingsley

DM. 2012. The genomic basis of adaptive evolution in threespine sticklebacks. Nature 484:55–61. DOI

10.1038/nature10944.

Jones JI, Clemmons DR. 2008. Insuline-like growth factors and their biding proteins: biological actions.

Endocrine Reviews 16(1):   doi: 10.1210/edrv-16-1-3

Kalabukhov NI. 1960. Comparative ecology of hibernating animals. Bulletin of the Museum of

Comparative Zoology at Harvard 124: 45-74. Doi: N/A

Kawamura K, Kumagai J, Sudo S, Chun S-Y, Pisarska M, Morita H, Toppari J, Fu P, Wade JD,

Bathgate RAD, Hsueh AJW. 2004. Paracrine regulation of mammalian oocyte maturation and male

germ cell survival. Proceedings of the National Academy of Sciences 101(19): 7323-7328. Doi:

10.1073.pnas.0307061101.

Kelt DA. 2011. Comparative ecology of desert small mammals: a selective review of the past 30 years.

Journal of Mammalogy, 92(6):1158–1178. doi: http://dx.doi.org/10.1644/10-MAMM-S-238.1

King, John A., (editor). Biology of *Peromyscus* (Rodentia). Special Publication No. 2, The American

Society of Mammologists, xiii + 594 pp, 1968. Doi: 10.2307/1378817

Kordonowy LK, MacManes MD. 2016. Characterization of a male reproductive transcriptome for

Peromyscus eremicus (cactus mouse). PeerJ 4:e2617(pp.1-23). Doi: 10.7717/peerj.2617.

Kordonowy L, Lombardo K, Green H, LaCourse S, Bolton E, MacManes M. 2016. Physiological and

biochemical changes associated with experimental dehydration in the desert adapted cactus mouse, Peromyscus

eremicus. BioRxiv. doi: https://doi.org/10.1101/047704

Kwon T-H, Frokiaer J, Nielsen S. 2013. Regulation of aquaporin-2 in the kidney: A molecular

mechanism of body-water homeostasis. Kidney Research Clinical Practice 32(3): 96-1023. Doi:

10.1016/j.krcp.2013.07.005.

Leijonhufvud P, Akerlof E, Pousette A. 1997. Structure of sperm activating protein. Molecular Human

Reproduction 3(3): 249-253. Doi:10.1093/molehr/3.3.249.

Le Roith D. 1997. Insulin-like growth factors. The New England Journal of Medicine 336(9): 633-640.

Doi: 10.1056/NEJM199702273360907

Li W, Godzik A. 2006. Cd-hit: a fast program for clustering and comparing large sets of protein or

nucleotide sequences. Bioinformatics 22(13):1658-1659. Doi: 10.1093/bioinformatics/btl158.

Liu J, Li G, Chang Z, Yu T, Liu B McMullen R, Chen P, Huang X. 2016. BinPacker: packing-based *de*

novo transcriptome assembly from RNA-seq data. PLos Computational Biology 12(2):e1004772 (pp.1-

15). Doi:10.1371/journal.pcbi.1004772.

Lorenzo FR, Huff C, Myllymäki M, Olenchock B, Swierczek S, Tashi T, Gordeuk V, Wuren T, Ri-Li G,

McClain DA, Khan TM, Koul PA, Guchhait P, Salama ME, Xing J, Semenza GL, Liberzon E, Wilson

A, Simonson TS, Jorde LB, Kaelin Jr WG, Koivunen P, Prchal JT. 2014. A genetic mechanism for

Tibetan high-altitude adaptation. Nature Genetics 46(9):951–956 DOI 10.1038/ng.3067.

Love MI, Huber W, and Anders S. 2014. Moderated estimation of fold change and dispersion for RNA-

seq data with DESeq2. Genome Biology 15:550 (pp. 1-21). Doi: 10.1186/s13059-014-0550-8.

MacManes MD. 2014. On the optimal trimming of high-througput mRNA sequence data. Frontiers in

Genetics 5:13 (pp.1-7). doi.org/10.3389/fgene.2014.00013.

MacManes MD. 2016. Establishing evidenced-based best practice for the de novo assembly and evaluation of

transcriptomes from non-model organisms. bioRxiv. doi: 10.1101/035642.

MacManes MD, Eisen MB. 2014. Characterization of the transcriptome, nucleotide sequence

polymorphism, and natural selection in the desert adapted mouse *Peromyscus eremicus*. PeerJ 2: e642.

doi.org/10.7717/peerj.642.

MacMillen RE, Garland T Jr. in *Advances in the Study of Peromyscus (Rodentia)* (Texas Tech

University Press, Lubbock, 1989, Kirkland LG Jr, Layne JN) 143-168.

MacMillen RE, Hinds DS. 1983. Water regulatory efficiency in heteromyid rodents: A model and its

application. Ecology 64(1): 152-164. DOI: 10.2307/1937337

Madden T. 2002 Oct 9 [Updated 2003 Aug 13]. The BLAST Sequence Analysis Tool. In: McEntyre J,

Ostell J, editors. The NCBI Handbook [Internet]. Bethesda (MD): National Center for Biotechnology

Information (US); 2002-. Chapter 16. Available from: http://www.ncbi.nlm.nih.gov/books/NBK21097/

Marra NJ, Eo SH, Hale MC, Waser PM, DeWoody JA. 2012. *A priori* and *a posteriori* approaches for

finding genes of evolutionary interest in non-model species: Osmoregulatory genes in the kidney

transcriptome of the desert rodent *Dipodomys spectabilis* (banner-tailed kangaroo rat). Comparative

Biochemistry and Physiology, Part D: Genomics Proteomics 7(4): 328-339. doi:

10.1016/j.cbd.2012.07.001.

Marra NJ, Romero A, DeWoody A. 2014. Natural selection and the genetic basis of osmoregulation in

heteromyid rodents as revealed by RNA-seq. Molecular Ecology 23(11): 2699-2711. doi:

10.1111/mec.12764.

Martin K, Wiebe KL. 2004. Coping mechanisms of alpine and arctic breeding birds: extreme weather

and limitations to reproductive resilience. Integrative and Comparative Biology 44(2):177-185. Doi:

10.1093/icb/44.2.177.

McCarthy DJ, Chen Y, Smyth GK. 2012. Differential expression analysis of multifactor RNA-Seq

experiments with respect to biological variation. Nucleic Acids Research 40(10): 4288-4297. Doi:

10.1093/nar/gks042

Moniri NH. 2016. Free-fatty acid receptor-4 (GPR120): Cellular and molecular function and its role in

metabolic disorders. Biochemical Pharmacology 110-111:1-15. doi:10.1016/j.bcp.2016.01.021

Moussad EE-DA, Brigstock DR. 2000. Connective Tissue Growth Factor: What's in a Name?

Molecular Genetics and Metabolism 71: 276-292. Doi: 10.1006/mgme.2000.3059.

Munshi-South J, Richardson JL. 2016 *Peromyscus* transcriptomics: understanding adaptation and gene

expression plasticity within and between species of deer mice. Seminars in Cell & Developmental

Biology *in press* (pp.1-9). doi: 10.1016/j.semcdb.2016.08.011

Nargund VH. 2015. Effects of psychological stress on male fertility. Nature Reviews Urology 12: 373-

382. Doi: 10.1038/nrurol.2015.112

Oh J, Woo JM, Choi E, Kim T, Cho BN, Park ZY, Kim YC, Kim DH, Cho C. 2005. Molecular,

biochemical, and cellular characterization of epididymal ADAMs, ADAM7 and ADAM28. Biochemical

and Biophysical Research Communications 331(4):1374-1383. doi:10.1016/j.bbrc.2005.04.067

Pan D. 2010. The hippo signaling pathway in development and cancer. Developmental Cell 19(4): 491-

505. Doi: 10.1016/j.devcel.2010.09.011.

Patro R, Duggal G, Kingsford C. 2015. Salmon: Accurate, Versatile and Ultrafast Quantification from

RNA-seq Data using Lightweight-Alignment. bioRxiv 021592; doi: http://dx.doi.org/10.1101/021592.

R Core Team. 2016. R: A Language and Environment for Statistical Computing. R Foundation for

Statistical Computing. Vienna, Austria. https://www.R-project.org

Robinson MD, McCarthy DJ, Smyth GK. 2010. edgeR: a Bioconductor package for differential

expression analysis of digital gene expression data. Bioinformatics 26(1): 139-140. Doi:

10.1093/bioninformatics.btp616.

Schiöth HB, Roshanbin S, Hägglund MG, Fredriksson R. 2013. Evolutionary origin of amino acid

transporter families SLC32, SLC36 and SLC38 and physiological, pathological and therapeutic aspects.

Molecular Aspects of Medicine 34(2-3): 571-585. Doi:10.1016/j.mam.2012.07.012

Schmidt-Nielsen K. in *Desert Animals: Physiological Problems of Heat and Water* (Oxford University

Press, New York, 1964, Schmidt-Nielsen K.) 129-138.

Schmidt-Nielsen B, Schmidt-Nielsen K, Brokaw A, Schneiderman H. 1948. Water conservation in

desert rodents. Journal of Cellular Physiology 32(3): 331-360. doi: 10.1002/jcp.1030320306.

Schmidt-Nielsen K, Schmidt-Nielsen B. 1952. Water metabolism of desert mammals 1. Physiological

Reviews 32(2): 135-166. PMID: 1492697.

Schurch NJ, Schofield P, Gierliński M, Cole C, Sherstnev A, Singh V, Wrobel N, Gharbi K, Simpson

GG, Owen-Hughes T, Blaxter M, Barton GJ. 2016. How many biological replicates are needed in an

RNA-seq experiment and which differential expression tool should you use? RNA 22(6): 839-

851.doi: 10.1261/rna.053959.115

Scott C. 2016. dammit: an open and accessible de novo transcriptome annotator.

www.camillescott.org/dammit

Sikes RS, Animal Care and Use Committee of the American Society of Mammalogists. 2016. 2016

Guidelines of the American society of Mammalogists for the use of wild mammals in research and

education. Journal of Mammalogy 97(3): 663-688. doi: 10.1093/jmammal/gyw078

Simão FA, Waterhouse RM, Ioannidis P, Kriventseva EV, Zdobnov EM. 2015. BUSCO: assessing

genome assembly and annotation completeness with single-copy orthologs. Bioinformatics 31(19):

3210-3212. doi: 10.1093/bioinformatics/btv351.

Smith-Unna R, Boursnell C, Patro R, Hibberd JM, Kelley S. 2016. TransRate: reference free quality
assessment of de-novo transcriptome assemblies. Genome Research 26: 1134-1144. DOI:
10.1101/gr.196469.115

Snel B, Lehmann G, Bork P, Huynen MA. 2000. STRING: a web-server to retrieve and display the
repeatedly occurring neighbourhood of a gene. Nucleic Acids Research: 28(18):3442-3444.
doi: 10.1093/nar/28.18.3442.

Somero GN. 2010. The physiology of climate change: how potentials for acclimatization and genetic
adaptation will determine 'winners' and losers'. Journal of Experimental Biology 213:912-920. Doi:
10/1242/jeb037473.
798

Soneson C, Love MI, Robinson MD. 2016. Differential analyses for RNA-seq: transcript-level estimates
improve gene-level inferences. F1000Reserach 4:1521 (pp 1-19). Doi: 10.12688/f1000research.7563.2

Song L. Florea L. 2015. Rcorrector: efficient and accurate error correction for Illumina RNA-seq reads.
Gigascience 4:48 (pp.1-8) doi: 10.1186/s13742-015-0089-y.

Stenmark H, Olkkonen VM. 2001. The Rab GTPase family. Genome Biology 2(5):pp.1-7. Doi:
10.1186/gb-2001-2-5-reviews3007

Stevens RD, Tello JS. 2009. Micro- and macrohabitat associations in Mojave desert rodent
communities. Journal of Mammalogy 90(2):388-403. Doi: 10.1644/08-MAMM-A-141.1

Storz JF, Runck AM, Moriyama H, Weber RE, Fago A. 2010. Genetic differences in hemoglobin
function between highland and lowland deer mice. Journal of Experimental Biology 213(15):2565–2574
DOI 10.1242/jeb.042598.

Szklarczyk D, Franceschini A, Wyder S, Forslund K, Heller D, Huerta-Cepas J, Simonovic M, Roth A,
Santos A, Tsafou KP, Kuhn M, Bork P, Jensen LJ, von Mering C. 2015. STRING v10: protein-protein
interaction networks, integrated over the tree of life. Nucleic Acids Research 43: D447-452. doi:
10.1093/nar/gku1003.

Urity VB, Issaian T, Braun EJ, Dantzler WH, Pannabecker TL. 2012. Architecture of kangaroo rat inner
medulla: segmentation of descending thin limb of Henle's loop. American Journal of Physiology-
Regulatory Integrative and Comparative Physiology 302(6): R720-R726.
doi:10.1152/ajpregu.00549.2011.

Veal R, Caire W. 1979. *Peromyscus eremicus*. Mammalian Species 118:1-6. Available at
http://www.science.smith.edu/msi/pdf/i0076-3519-118-01-0001.pdf.

Vimtrup BJ, Schmidt-Nielsen B. 1952. The histology of the kidney of kangaroo rats. The Anatomical
Record 114(4): 515-528. doi:10.1002/ar.1091140402.

Vitavska O, Wieczorek H. 2013. The SLC45 gene family of putative sugar transporters. Molecular
Aspects of Medicine 34(2-3): 655-660. Doi:10.1016/j.mam.2012.05.014

Vorhies CT in Water Requirements of Desert Animals in the Southwest. (College of Agriculture,
University of Arizona, Tucson, Agricultural Experiment Station.) Technical Bulletin No. 107: 487-525.
http://hdl.handle.net/10150/190625

Wagner GF, Dimattia GE. 2006. The stanniocalcin family of proteins. Journal of Experimental Zoology
Part A 305A(9): 769-780. Doi: 10.1002/jez.a.313

Walsberg GE. 2000. Small mammals in hot deserts: some generalizations revisited. BioScience 50(2):
109-120. doi: 10.1641/0006-3568(2000)050[0109:SMIHDS]2.3.C

Wingfield JC. 2013. The ecology of stress: ecological processes and the ecology of stress: the impacts of
abiotic environmental factors. Functional Ecology 27:37-44. Doi: 10/1111/1365-2435.12039

Wingfield JC, Kelley JP, Angelier F. 2011. What are extreme environmental conditions and how do
organisms cope with them? Current Zoologist 57(3):373-374. Doi: 10/1093/czoolo/57.3.363

Wingfield JC, Sapolsky RM. 2003. Reproduction and resistance to stress: when and how. Journal of
Neuroendocrinology 15:711-724. doi/10.1046/j.1365-2826.2003.01033.x

Xiao X, Mruk DD, Cheng CY. 2013. Intercellular adhesion molecules (ICAMs) and spermatogenesis.
Human Reproduction Update 19(2): 167-186. doi:10.1093/humupd/dms049.