Stem cell transcriptional profiles from mouse subspecies

reveal *cis*-regulatory evolution at translation genes

Noah M. Simon[1,2,3], Yujin Kim[4,5], Diana M. Bautista[6], James R. Dutton[4,5],

Rachel B. Brem[2,3]*

[1]Biology of Aging Doctoral Program, Leonard Davis School of Gerontology,

University of Southern California, Los Angeles, CA 90089

[2]Buck Institute for Research on Aging, Novato, CA 94945, USA

[3]Department of Plant and Microbial Biology, University of California, Berkeley,

Berkeley CA 94720, USA

[4]Stem Cell Institute, University of Minnesota, Minneapolis, MN 55455, USA

[5]Department of Genetics, Cell Biology, and Development, University of

Minnesota, Minneapolis, MN 55455, USA

[6]Department of Molecular and Cell Biology, University of California, Berkeley,

Berkeley CA 94720

*Correspondence to rbrem@berkeley.edu

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Abstract

A key goal of evolutionary genomics is to harness molecular data to draw inferences about selective forces that have acted on genomes. The field progresses in large part through the development of advanced molecular-evolution analysis methods. Here we explored the intersection between classical sequence-based tests for selection and an empirical expression-based approach, using stem cells from *Mus musculus* subspecies as a model. Using a test of directional, *cis*-regulatory evolution across genes in pathways, we discovered a unique program of induction of translation genes in stem cells of the Southeast Asian mouse *M. m. castaneus* relative to its sister taxa. As a complement, we used sequence analyses to find population-genomic signatures of selection in *M. m. castaneus,* at the upstream regions of the translation genes, including at transcription factor binding sites. We interpret our data under a model of changes in lineage-specific pressures across *Mus musculus* in stem cells with high translational capacity. Together, our findings underscore the rigor of integrating expression and sequence-based methods to generate hypotheses about evolutionary events from long ago.

Introduction

The central objective of molecular-evolution research is to draw inferences about changing selective pressures between lineages, based on clues from omics data. Since its inception, the field has relied in large part on model-fitting methods using DNA sequence (Kreitman 2000) and gene expression (Price et al. 2022). Complementing the latter, empirical genome ranking/bootstrapping approaches have emerged in the more recent literature (Ferguson and Chang, 2020), for which methods development and refinement remain an active area of research (Berg *et al.*, 2019; Sohail *et al.*, 2019; Johri *et al.*, 2020; Price *et al.*, 2022).

One clear-cut empirical molecular-evolution strategy (Fraser *et al.*, 2010, 2011; Bullard *et al.*, 2010; Martin *et al.*, 2012; York *et al.*, 2018; Gokhman *et al.*, 2021) takes as input measurements of *cis*-regulatory variation from expression profiles. The test identifies cases in which, among the unlinked genes of a pathway subject to *cis*-regulatory change, alleles in one taxon tend to drive expression mostly up, or mostly down, relative to another taxon. This pattern of independent genetic variants with similar effects at similar genes is unlikely under neutral expectations (Orr, 1998). It thus serves as a suggestive signature of changes in selective pressure on the pathway between lineages. At this point, rigorous evolutionary inference requires additional follow-up, including sequence-based tests to distinguish between positive and relaxed

selection as the driver of expression divergence. A number of studies in yeast

have made this link (Martin *et al.*, 2012; Fraser *et al.*, 2012; Roop *et al.*, 2016);

in metazoans, it remains an open question whether results from expression-

based *cis*-regulatory pathway tests can be validated with sequence-based

evidence for selection (though see Mack et al. 2023).

In the current work, we set out to harness the diversity among mouse lineages

in pluripotent stem cell expression programs, to model the integration of

expression- and sequence-based tests for selection in multi-gene pathways.

We focused on the mouse *Mus musculus castaneus (M. m. castaneus)*. This

subspecies is endemic to southeast Asia and diverged 0.5-1 MYA from other

Mus musculus (Chapman and Ruddle, 1972; Sangster *et al.*, 1993). Previous

surveys have established divergence between *M. m. castaneus* and laboratory

strains in terms of gene expression (Fraser *et al.*, 2011; Xiong *et al.*, 2014;

Chappell *et al.*, 2017; Tkatchenko *et al.*, 2019; Chou *et al.*, 2022) and

phenotype (Johnson *et al.*, 1997, 2006; Beamer *et al.*, 1999; Yu *et al.*, 2007;

Koturbash *et al.*, 2011; French *et al.*, 2015; Omura *et al.*, 2015; Chappell *et al.*,

2017; Hsiao *et al.*, 2020; Chou *et al.*, 2022). Our goal was to use stem cell

transcriptomes to identify pathways subject to directional *cis*-regulatory change

between *M. m. castaneus* and laboratory mice. We earmarked one pathway

hit, a set of translation genes at which *M. m. castaneus* alleles acted in *cis* to

drive uniquely high expression, for independent validation analyses with

sequence data. At these loci, population-genomic tests revealed signals of

unique evolution in *M. m. castaneus* in the regions upstream of genes.

Together, our results establish the utility of an expression-based molecular-

evolution test with sequence-based follow-up, in a model mouse tissue.

Materials and Methods

RNA-seq data sources

For our initial screen for directional *cis*-regulatory variation in pathways, we

used transcriptional profiles of CAST/EiJ male x 129/SvImJ female F1 hybrid

embryonic stem cells (NCBI [National Center for Biotechnology Information]

accession GSE60738, samples SRR1557132, SRR1557133, SRR1557112,

and SRR1557123; Marks et al. 2015). For validation and follow-up we used

additional transcriptional profiles from reciprocal crosses of CAST/EiJ x

C57BL/6J hybrid pluripotent stem cells (NCBI accession GSE90516, samples

SRR5054337-5054348 and SRR5054353-5054364; Werner et al. 2017);

homozygous embryonic stem cells from a panel of *M. musculus* subspecies

(EBI [European Bioinformatics Institute; Sarkans et al. 2018] accession E-

MTAB-7730 [Skelly et al. 2020]); and in-house cultures of CAST/EiJ male x

129/SvImJ female F1 hybrid embryonic stem cells, and homozygous induced

pluripotent stem cells (see below).

For validation of the role of candidate transcription factors in translation

regulation in stem cells, we analyzed published transcriptional data from

trophoblast stem cells knocked out for *Eed* (Weigert *et al.*, 2023; NCBI

accession GSE198076, samples SRR18260888, SRR18260892,

SRR18260902, and SRR18260903) and E8.5 embryos knocked out for *Ehmt2*

(Auclair *et al.*, 2016; NCBI accession GSE71500, samples SRR2133432,

SRR2133433, SRR2133434, and SRR2133435), respectively.

RNA-seq read mapping

For analysis of CAST/EiJ x 129/SvImJ and CAST/EiJ x C57BL/6J hybrid stem

cell transcriptomes from (Marks *et al.*, 2015), we downloaded raw reads and

mapped with the STAR aligner (Dobin et al 2013) to a concatenated genome

file containing chromosomal sequences from both parent strains (CAST/EiJ

and either 129/SvImJ or C57BL/6J).

For analysis of transcriptomes of homozygous CAST/EiJ and C57BL/6J stem

cells cultured in-house (see below), we downloaded raw reads and mapped

them to the corresponding reference genome for the respective strain

(CAST/EiJ or GRCm38). For validation of the role of candidate transcription

factors in translation regulation in stem cells, we mapped RNA-seq from *Ehmt2*

knockout embryos (NCBI accession GSE71500; Auclair *et al.*, 2016) and their

corresponding wild-type to the GRCm38 reference, and we mapped RNA-seq

from *Eed* knockout stem cells and their corresponding wild-type to the CD-1

reference genome (NCBI accession GSE209815; Jung *et al.*, 2023).

RNA-seq normalization and quantification

Reads that mapped ambiguously to more than one locus were discarded.

Read counts were generated during the STAR alignment step using the '--

quantMode GeneCounts' option. For analysis of transcriptomes of

homozygous stem cells from (Skelly *et al.*, 2020) (genotypes C57BL/6J, A/J,

129S1/SvImJ, NZO/HILtJ, NOD/ShiLtJ, WSB/EiJ, CAST/EiJ, and PWD/PhJ),

we downloaded mapped read counts from EBI's ArrayExpress database

(Sarkans *et al.*, 2018). For each data set in turn, normalized (TPM, transcripts

per million) counts were generated by dividing read counts per gene by

transcript length (using annotations from the Ensembl database, build 102) and

then dividing by library size. Normalized read counts for hybrids and

homozygous strains are reported in Table S1 and S2, respectively.

Hybrid RNA-seq mapping quality control

To eliminate potential artifacts from allele-specific mapping errors in hybrid

RNA-seq analyses, we performed a simulated RNA-seq experiment using the

Polyester package in R (Frazee *et al.*, 2015) as follows. For the CAST x 129

hybrid stem cell transcriptome from (Marks *et al.*, 2015), we generated two

replicates of simulated reads from the hybrid genome with ~200 reads for each

annotated transcript. These simulated reads were then mapped back to the

genome with STAR as above. For a given gene with called orthologs in the

CAST and 129 genomes, for each allele in turn we tabulated the ratio between

the number of successfully mapped simulated reads and the number of

simulated reads that went into the mapping, as a report of the extent of artifact-

free mapping. We converted each such ratio to a percentage; we then took the

absolute value of the difference between the ratio for the 129 and CAST allele

as a report of the difference in mapping fidelity between them. We filtered out

genes for which the latter parameter exceeded 5%, a total of 1,852 genes of

10,380 initial homologs in the data from (Marks *et al.*, 2015) (Table S3). We

repeated this procedure for CAST x BL6 hybrid stem cell transcriptomes from

(Werner *et al.*, 2017), resulting in exclusion of 2,886 genes out of an initial

8,887 homologs expressed in the RNA-seq data (Table S3).

In silico screen for directional _cis_-regulatory variation

For our initial screen of polygenic, directional cis-regulatory variation in

pathways, we harnessed profiles from CAST/EiJ male x 129/SvImJ female F1

hybrid pluripotent stem cells (Marks et al. 2015). We generated a list of one-to-

one orthologous genes (Table S4 between CAST/EiJ and 129/SvImJ from the

Ensembl database (build 102) using biomaRt (Durinck *et al.*, 2009). At a given

gene, we tested for differential expression between the CAST/EiJ and

129/SvImJ alleles using the reads mapping to each as input into DESeq2

(Love et al. 2014). We eliminated from further analysis genes with fewer than

10 total reads across all samples. Differential expression results are reported in

Table S5.

For each gene with differential allele-specific expression at adjusted $P < 0.05$,

we assigned a quantitative sign statistic s, equal to the $\log_2$(129 allele

expression/CAST allele expression). We assigned all genes without differential

allele-specific expression to have $s = 0$. We downloaded gene annotations in

Gene Ontology 'biological process' terms from the AmiGO database (Carbon

et al., 2009). We eliminated from further testing all terms containing fewer than

10 genes with significant differential allele-specific expression. For a given

remaining term containing n genes, we summed the s values across the

component genes to yield a summary statistic S_{true}. To evaluate significance by

resampling, we randomly sampled n genes from the total set with expression

data and summed their s values, generating a resampled summary statistic

$S_{resample}$. We carried out this calculation 10,000 times and used as a two-sided

P value the proportion of resamples in which $|S_{resample}| \geq |S_{true}|$. We corrected

for multiple testing with the Benjamini-Hochberg method. P values are reported

in Table S6. All further expression and molecular-evolution analyses focused

on genes in the top-scoring term, GO:0006412, translation.

Induced pluripotent stem cell derivation and culture

To establish C57BL/6J and CAST/EiJ induced pluripotent stem cell lines, we

used mouse embryonic fibroblasts E13.5 obtained from Jackson Laboratories

(Bar Harbor, ME, USA) as input into a stem cell derivation protocol as

previously described (Terzic *et al.*, 2016). Briefly, we used octamer-binding

transcription factor-4 (Oct4), sex-determining region y-box 2 (Sox2), and

Kruppel-like factor-4 (Klf4) as reprogramming factors, introduced using pMXs

retroviral vectors. A 129S6/SvEv embryonic stem cell line was obtained from

Millipore Sigma, (Catalog no. SCR012), Burlington, MA, USA). A CAST/EiJ x

129/SvImJ mouse embryonic stem cell line was generously provided by Joost

Gribnau (Marks *et al.*, 2015).

Stem cells were cultured on irradiated mouse embryonic fibroblasts (R and D

Systems, Minneapolis, MN, USA) in miPSC medium: knockout DMEM with 4.5

g/ L d-glucose (Gibco, Grand Island, NY, USA), 10% knockout serum

replacement (KSR) (Gibco), 10% fetal bovine serum (FBS) (HyClone, Logan,

UT, USA), 1× MEM nonessential amino acids (MEM NEAA) (Gibco), 1×

GlutaMAX (Gibco), 0.1 mM 2-mercaptoethanol (BME) (Life Technologies,

Grand Island, NY, USA), and 0.02% ESGRO-LIF (Millipore, Billerica, MA,

USA). Cells were incubated at 37°C in 5% CO_2.

RNA isolation and sequencing

RNA was extracted from undifferentiated pluripotent stem cell cultures (two

replicates each of BL6 and CAST) following feeder depletion using the

RNAqueous™-Micro Total RNA Isolation Kit (Thermo Fisher Scientific) and on-

column DNase treatment (QIAGEN, Hilden, Germany). RNA samples were

processed into mRNA libraries and sequenced on an Illumina NovaSeq 6000

Sequencing System, yielding ~20M paired-end 150 bp reads per sample.

Read-mapping was as in RNA-seq normalization and quantification, above.

Population genomic analysis

We downloaded resequencing data from wild populations of *M. m. domesticus*

(from France, Germany and Iran), *M. m. musculus* (from Afghanistan, Czech

Republic and Kazakhstan), and *M. m. castaneus* (from northwest India) (Harr

et al., 2016). VCF files were used to map SNPs from each individual mouse

onto the C57BL/6J reference genome (GRCm38) using the 'consensus'

command from bcftools software (Danecek *et al.*, 2021). Default options for

'bcftools consensus' algorithm were used, meaning that in the case of the

heterozygous loci, the reference GRCm38 allele would be used. Sequences

upstream of the transcription start site for each gene were extracted utilizing

the pybedtools Python package (Quinlan and Hall, 2010; Dale *et al.*, 2011). In

order to assess only a single transcript per gene, gffread (Pertea and Pertea,

2020) was used to extract coding sequences (CDS) from each annotated

transcript. A custom Python script was used to check whether each CDS

contained an in-frame start and stop codon, signifying a valid open reading

frame (ORF). For genes with multiple transcripts containing a valid ORF, the

longest transcript was selected. Genes without an in-frame ORF were

eliminated from further analysis.

We carried out selection tests separately in a comparison of each *M. m.*

domesticus or *M. m. musculus* population in turn against the *M. m. castaneus*

population. In each case, for a given gene, we generated an alignment of the

50kb region upstream of the gene start site and split it into ten 5kb windows. In

each window of each gene, we tabulated the number of sites harboring

polymorphisms across one or both populations, and the number of sites with

alleles fixed in each population and divergent between them (P_u and D_u

respectively); these data are reported in Table S7. For a given window, we

took the sum of D_u and, separately, the sum of P_u across the translation GO

term genes, and we calculated the ratio $R_{translation} = D_u/P_u$. To evaluate

significance by resampling, for a given window we randomly sampled 10 genes

from the total set with sequence data; calculated P_u and D_u, their sums across

genes, and their ratio $R_{resample}$ as above; and over 10,000 resamples, we used

as a one-sided P value the proportion of resamples in which $R_{resample} \geq$

$R_{translation}$. To check robustness of the signal, we also employed a leave-one-out

strategy, where we repeated the resampling test over 10,000 permutations

with each translation gene excluded in turn; these data are reported in Table

S8.

Analysis of transcription factor binding sites

In a public resource of chromatin immunoprecipitation sequencing data sets

(Kolmykov *et al.*, 2021), we tabulated binding sites for 680 transcription factors

across the mouse genome. For each factor, D_u/P_u was calculated as above for

binding sites within 50kb upstream of genes, and then separately in the

remaining non-binding site regions. These values were then used as input for a

two-factor ANOVA, testing for an interaction effect between binding site identity

and membership in the translation GO term for each factor. D_u/P_u between the

M. m. castaneus population and each *M. m. domesticus* or *M. m. musculus*

population were used as replicates in the ANOVA analysis. ANOVA results are

reported in Table S9. Multiple sequence alignment was edited for display using

Jalview software (Waterhouse *et al.*, 2009). Binding site data for the top 3 hits

from this analysis are reported in Table S10.

Results

A screen for directional *cis*-regulatory change in mouse stem cell pathways

As a first testbed for analyses of pathway *cis*-regulatory change, we used

transcriptional profiles (Marks *et al.*, 2015) of embryonic stem cells of an F1

hybrid background from a cross between two homozygous mouse strains: a

male *M. m. castaneus* (CAST/EiJ, hereafter CAST), and a female of the

129/SvImJ laboratory strain genotype (hereafter 129), which is of admixed

origin (Frazer *et al.*, 2007; Yang *et al.*, 2007, 2011). In the F1, because the two

subspecies' alleles of a given gene are in the same nucleus, any difference in

allele-specific expression between them can be attributed to genetic variation

acting in *cis*, *i.e.,* not mediated through a soluble factor (Wittkopp *et al.*, 2004).

We implemented a pipeline of allele-specific read-mapping taking account of

the potential for mapping artifacts (see Methods); and we retained for analysis

all genes exhibiting significant expression divergence between the species'

alleles (962 genes at a 0.05 *p*-value threshold). We tabulated the directional

effect for each gene—whether the CAST allele was more highly expressed

than the laboratory-strain allele, or *vice versa*—using the $\log_2$-transformed fold-

change between expression of the 129 allele and CAST allele. Then, to

formulate our test, we used as pathways groups of genes of common function,

each comprised of a Gene Ontology biological process term. For each such

group, we quantified the agreement in the direction of allelic expression

differences between species across the gene members. We evaluated

significance based on resampling (Table S6). Of the complete survey results,

one pathway showed significant signal: a cohort of genes from the GO term for

translation, at which the CAST allele was upregulated relative to that of 129 in

the hybrid at a 2-fold excess of genes relative to those with higher expression

from the 129 allele (Table S6, Figure 1A, and Figure S1A). This represented a

potential case in which selective pressures on regulation of the respective loci

had changed between the *M. m. castaneus* and laboratory-strain lineages.

M. m. castaneus *cis*-regulatory alleles drive high expression of

translation genes in stem cells

As a first verification of the trend for *cis*-regulatory alleles from *M. m. castaneus*

driving high expression of translation genes, we repeated the culture and

sequencing of CAST x 129 hybrid stem cells. The results confirmed the robust

directional imbalance in allele-specific expression among translation genes,

with the CAST allele expressed more highly across the set 2.54-fold more

often than the 129 allele (Figure 1A-B). Similarly, we analyzed the

transcriptome of hybrid stem cells from a cross between CAST and the

admixed C57BL/6J laboratory strain (hereafter BL6; Werner et al. 2017), and

observed a 2.48-fold imbalance favoring high expression by the CAST allele

among translation genes (Figure 1A and Figure S1B). Together, these data

establish that *M. m. castaneus cis*-regulatory alleles at translation genes

encode a unique activating program relative to those encoded by 129 and BL6

alleles, in stem cells.

Homozygous *M. m. castaneus* stem cells are distinguished by high

expression of translation genes

We expected that, if regulatory divergence between CAST and other lineages

at translation genes had been important for fitness in the organismal and

ecological context, it would be apparent in the context of homozygous strains,

which integrate the effects of genetic factors acting both in *cis* and in *trans*

(Signor and Nuzhdin, 2018). Consistent with this prediction, translation genes

were more highly expressed in CAST homozygous stem cells relative to those

of 129 and BL6, in a published data resource (Skelly *et al.*, 2020) and in our

own culture and sequencing (Figure 1A and 1C and Figure S1C-D). The trend

persisted in comparisons between CAST homozygous stem cells and those of

other mouse lineages (Figure 1A and Figure S1E-I). Interestingly, the tendency

for high expression of translation genes by CAST in homozygous stem cells

was of larger magnitude than we had noted in our analyses of *cis*-regulatory

variation in hybrids (Figure 1A), indicating that the latter was reinforced by a

more pervasive effect of divergence in *trans*-acting regulators. Given these

signatures of directional *cis*- and *trans*-acting variation between *Mus*

subspecies, we considered the translation gene cohort to be a compelling

candidate for molecular-evolution follow-up, to investigate the underlying

selective pressures.

ial *M. m. castaneus* harbors unique and conserved alleles in spatial

windows of translation gene upstream regions

We next turned to sequence-based tests of selection to explore the

evolutionary history of translation gene regulation between mouse lineages.

We reasoned that analysis of multiple *M. musculus* subspecies could help

polarize any sequence variation we would detect at the loci. Toward this end,

we developed an analysis pipeline for single-nucleotide variants from wild-

caught mice of the *Mus* genus (Harr *et al.*, 2016), in which we tabulated

divergence between subspecies, normalized by intra-subspecies

polymorphism, in spatial windows of upstream regions for every gene in the

genome. We then used these values as input into resampling tests for

enrichment of high values of this normalized quantity in translation genes

relative to a genomic null, representing a hallmark of positive selection

(Kreitman, 2000). In a first such approach, we classified variant sites based on

their spatial positions in upstream loci (Figure 2A). The results revealed

elevated sequence divergence between *M. m. castaneus* on the one hand and

populations of *M. m. musculus* and *M. m. domesticus* on the other (Figure 2A-

B). The signal was most marked in a window 25 kb upstream of translation

336 start among genes of the translation GO term and fell off gradually on either

337 side of this peak (Figure 2A), as expected if *M. m. castaneus* had acquired

338 unique alleles in *cis*-regulatory elements with positional preferences across this

339 gene cohort. Such a trend was consistent with the unique expression program

340 in *M. m. castaneus* that we had noted in stem cells as a product in part of *cis*-

341 regulatory change (Figure 1A).

342 **A role for transcription factors in divergently regulated translation genes**

343 As a complement to our focus on the spatial positioning of *cis*-regulatory

344 sequence changes in translation genes, we developed a separate molecular-

345 evolution approach classifying variants on the basis of binding sites by

346 transcriptional regulators (Kolmykov *et al.*, 2021). For the sites bound by a

347 given regulator, we again tested for elevated divergence between *M. m.*

348 *castaneus* and its relatives, relative to intra-subspecies polymorphism, across

349 the genes of our translation cohort (Figure 3A-B). Three regulators (Ehmt2,

350 Eed, and Msx1) exhibited robust signal in this scheme, with binding sites at

351 translation genes enriched for divergence between *M. m. castaneus* and its

352 sister lineages (Table S9 and Figure 3A-B). In *Eed* knockout stem cell and

353 *Ehmt2* knockout early embryonic transcriptomes (Auclair *et al.*, 2016; Weigert

354 *et al.*, 2023), we detected a directional impact of these factors on expression of

355 translation genes (Figure 3C and Figure S2), consistent with an inference of

356 direct regulatory function at translation genes. Interestingly, unique alleles in

357 *M. m. castaneus* in binding sites of these regulators did not fall in the window

that had been top-scoring in our spatially resolved analysis of upstream

regions (at 25 kb upstream of coding start; Figure 2 and Table S10),

suggesting a role in the latter of mechanisms beyond Ehmt2, Eed, and Msx1

binding. As such, results from the two test schemes represent independent

evidence for positive selection in distinct elements of the upstream regions of

translation genes—which together represent compelling candidates for the

mechanism of their non-neutral expression variation across mouse lineages,

and its manifestation in stem cells.

Discussion

Empirical molecular evolution tests, once developed, must be validated to

establish their rigor and utility for the field. A compelling means toward this end

is to integrate results from an emergent test strategy with those of more classic

tools, when they complement each other in support of an evolutionary

inference. In the current work, we have detected sequence signatures of

selection between mouse subspecies in a cohort of translation genes that also

exhibits directional, polygenic *cis*- and *trans*-acting regulatory variation in stem

cells. Against a backdrop of other case studies focused on *cis*-regulatory

change as detected in transcriptomes (Fraser *et al.*, 2010, 2011; Bullard *et al.*,

2010; Martin *et al.*, 2012; York *et al.*, 2018; Agoglia *et al.*, 2021; Gokhman *et*

al., 2021; Mack *et al.*, 2023; Wang *et al.*, 2024), our results represent a proof of

concept for sequence-based validation of this approach in mammals.

Our analyses implicate several transcriptional regulators in the divergence

between mouse subspecies in translation genes. The variation we have

detected in translation gene binding sites for the polycomb repressive complex

2 component Eed (Mozzetta *et al.*, 2014; Obier *et al.*, 2015; Li *et al.*, 2018), as

a candidate determinant of stem cell expression change between mice,

dovetails with the known role for Eed in development and stem cell

differentiation (Young, 2011; Obier *et al.*, 2015). The polycomb repressive

complex 2 more broadly has been shown to regulate pluripotency of stem cells

through methylation of the DNA encoding ribosomal RNAs (Zhang *et al.*,

2020); our focused study of Eed suggests that its effects in stem cells are also

mediated by other translation components, as reflected in binding at these loci

in stem cells (Li *et al.*, 2018; Oksuz *et al.*, 2018; Kaaij *et al.*, 2019; Kriz *et al.*,

2021). Likewise, the evolutionary and genomic profiles of binding sites at

translation genes for the methyltransferase Ehmt2, including their inferred

impact on expression in stem cells, dovetail with the known function of this

factor in stem cell maintenance and differentiation (Ikegami *et al.*, 2007; Leitch

et al., 2013; Boroviak *et al.*, 2014; Auclair *et al.*, 2016; Kim *et al.*, 2020) and

direct binding by Ehmt2 to translation genes in stem cells (Mozzetta *et al.*,

2014). As such, our results suggest that the known roles of Eed and Ehmt2 in

stem cell identity and differentiation may be mediated by regulation of protein

synthesis, and that evolution used these binding sites in part to tune

expression of translation factors between mouse lineages. A similar function

may ultimately also emerge for the development regulator Msx1, whose

binding sites were also top-scoring in our molecular-evolution analyses across

Mus. That said, we expect that our genomic approach affords only partial

coverage and power in discovering elements of the mechanism of mouse

subspecies divergence at translation genes, and thus that many other

contributing transcription factors likely remain to be identified.

Our finding of divergence at translation genes between mouse lineages

echoes reports of *cis*-regulatory change in the translation machinery of other

organismal systems, from yeasts (Tanay *et al.*, 2005; Hogues *et al.*, 2008; Li

and Fay, 2017; Sorrells *et al.*, 2018) to animal ancestors distributed over deep

time (Brown *et al.*, 2008). This literature leaves open the question of what

ecological forces might drive divergence of translation gene expression, and

the phenotypes that would mediate such effects. Under a model in which

ribosomal protein dosage governs the readiness of a cell to divide (Polymenis

and Aramayo, 2015), adaptive regulatory variation in translation pathways may

often reflect species-unique logic of cell growth decisions.

In metazoan stem cells in particular, translation plays a critical role in

differentiation. Inducing (Easley *et al.*, 2010) or compromising (Khajuria *et al.*,

2018) stem cell translation can drive qualitative differences in differentiation

behavior. According to current models, high expression of translation genes in

stem cells (Sampath *et al.*, 2008) sets up a poised state to enable rapid protein

production in their differentiated progeny (Gabut *et al.*, 2020). On the basis of

this tight link between translation and differentiation, it is tempting to speculate

that the variation we have seen among *Mus musculus* lineages in stem-cell

translation gene expression was driven by selection in distinct niches for

unique differentiation properties. If so, the loci we study here would act in a

complex genetic architecture of stem cell differentiation, alongside other

variants characterized across mouse lineages (Ortmann *et al.*, 2020; Skelly *et al.*, 2020). Future work will explore the phenotypic relevance of translation

al., 2020). Future work will explore the phenotypic relevance of translation

gene divergence across *Mus*, and its role in the slew of traits (Johnson *et al.*,

1997; Beamer *et al.*, 1999; Johnson *et al.*, 2006; Yu *et al.*, 2007; Koturbash *et*

al., 2011; French *et al.*, 2015; Chappell *et al.*, 2017; Hsiao *et al.*, 2020; Chou *et*

al., 2022) that distinguish *M. m. castaneus* from the rest of its genus.

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438   Author Contributions

439   NMS, RBB, DMB, and JRD conceived of the study and provided direction.

440   NMS processed the sequencing data and carried out all *in silico* analyses. JRD

441   derived the iPSC lines used for sequencing and neuron differentiation. YK

442   performed the cell cultures and RNA isolations. All authors contributed to and

443   reviewed the final manuscript.

Competing Interests

The authors declare no competing financial interests.

Data Archiving

RNA-seq data generated for this study was deposited in the GEO database (https://www.ncbi.nlm.nih.gov/geo/query/acc.cgi?acc=GSE234761).

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Figure captions

Figure 1: Directional *cis*-regulatory variation in translation gene expression in stem cells between mouse subspecies. (A) The first three columns report results from analyses of differential allele-specific expression across genes from the Gene Ontology term GO:0006412, translation, in stem cells of F1 hybrids (left columns) between CAST and either C57BL/6 (from Werner et al. 2017) or 129 ((Marks *et al.*, 2015) or this study) as indicated. The remaining columns report results from analyses of differential expression across translation genes between homozygous mouse stem cells in comparisons between CAST and either 129 (this study or (Skelly *et al.*, 2020)), C57BL/6J, A/J, NZO/HILtJ, NOD/ShiLtJ, WSB/EiJ, CAST/EiJ, or PWD/PhJ (Skelly *et al.*, 2020) as indicated. The first and second rows report the percentage of translation genes in which the CAST allele is expressed higher or lower, respectively, than the allele of the other parent, and the third row reports the ratio of these quantities. **(B)** Each point reports allele-specific expression of a translation gene in CASTx129 hybrid stem cells cultured in this study: the *x*-axis reports the $\log_2$ ratio of expression of the respective strain alleles, and the *y*-axis reports the $\log_{10}$ of the significance of the difference. Point colors report significance of differential allele-specific expression (p-adj, adjusted *p*-value). Red and black text inlays report the percentage of translation genes where expression of the respective parental allele is higher with and without filtering for significance, respectively. **(C)** Data are as in (B),

but from a comparison between homozygous CAST and 129 stem cells

cultured in this study.

Figure 2: Elevated nucleotide divergence between *M. m. castaneus* and

other subspecies in regions upstream of translation genes. (A) Each bar

reports results from a metagene analysis of nucleotide divergence between a

M. m. castaneus population from India and a *M. m. musculus* population from

Afghanistan in a 5000-bp window starting at the indicated distance upstream of

transcription start sites. The *y*-axis reports the ratio of the average number of

sites in the respective window exhibiting divergence between the subspecies

(D_u) to the average number of polymorphic sites (P_u) across genes from the

Gene Ontology term GO:0006412, translation, normalized against the

analogous quantity for all other genes in the genome. *: $p > 0.05$ from a one-

sided resampling test for elevated D_u/P_u in the translation gene cohort against

a genomic null. **(B)** Summary of the analysis of D_u/P_u between the *M. m.*

castaneus population and 6 *M. m. musuclus* and *M. m. domesticus*

populations. Normalized D_u/P_u and resampling p-values are reported for the

5kb window from 25kb to 20kb upstream for each comparison.

Figure 3: Signatures of species divergence and regulatory function of

Eed at translation genes. (A) In a given row, cells report the ratio of inter-

subspecies divergence (D_u) to within-subspecies polymorphism (P_u) at Eed

binding sites upstream of genes from the Gene Ontology term GO:0006412,

translation, or all other genes of the genome, in a comparison between *M. m.*

castaneus and the indicated population of a sister taxon. Divergence is

significantly enriched in Eed binding sites upstream of translation genes (see

Table S9). **(B)** An Eed binding footprint upstream of the translation gene Eif4e2

as an example of mouse subspecies variation. Each row reports the sequence

of one *M. m. castaneus, domesticus,* or *musculus* individual (*M. m. c.*, *M. m. d.*,

or *M. m. m.* respectively). **(C)** Data and symbols are as in Figure 1B and 1C

except that each point reports a comparison of expression of one translation

gene between *Eed* knockout and wild-type mouse stem cells (Weigert et al

2023), and red and black text inlays report the percentage of translation genes

where expression was higher in the indicated genotype with and without

filtering for significance, respectively (DE, differential expression).

Supplementary figure captions

Figure S1: Directional *cis*-regulatory variation in translation gene expression in stem cells between mouse subspecies. Data are as in Figure 1B **(A-B)** or Figure 1C **(C-I)** of the main text except that transcriptomes analyzed were as follows. **(A)** CAST x 129 hybrid stem cells (Marks *et al.*, 2015). **(B)** CAST x BL6 hybrid stem cells (Werner *et al.*, 2017). **(C)** CAST and 129 homozygous stem cells (Skelly *et al.*, 2020). **(D)** CAST and BL6 homozygous stem cells (Skelly *et al.*, 2020). **(E)** CAST and WSB/EiJ homozygous stem cells (Skelly *et al.*, 2020). **(F)** CAST and PWD/PhJ homozygous stem cells (Skelly *et al.*, 2020). **(G)** CAST and A/J homozygous stem cells (Skelly *et al.*, 2020). **(H)** CAST and NOD/ShiLtJ homozygous stem cells (Skelly *et al.*, 2020). **(I)** CAST and NZO/HILtJ homozygous stem cells (Skelly *et al.*, 2020).

Figure S2: Ehmt2 knockout induces translation genes. Data are as in Figure 3C of the main text except that transcriptomes analyzed were wild-type and Ehmt2 knockout stem cells (Auclair et al. 2016).

Supplementary table captions

Table S1. Allele-specific normalized read counts from RNA-seq of stem cells from hybrids involving CAST. Each tab reports allele-specific expression measurements in one genotype of hybrid embryonic stem cells. Tab TPM_GSE234761_CAST_M_x_129_F contains results from four replicate transcriptomes of CAST/EiJ male x 129/SvImJ female F1 hybrid cells from (Marks *et al.*, 2015). Tab TPM_GSE60738_CAST_M_x_129_F contains results from four replicate transcriptomes of replicates of CAST/EiJ male x 129/SvImJ female F1 hybrid cells from this study. Tab TPM_GSE90516_CAST_M_x_BL6_F contains results from 12 replicate transcriptomes of CAST/EiJ male x C57BL/6J female cells (Werner *et al.*, 2017). Tab TPM_GSE90516_CAST_F_x_BL6_M contains results from 11 replicate transcriptomes of CAST/EiJ female x C57BL/6J male cells (Werner *et al.*, 2017). In each tab, the first column reports Ensembl gene accession numbers. The remaining columns are TPM (transcripts per million) values for reads mapped non-ambiguously to either parental allele for each gene (two columns for each biological sample).

Table S2. Normalized read counts from RNA-seq of stem cells from homozygous strains. Each tab reports expression measurements from one data source of experimental profiles of cell culture from homozygous mouse strains: pluripotent stem cells (PSCs) of BL6 and CAST/EiJ genotypes from

this study; embryonic stem cells of eight genotypes from (Skelly et al. 2020);

Eed knockouts in trophoblast stem cells from (Weigert *et al.*, 2023); and *Ehmt2*

knockouts in E8.5 embryos (Auclair *et al.*, 2016). In each tab, the first column

reports Ensembl gene accession numbers. The remaining columns report the

TPM (transcripts per million) values. Tab

TPM_GSE234761_homozyg_CAST_129 contains results from two replicate

transcriptomes of CAST PSCs and four replicate transcriptomes of 129 PSCs.

Tab TPM_E_MTAB_7730_8_strains contains results from ESC replicate

transcriptomes from eight strains (3 replicates CAST, 35 replicates PWD, 3

replicates WSB, 36 replicates BL6, 3 replicates 129, 3 replicates AJ, 22

replicates NOD, 3 replicates NZO). Tab TPM_GSE198076_EED_KO contains

results from two replicate transcriptomes of CD-1 wildtype (WT) trophoblast

stem cells (TSCs) and two replicate transcriptomes of *Eed* knockout TSCs.

Tab TPM_GSE71500_EHMT2_KO contains results from two replicate

transcriptomes of WT BL6 embryos (E8.5) and two replicate transcriptomes of

Ehmt2 knockout embryos (E8.5).

Table S3. Mapping results from simulated RNA-seq experiments using

hybrid genomes. The first and second tabs report the mapped read counts

from simulated RNA-seq of CAST x 129 and CAST x BL6 hybrids, respectively.

In a given tab, for each row, the first column lists the Ensembl gene id for the

gene analyzed; the second through fourth columns report the mapping

percentage (the number of simulated reads mapped, normalized by the total

number of simulated reads) for the indicated allele in the indicated simulated

replicate; the last column reports the absolute value of the difference of

mapping percentages between the CAST allele and that of the indicated strain

(129 or BL6).

Table S4. Accession ID numbers for gene homologs across strains. Gene

and transcript accession numbers used to compare homologous alleles

between strains. GRCm38_ensembl_gene_ID, Ensembl gene IDs for the

C57BL/6 reference. 129_MGP_gene_ID, 129S1/SvImJ Mouse Genome

Project IDs. CAST_MGP_gene_ID, CAST/EiJ Mouse Genome Project IDs.

gene_name, gene names from the Mouse Genome Informatics database.

GRCm38_ensembl_transcript_ID is the Ensembl transcript ID for the longest

in-frame ORF for that gene; empty cells in this column are genes with no valid

ORF in the GRCm38 genome using the Ensembl build 102 annotations.

Table S5. Differential expression test results in hybrid and homozygous

stem cells. Each tab reports results of a genome-wide survey of differential

expression between homologous alleles in interspecies hybrid stem cells;

between homozygote stem cells of different species genotypes; or between a

regulator gene knockout and its isogenic wild-type in stem cell or early embryo

samples. Log2FoldChange, log-transformed ratio of normalized expression of

the indicated alleles or homozygote samples; Pvalue, Wald test p-value; Padj,

Benjamini-Hochberg adjusted p-value. Tab

DESeq_GSE60738_CASTx129_hyb, comparison of CAST vs 129 allele-

specific expression in F1 hybrid stem cells from (Marks *et al.*, 2015). Tab

DESeq_GSE234761_CASTx129_hyb, comparison of CAST vs 129 allele-

specific expression in F1 hybrid stem cells from this study. Tab

DESeq_GSE90516_CAST_x_BL6_hyb, comparison of CAST vs BL6 allele-

specific expression in F1 hybrid stem cells from (Werner *et al.*, 2017). Tab

DESeq_GSE234761_CASTvs129, comparison of CAST vs 129 expression in

homozygous stem cells from this study. Tab

DESeq_E_MTAB_7730_CASTvs129, comparison of CAST vs 129 expression

in homozygous stem cells from (Skelly et al. 2020). Tab

DESeq_E_MTAB_7730_CASTvsBL6, comparison of CAST vs BL6 expression

in homozygous stem cells from (Skelly et al. 2020). Tab

DESeq_E_MTAB_7730_CASTvsWSB, comparison of CAST vs WSB

expression in homozygous stem cells from (Skelly et al. 2020). Tab

DESeq_E_MTAB_7730_CASTvsPWD, comparison of CAST vs PWD

expression in homozygous stem cells from (Skelly et al. 2020). Tab

DESeq_E_MTAB_7730_CASTvsAJ, comparison of CAST vs AJ expression in

homozygous stem cells from (Skelly et al. 2020). Tab

DESeq_E_MTAB_7730_CASTvsNOD, comparison of CAST vs NOD

expression in homozygous stem cells from (Skelly et al. 2020). Tab

DESeq_E_MTAB_7730_CASTvsNZO, comparison of CAST vs NZO

expression in homozygous stem cells from (Skelly et al. 2020). Tab

DESeq_GSE198076_EED_KO, comparison of wildtype vs *Eed* knockout

expression in stem cells from (Weigert *et al.*, 2023). Tab

DESeq_GSE71500_EHMT2_KO, comparison of wildtype vs *Ehmt2* knockout

expression in E8.5 embryos from (Auclair *et al.*, 2016).

Table S6. A screen for directional allele-specific expression in CAST x

129 hybrid embryonic stem cells. Each row reports results of a statistical test

for directional allele-specific expression variation in CAST/EiJ male x

129/SvImJ female F1 hybrid embryonic stem cells (Marks *et al.*, 2015) in the

indicated Gene Ontology term. total_number_of_genes, number of genes in

the term with analyzable expression measurements; sum_sign_statistic, sum

across genes of the term of $\log_2$(129 allele expression/CAST allele

expression); resampling_P_val, resampling-based significance of the

enrichment for high absolute value of the sign statistic in the term; Adjusted_P,

p-value after Benjamini-Hochberg correction for multiple testing.

Table S7. Counts of divergent and polymorphic sites in regions upstream

of gene start sites, in comparisons between wild mouse populations.

Shown are results of analyses of genome sequences from wild mouse

populations taken from Harr et al. (2016). Each of the first six tabs reports

results from a comparison between a *M. m. castaneus* population (captured in

northwest India) and one population of another subspecies: *M. m. domesticus*

populations from France (Mmd_FRA), Germany (Mmd_GER), and Iran

(Mmd_IRA); or *M. m. musculus* populations from Kazakhstan (Mmm_KAZ), the

Czech Republic (Mmm_CZE), and Afghanistan (Mmm_AFG). In a given one of

the first six tabs, the first three columns report Ensembl (build 102) transcript

IDs, gene IDs, and gene names, and each of the remaining columns reports

the number of divergent sites, D_u, or the number of polymorphic sites, P_u, from

the 5-kb window at the indicated position upstream of the transcription start

site for the indicated gene. In the final two tabs, each row reports the mean D_u

and P_u values at the indicated upstream window across all genes in the

genome and genes of the translation GO term respectively, in a comparison

between *M. m. castaneus* and the indicated population of *M. m. domesticus* or

M. m. musculus.

Table S8. Tests for enriched divergence relative to polymorphism in the

upstream regions of translation genes, in comparisons between wild

mouse populations. In the first tab, each column reports results of resampling

tests for elevated values of the ratio between the number of divergent sites, D_u,

and the number of polymorphic sites, P_u, between *M. m. castaneus* and the

indicated population of *M. m. domesticus* or *M. m. musculus* from Table S6, in

upstream regions of translation genes. The first column reports the 5-kb

window upstream of the transcription start site subject to analysis. Each of the

following pairs of columns reports, respectively, the mean D_u/P_u across the

translation gene cohort in the indicated population in a comparison against *M.*

m. castaneus, and the *p*-value from a resampling test for high mean D_u/P_u in

translation genes relative to the genomic null. The second tab reports the

results of a calculation in which we repeated the resampling analyses of the

first tab for the window at 25kb upstream but left out each gene in turn. In a

given row, the first column reports the gene left out, and the remaining pairs of

columns report D_u/P_u across the translation gene cohort and the *p*-value from a

resampling test as in the first tab.

Table S9. Signatures of selection between mouse subspecies in

transcription factor binding sites upstream of translation genes. Each row

reports results of a two-factor ANOVA testing, across *M. m. musculus* and M*.*

m. domesticus populations, for an elevated value of the ratio of sites divergent

from *M. m. castaneus*, D_u, to the number of sites polymorphic within

populations, P_u, in binding sites of the indicated transcription factor upstream of

genes of the translation GO term, relative to non-binding sites and non-

translation genes (see Figure 3A of the main text). The second and third

columns report the ANOVA F statistic and nominal *p*-value, respectively, and

the third reports the *p*-value after Benjamini-Hochberg correction for multiple

testing.

Table S10. Counts of divergent and polymorphic sites in binding sites of

top three candidate transcriptional regulators. Each row reports counts

upstream of a gene, of sequence variation between wild mouse populations in

a binding site for Eed, Ehmt2, or Msx1 from the GTRD database (Kolmykov *et*

al., 2021). The first column reports the Ensembl transcript ID. The next three

columns report the chromosome, strand and transcription start site (TSS)

position for that gene, respectively. Column E reports whether or not the gene

is part of the translation GO term. Column F reports the transcription factor to

the binding site for that row. Columns G-H report the chromosomal positions of

the start and end of the binding site, respectively. Columns I-J report the

upstream distance of the binding site start and end, respectively, relative to the

TSS. Column K reports the length of the binding site which fall within the first

50kb upstream of the TSS. The remaining pairs of columns report the

divergent sites within the binding site, d_{TF}, or the number of polymorphic sites

within the binding site, p_{TF}, from a comparison between a *M. m. castaneus*

population (captured in northwest India) and one population of another

subspecies: *M. m. domesticus* populations from France (Mmd_FRA), Germany

(Mmd_GER), and Iran (Mmd_IRA); or *M. m. musculus* populations from

Kazakhstan (Mmm_KAZ), the Czech Republic (Mmm_CZE), and Afghanistan

(Mmm_AFG).

A — *M. m. castaneus vs M. m. musculus* (AFG)

Y-axis: Normalized D_u/P_u

X-axis: Window upstream start position (-50kb, -45kb, -40kb, -35kb, -30kb, -25kb, -20kb, -15kb, -10kb, -5kb)

B

M. m. castaneus population vs other population:		-25kb to -20kb Normalized D_u/P_u	Resampling *P*
M. m. musculus	Czech Rep.	*1.168*	*0.027*
	Afghanistan	*1.202*	*0.013*
	Kazakhstan	*1.153*	*0.034*
M. m. domesticus	Iran	*1.176*	*0.032*
	Germany	1.012	0.430
	France	1.076	0.140

A

M. m. castaneus population vs other population:		D_u/P_u in Eed binding sites		D_u/P_u in remaining sequence	
		Translation GO genes	Rest of genome	Translation GO genes	Rest of genome
M. m. domesticus	France	0.133	0.061	0.063	0.060
	Germany	0.128	0.061	0.057	0.060
	Iran	0.080	0.040	0.035	0.035
M. m. musculus	Afghanistan	0.106	0.031	0.048	0.036
	Czech Rep.	0.136	0.047	0.057	0.045
	Kazakhstan	0.119	0.043	0.056	0.043

B

M.m.c. India

M.m.d. Germany

M.m.d. France

M.m.d. Iran

M.m.m. Czech Rep.

M.m.m. Afghanistan

M.m.m. Kazakhstan

M. m. castaneus

M. m. domesticus

M. m. musculus

C

KO up

WT up

24.1% of genes
74.7% of genes

50.6% of genes
25.3% of genes

DE significance
- p-adj < 0.05
- p-adj > 0.05

-log₁₀(p-adj)

log₂(Wildtype expression/Eed knockout expression)