

# Comparative Transcriptomics Analyses across Species, Organs, and Developmental Stages Reveal Functionally Constrained lncRNAs

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1	Article
2	Comparative transcriptomics analyses across species, organs and developmental stages
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#### 27

#### Abstract

The functionality of long non-coding RNAs (IncRNAs) is disputed. In general, IncRNAs are under 28 29 weak selective pressures, suggesting that the majority of lncRNAs may be non-functional. However, 30 although some surveys showed negligible phenotypic effects upon IncRNA perturbation, key biological 31 roles were demonstrated for individual lncRNAs. Most lncRNAs with proven functions were implicated 32 in gene expression regulation, in pathways related to cellular pluripotency, differentiation and organ 33 morphogenesis, suggesting that functional lncRNAs may be more abundant in embryonic 34 development, rather than in adult organs. To test this hypothesis, we perform a multi-dimensional 35 comparative transcriptomics analysis, across five developmental time-points (two embryonic stages, 36 newborn, adult and aged individuals), four organs (brain, kidney, liver and testes) and three species 37 (mouse, rat and chicken). We find that, overwhelmingly, IncRNAs are preferentially expressed in adult 38 and aged testes, consistent with the presence of permissive transcription during spermatogenesis. 39 LncRNAs are often differentially expressed among developmental stages and are less abundant in 40 embryos and newborns compared to adult individuals, in agreement with a requirement for tighter 41 expression control and less tolerance for noisy transcription early in development. For differentially 42 expressed IncRNAs, we find that the patterns of expression variation among developmental stages are 43 generally conserved between mouse and rat. Moreover, IncRNAs expressed above noise levels in 44 somatic organs and during development show higher evolutionary conservation, in particular at their 45 promoter regions. Thus, we show that functionally constrained lncRNA loci are enriched in developing 46 organs, and we suggest that many of these loci may function in an RNA-independent manner.

#### 47 Introduction

Long non-coding RNAs (IncRNAs, loosely defined as transcripts without protein-coding potential, at least 200 nucleotides long) are an excellent illustration of the ongoing conceptual tug-ofwar between biochemical activity and biological function (Graur et al. 2013; Doolittle 2018). Recent sequencing-based studies identified thousands of IncRNAs in vertebrates (Guttman et al. 2009; Khalil et al. 2009; Iyer et al. 2015; Pertea et al. 2018). While this class of transcripts includes IncRNAs with undisputed biological roles, such as *Xist* (Brown et al. 1991) or *H19* (Brannan et al. 1990), experimental validations are lacking for the great majority of IncRNAs and their functionality is controversial.

The first functional characterizations of individual IncRNAs forged the idea that they are important contributors to gene expression regulatory networks. This has been unequivocally proven for some IncRNAs, such as *Xist*, whose transcription and subsequent coating of the X chromosome triggers a complex chain of molecular events leading to X inactivation in placental mammals (Munschauer et al. 2018). Other proposed mechanisms for gene expression regulation by IncRNAs include directing chromatin-modifying complexes at specific genomic locations, to control gene

61 expression in trans (Rinn et al. 2007); providing decoy targets for microRNAs (Cesana et al. 2011); 62 enhancing neighboring gene expression through an RNA-dependent mechanism (Ørom et al. 2010). 63 Biological functions unrelated to gene expression regulation were also proposed for IncRNAs. For 64 example, the NORAD IncRNA was shown to assemble a topoisomerase complex critical for genome stability (Munschauer et al. 2018), while the X-linked Firre IncRNA is involved in chromatin super-loop 65 66 formation on the inactive X chromosome (Hacisuleyman et al. 2014; Barutcu et al. 2018). Additional evidence that individual IncRNAs are undoubtedly biologically relevant comes from associations with 67 68 human diseases, including cancer, as for the SAMMSON IncRNA (Vendramin et al. 2018).

69 Initial studies of IncRNA functionality generally asserted that biological functions are directly 70 carried out by the transcribed RNA molecules. For some lncRNAs, this hypothesis was supported by 71 thorough functional tests, including rescue experiments showing that phenotypic effects of IncRNA 72 locus deletion can be reversed by expressing the lncRNAs in trans (Munschauer et al. 2018). Thus, for 73 a subset of IncRNA loci, their biological function is undoubtedly achieved by the non-coding RNA 74 molecule. However, in some cases, lncRNA function resides in the act of transcription at a given 75 genomic location, rather than in the product of transcription (Latos et al. 2012). In other cases, 76 biological functions are carried out by other elements embedded in the IncRNA genomic loci (Bassett 77 et al. 2014). For example, transcription of Linc-p21, originally described as a cis-acting enhancer 78 IncRNA, is not needed to regulate neighboring gene expression, which is instead controlled by multiple 79 enhancer elements within the locus (Groff et al. 2016). Genetic engineering of multiple IncRNA loci in 80 mouse likewise indicated that lncRNA transcripts are dispensable, and that gene expression regulation 81 by IncRNA loci is instead achieved by the process of transcription and splicing, or by additional 82 regulatory elements found in lncRNA promoters (Engreitz et al. 2016; Anderson et al. 2016). 83 Furthermore, some attempts to look for IncRNA function through genetic engineering approaches 84 showed that the tested IncRNA loci are altogether dispensable (Amândio et al. 2016; Zakany et al. 85 2017; Goudarzi et al. 2019). These recent observations signal a paradigm shift in IncRNA biology, as it 86 is increasingly acknowledged that, even when phenotypic effects can be unambiguously mapped to 87 IncRNA loci, they may not be driven by the IncRNA transcripts themselves.

88 Importantly, this new perspective on IncRNA biology had been predicted by evolutionary 89 analyses, traditionally used to evaluate the functionality of diverse genomic elements (Haerty and 90 Ponting 2014; Ulitsky 2016). Evolutionary studies of IncRNAs in vertebrates agree that selective 91 constraint on IncRNA primary sequences is weak, though significantly above the genomic background 92 (Ponjavic et al. 2007; Kutter et al. 2012; Necsulea et al. 2014; Washietl et al. 2014; Hezroni et al. 2015). 93 These observations are compatible with the hypothesis that many of the lncRNAs detected with sensitive transcriptomics techniques may be non-functional noise (Ponjavic et al. 2007), or that their 94 95 function is carried out by small conserved elements, such that the selective constraint signal on the

entire IncRNA locus is overall weak (Ulitsky 2016). They also indicate that IncRNA functionality may not 96 97 reside in the primary transcribed sequence. Indeed, mammalian IncRNA promoters show higher levels 98 of sequence conservation, similar to protein-coding gene promoters (Necsulea et al. 2014), as 99 expected if they carry out additional regulatory functions independently of the transcribed RNA 100 molecule. Moreover, it was previously reported that, in multi-exonic lncRNAs, splicing regulatory 101 elements are more conserved than exonic sequences (Schüler et al. 2014; Haerty and Ponting 2015), 102 in agreement with the recent finding that lncRNA splicing can contribute to neighboring gene 103 regulation (Engreitz et al. 2016). Thus, detailed evolutionary analyses of IncRNA loci can bring insights 104 into their functionality, and can help prioritize candidates for experimental validation.

105 Most comparative IncRNA studies were so far restricted to adult organ transcriptomes. These 106 comparisons showed that IncRNAs are preferentially expressed in adult testes, during 107 spermatogenesis (Soumillon et al. 2013). This process is characterized by a permissive chromatin 108 environment, which can promote non-functional transcription (Soumillon et al. 2013). The resulting 109 IncRNA datasets may thus be enriched in non-functional transcripts. Additional lines of evidence 110 suggest that the search for functional lncRNAs should be extended beyond adult organ transcriptomes. 111 For example, involvement in developmental phenotypes was proposed for many experimentally-112 tested lncRNAs (Sauvageau et al. 2013; Ulitsky et al. 2011; Grote et al. 2013), and an enrichment for 113 developmental transcription factor binding was reported for the promoters of conserved lncRNAs 114 (Necsulea et al. 2014). These observations motivated us to add a temporal dimension to comparative 115 IncRNA transcriptomics studies. Therefore, here we characterize IncRNAs across species, organs and 116 developmental stages. We analyze the spatial and temporal expression patterns of protein-coding 117 genes and IncRNAs, in conjunction with their evolutionary conservation. We find that, while IncRNAs 118 are overall poorly conserved during evolution in terms of primary sequence or expression patterns, 119 higher frequencies of constrained IncRNAs are observed in embryonic transcriptomes. For many of 120 these loci, biological function may be RNA-independent, as the highest levels of sequence conservation 121 are observed on promoter regions and on splice signals, rather than on IncRNA exonic sequence. Our 122 results are thus compatible with unconventional, RNA-independent functions for evolutionarily 123 conserved lncRNA loci, in particular for those that are expressed during embryonic development.

- 124 Results
- 125

#### Comparative transcriptomics across species, organs and developmental stages

To study protein-coding and IncRNA expression patterns across both developmental and evolutionary time, we generated strand-specific RNA-seq data for mouse and rat, for four major organs (brain, kidney, liver and testes) and five developmental time points, including two embryonic stages, newborn, young and aged adult individuals (Figure 1A, Supplementary Table 1, Materials and 130 methods). The selected time points allow us to obtain a broad view of major organ ontogenesis and to 131 capture drastic physiological changes during development (Theiler 1989). We chose to include in our 132 study both young adult (8-10 weeks old) and aged adult individuals (12 to 24 months old), thus 133 completing our overview of temporal patterns of gene expression variation. At the earliest embryonic 134 stage (day 13.5 post-conception for mouse, day 15 for rat), we dissected only the three somatic organs. 135 Our experimental design for mouse and rat thus comprises 19 organ / developmental stage 136 combinations. To obtain a broader evolutionary perspective we generated comparable RNA-seq data 137 for the chicken, for the two earliest developmental stages (Figure 1A, Supplementary Table 1). We 138 generated between 2 and 4 biological replicates for each species/organ/developmental stage 139 combination (Supplementary Table 1). Additional RNA-seq samples from previous publications were 140 included in the IncRNA annotation process, to increase detection sensitivity (Supplementary Table 2).

141 The organs and developmental stages included in our study differ greatly in terms of their 142 cellular composition diversity. To verify that our whole-organ RNA-seq data reflects cellular 143 composition heterogeneity, we assessed the expression patterns of cell population markers derived 144 from single-cell transcriptomics studies (Tabula Muris Consortium 2018; Green et al. 2018) in our 145 samples (Supplementary Figure 1, Supplementary Table 3, Supplementary Methods). This analysis 146 confirms that our transcriptome collection reflects expected developmental patterns. For example, 147 mature oligodendrocyte cell markers are systematically highly expressed in adult brain, while 148 oligodendrocyte precursor markers are more highly expressed in the earliest developmental stages 149 (Supplementary Figure 1). To further characterize our transcriptome collection, we sought to identify 150 genes that could serve as markers for organ/developmental stage combinations. To do this, we 151 selected genes that have narrow expression distributions, and for which maximum expression is 152 observed in the same organ/developmental stage combination in mouse and rat (Supplementary 153 Methods, Supplementary Table 4). Gene ontology enrichment analyses for these lists of genes are 154 coherent with the cellular composition and biological processes at work. Thus, genes involved in 155 forebrain neuron differentiation are over-represented in the mid-stage embryonic brain, while 156 processes related to synaptic transmission are enriched among genes specifically expressed in adult 157 brain (Supplementary Figure 2). In the kidney, the early developmental stages are enriched in genes 158 involved in metanephric development (Supplementary Figure 3). The newborn liver stands out due to 159 its strong enrichment in genes involved in immune response, while metabolic processes are over-160 represented in the adult liver (Supplementary Figure 4). Embryonic testes samples express genes 161 implicated in gamete generation and gene silencing by miRNAs, including the Piwi-like genes, while 162 adult testes transcriptomes are dominated by genes involved in spermatogenesis (Supplementary Figure 5). These patterns confirm that our whole-organ transcriptome collection captures the cell 163 164 composition changes and physiological transitions that occur during organ development.

#### Developmental expression patterns are well conserved among species for protein-coding genes

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To gain a first glimpse into the evolution of developmental gene expression patterns, we 166 167 performed a principal component analysis (PCA) for 10,363 protein-coding genes shared among 168 mouse, rat and chicken (Figure 1B, Supplementary Methods). This analysis revealed that the main 169 source of gene expression variability among species, organs and developmental stages is the 170 distinction between adult and aged testes and the other samples, which are separated on the first PCA 171 axis (Figure 1B). In contrast, embryonic and newborn testes are grouped with kidney samples from 172 similar developmental stages, in agreement with the common developmental origin of the kidney and the gonads (Nel-Themaat et al. 2010). The first axis of the PCA, which explains 67% of the total 173 174 expression variance, also correlates with the developmental stage for the brain: samples derived from 175 adult and aged individuals have higher coordinates on this axis than embryonic and newborn samples 176 (Figure 1B, Supplementary Figure 6, Kruskal-Wallis test p-value 0.003). The second PCA axis (10% 177 explained variability) mainly reflects the difference between brain and the other organs, but is also 178 associated with the developmental stage for kidney and liver (Figure 1B, Supplementary Figure 6, Kruskal-Wallis test p-value 4e<sup>-4</sup> for kidney, 4e<sup>-5</sup> for liver). However, we note that the association 179 between PC2 and developmental stages for kidney and liver may be confounded by differences in RNA 180 181 among developmental stages for these organs (Supplementary Methods, Supplementary Figure 6).

182 While mouse and rat samples are almost undistinguishable on the PCA factorial map, there is 183 considerably higher expression divergence between chicken and the two rodent species (Figure 1B). 184 However, differences among major organs are stronger than differences among species, even at these 185 broad evolutionary distances: brain samples all cluster together, irrespective of the species of origin, 186 and are clearly separated from kidney and liver samples on the second PCA axis (Figure 1B). These 187 patterns of gene expression variations are confirmed by a hierarchical clustering analysis based on 188 Spearman's correlation coefficients between pairs of samples (Figure 1C). The strongest clustering is 189 observed for adult and aged testes samples, followed by brain samples (Figure 1C).

The grouping among samples derived from similar organs and developmental stages, 190 191 irrespective of the species of origin, is stronger for genes that are associated with embryonic 192 development and with gene expression regulation (Supplementary Methods, Supplementary Figure 193 6C, D). For this set of genes, both the principal component analysis and the hierarchical clustering 194 analysis show a near-perfect separation of organs and developmental stages, for all three species 195 (Supplementary Figure 6C, D). Chicken samples, which cluster apart from rodent samples in whole 196 transcriptome analyses, are now grouped with the corresponding organs and developmental stages 197 from mouse and rat. Our transcriptome collection can thus reveal highly conserved expression 198 patterns for regulators of embryonic development, across amniotes.

199

#### Variations in transcriptome complexity among organs and developmental stages

200 We next sought to assess the transcriptome complexity in different organs across 201 developmental stages. To predict lncRNAs, we used the RNA-seq data to reconstruct gene models with 202 StringTie (Pertea et al. 2015), building on existing genomic annotations (Cunningham et al. 2019). We 203 verified the protein-coding potential of newly annotated transcripts, based on the codon substitution 204 frequency score (Lin et al. 2007, 2011) and on sequence similarity with known proteins, and we applied 205 a stringent series of filters to reduce contaminations from un-annotated protein-coding UTRs and 206 other artefacts (Materials and methods). We thus obtain a total of 18858 candidate lncRNAs in the 207 mouse, 20159 in the rat and 5496 in the chicken, including both newly-annotated and previously 208 known IncRNAs transcribed in our samples (Supplementary Dataset 1). The relative sizes of each 209 species' IncRNA repertoires are consistent with previous studies (Necsulea et al. 2014; Sarropoulos et 210 al. 2019). We note however that our power to detect IncRNAs in chicken is limited, due to the narrower 211 organ and developmental stage sampling in this species (Supplementary Tables 1,2). Most candidate 212 IncRNAs are expressed at very low levels. When imposing a minimum normalized expression level 213 (transcript per million, or TPM) of 1, in at least one sample, the numbers of candidate lncRNAs falls to 214 12199, 15319 and 2892 in the mouse, rat and chicken, respectively (Supplementary Datasets 2, 3).

215 The differences in IncRNA content among species may be affected by RNA-seq read coverage 216 and sample distribution, as well as genome sequence and annotation quality. To correct for the effect 217 of RNA-seq read coverage, we down-sampled the RNA-seq data to obtain the same number of uniquely 218 mapped reads for each organ/developmental stage combination within each species (Supplementary 219 Methods). After this procedure, the number of detectable protein-coding genes (supported by at least 220 10 uniquely mapped reads) still shows broad variations among organs and developmental stages, with 221 the highest numbers of genes detected in the testes, for all time points (Figure 2A). Large numbers of 222 protein-coding genes (between approximately 12800 and 16700) are detected in all samples. In 223 contrast, for IncRNAs, the pattern is much more striking: the young and aged adult testes express 224 between 11000 and 12000 lncRNAs, in both mouse and rat, while in somatic organs and earlier 225 developmental stages we can detect only between 1800 and 4800 lncRNAs (Figure 2B). This 226 observation is in agreement with previous findings indicating that during spermatogenesis the 227 chromatin environment is highly permissive to transcription (Soumillon et al. 2013).

228

# Spatial and temporal expression patterns for protein-coding genes and IncRNAs

We next compared spatial and temporal expression patterns between protein-coding genes and IncRNAs. In agreement with previous findings (Soumillon et al. 2013), we show that IncRNAs are overwhelmingly preferentially expressed in the testes (Figure 3A). Indeed, more than 60% of IncRNAs reach their maximum expression level in this organ, compared to less than 35% of protein-coding 233 genes, for both mouse and rat (Figure 3A; Chi-square test p-value  $<1e^{-10}$ ). Almost 80% of lncRNAs are 234 preferentially expressed in young or aged adult samples, which is significantly higher than the fraction 235 observed for protein-coding genes (less than 65%, Chi-square test p-value  $<1e^{-10}$ , Figure 3B).

We found that between 57% and 80% of protein-coding genes are significantly differentially 236 237 expressed (DE) among developmental stages, at a false discovery rate (FDR) below 1%, in each organ 238 and species (Figure 3C, Supplementary Dataset 4). The proportions of DE IncRNAs are significantly 239 lower than the proportions of DE protein-coding genes in somatic organs, between 17% and 41% (Chisquare test, p-value  $<1e^{-10}$ ). In the testes, we observed higher proportions of DE lncRNAs (63% in 240 mouse and 67% in rat), but these values were still significantly lower than those observed for protein-241 coding genes (77% in mouse and 79% in rat; Chi-square test, p-value <1e<sup>-10</sup>; Figure 3C). We suspected 242 that the lower proportion of DE IncRNAs could be due to their low expression levels, as total read 243 244 counts affect the sensitivity of DE tests (Anders and Huber 2010). Indeed, IncRNAs are expressed at 245 much lower levels than protein-coding genes (Supplementary Figure 7). To control for this, we down-246 sampled the read counts observed for protein-coding genes, bringing them to the same average counts 247 as IncRNAs but preserving relative gene abundance (Materials and methods). Strikingly, after downsampling, we observe higher proportions of DE loci for IncRNAs compared to protein-coding genes 248 (Figure 3C). The differences are statistically significant (Chi-square test, p-value <1e<sup>-10</sup>) in all but one 249 250 species / organ combination (mouse kidney, Chi-square test, p-value 0.15). We also observed that the 251 expression amplitude among developmental stages are more important for lncRNAs than for proteincoding genes (Wilcoxon test, p-value <1e<sup>-10</sup>, Supplementary Figure 8A), as expected given the lower 252 253 IncRNA expression levels, which preclude detecting subtle expression shifts among time points. Finally, 254 we observe that the developmental stage with maximum expression is generally different between 255 protein-coding genes and IncRNAs, even when considering genes that are significantly DE among 256 stages. For all organs, DE IncRNAs tend to show highest expression levels in the young and aged adults, while DE protein-coding genes are more homogeneously distributed among developmental stages 257 (Chi-square test, p-value  $<1e^{-10}$ , Figure 3D, Supplementary Figure 8B). 258

259 Similar conclusions are reached when performing DE analyses between consecutive time 260 points (Supplementary Figure 9, Supplementary Dataset 4). For both protein-coding genes and 261 IncRNAs, the strongest expression changes are observed between newborn and young adult individuals. Almost 10000 IncRNAs are significantly up-regulated between newborn and young adult 262 263 testes, confirming the strong enrichment for IncRNAs during spermatogenesis (Supplementary Figure 9). As expected, the lowest numbers of DE genes are observed at the transition between young and 264 aged adult organs. At this time point, we observe more changes for the rat than for the mouse, 265 266 potentially due to a higher proportion of immune cell infiltrates in the rat aged organ samples. Genes

associated with antigen processing and presentation tend to be expressed at higher levels in aged
adults than in young adults, for mouse kidney, rat brain and liver (Supplementary Dataset 4).

269

# Stronger selective constraint on IncRNAs expressed earlier in development

We next analyzed the long-term evolutionary sequence conservation for lncRNAs, in conjunction with their spatio-temporal expression patterns (Supplementary Table 5). We used the PhastCons score (Siepel et al. 2005) for placental mammals (Casper et al. 2018), to assess sequence conservation for various aspects of mouse lncRNAs: exons, promoters (defined as 400 bp regions upstream of the transcription start site), splice sites (first and last two bases of the introns). As approximately 20% of lncRNAs overlap with exonic regions on the opposite strand (Supplementary Dataset 1), we masked exonic regions from other genes before evaluating sequence conservation.

277 As previously observed (Ponjavic et al. 2007; Haerty and Ponting 2013), exonic and splice site 278 sequence conservation is much lower for IncRNAs (median exonic score 0.094, median splice site score 279 0.075) than for protein-coding genes (median exonic score 0.42, median splice site score 0.85, Wilcoxon test p-value < 1e<sup>-10</sup>, Supplementary Figure 10). Exonic IncRNA conservation scores are 280 significantly above the conservation observed for intergenic regions genome-wide (median score 281 0.076, Wilcoxon test, p-value < 1e<sup>-10</sup>, Supplementary Figure 10). Interestingly, intergenic regions found 282 in the vicinity of IncRNA loci (Supplementary Methods) had slightly lower conservation scores than all 283 intergenic regions, on average (median 0.072, Wilcoxon test, p-value  $< 1e^{-6}$ , Supplementary Figure 10). 284 Promoter conservation levels are more comparable between protein-coding genes (median score 285 286 0.17) and IncRNAs (median score 0.08), though still significantly higher for the former (Wilcoxon test, p-value < 1e<sup>-10</sup>, Supplementary Figure 10). Among lncRNA classes, the highest levels of promoter 287 288 sequence conservation (median 0.14) are observed for bidirectional promoters shared with protein-289 coding genes (Supplementary Figure 10).

290 We next analyzed sets of protein-coding genes and lncRNAs that are expressed above noise 291 levels (TPM>=1, averaged across all biological replicates) in each organ / developmental stage combination (Supplementary Table 6). For all examined regions and for both categories of genes, the 292 spatio-temporal expression pattern is associated with the level of sequence conservation. Globally, 293 294 sequence conservation is higher for genes that are expressed earlier in development than for genes 295 expressed later in development, and reaches its lowest values for genes expressed in adult and aged 296 testes (Figure 4). For exonic sequences and splice sites, the amount of sequence conservation is 297 significantly lower for IncRNAs than for protein-coding genes, irrespective of the organ and developmental stage in which they are expressed (Wilcoxon test, p-value < 1e<sup>-10</sup>, Figure 4A, C). 298 299 However, for promoter regions, the differences between the two gene categories are weaker, and are 300 not statistically significant for the mid-stage embryonic brain (median 0.21 for protein-coding genes, 0.20 for IncRNAs, Wilcoxon test, p-value 0.08) and kidney (median 0.20 for protein-coding genes, 0.21
 for IncRNAs, Wilcoxon test, p-value 0.76), and for the late embryonic kidney (median 0.20 for protein coding genes, 0.19 for IncRNAs, Wilcoxon test, p-value 0.15). As noted before, the highest levels of
 IncRNA promoter conservation are observed for IncRNAs that have bidirectional promoters shared
 with protein-coding genes or other non-coding loci (Supplementary Figure 11A-C).

306 Finally, we asked whether the highest level of evolutionary sequence conservation is seen at 307 exons, promoter or splice site regions, for lncRNA loci taken individually. Here again, the answer 308 depends on the expression pattern: for IncRNAs detected in somatic organs and in the developing testes, there is significantly higher conservation for promoters than for exons (Wilcoxon test, p-value 309 < 1e<sup>-3</sup> for all organ / developmental stage combinations, Supplementary Figure 11D, E). We also 310 311 observe significantly higher conservation for splice sites than for exons (Wilcoxon test, p-value < 312 0.005), in all samples except aged liver (Wilcoxon test, p-value 0.052). However, when we consider 313 IncRNAs that are expressed above noise levels in the young and aged adult testes (which constitute 314 the great majority of loci), the conservation scores are slightly but significantly higher for exons than for promoters or splice sites (Wilcoxon test, p-value  $< 1e^{-9}$ , Supplementary Figure 11D, E). 315

316

# Detection of homologous IncRNAs across species

317 We next sought to assess the conservation of IncRNA repertoires in mouse, rat and chicken. 318 We detected IncRNA separately in each species, using only RNA-seq data and existing genome 319 annotations, as previously suggested (Hezroni et al. 2015). We then searched for putative 1-to-1 320 orthologous IncRNAs between species using pre-computed whole-genome alignments as a guide 321 (Materials and methods), to increase the sensitivity of orthologous gene detection in the presence of 322 rapid sequence evolution (Washietl et al. 2014). The orthologous IncRNA detection procedure involves 323 several steps, including the identification of putative homologous (projected) loci across species, 324 filtering to remove large-scale structural changes in the loci, and intersection with predicted loci in the 325 target species (Materials and methods). As illustrated in Figure 5, for comparisons between rodents the extent of sequence divergence is low enough that more than 90% of 18858 IncRNA loci are 326 327 successfully projected from mouse to rat (Figure 5A, Supplementary Dataset 5). However, only 54% of 328 projected loci have detectable transcription in the target species (at least 10 uniquely mapped reads). 329 Only 23% of mouse IncRNA loci have predicted 1-to-1 orthologues in the rat, and only 15% are 330 orthologous to confirmed IncRNA loci in the rat (Figure 5A, Supplementary Dataset 5). The 1493 mouse IncRNAs that have non-IncRNAs orthologues in the rat are generally matched with loci discarded 331 332 because of low read coverage, minimum exonic length or distance to protein-coding genes 333 (Supplementary Dataset 5). Cases of IncRNA-protein-coding orthologues are rare at this evolutionary 334 distance (Supplementary Dataset 5), and they may stem from gene classification errors. We note that

orthologous IncRNA gene structures are highly divergent across species, in terms of exonic length or number of exons (Supplementary Figure 12). At larger evolutionary distances, the rate of sequence evolution is the main factor hampering detection of orthologous IncRNAs. Only 2613 (14%) of mouse IncRNAs could be projected on the chicken genome, and after subsequent filtering we detect only 66 mouse – chicken IncRNA orthologues, and 30 IncRNAs with orthologues in all three species (Supplementary Dataset 5, Supplementary Table 7).

341 Conserved IncRNAs differ from species-specific IncRNAs in terms of expression patterns. While 342 only subtle differences can be observed when comparing mouse-rat orthologous lncRNAs to the 343 mouse-specific IncRNA set, IncRNAs that are conserved between mouse and chicken are enriched in 344 somatic organs and early developmental stages (Figure 5B, C). For example, only 15% of mouse-345 specific IncRNAs reach their maximum expression in the brain, which is significantly lower than the 346 observed proportion for mouse lncRNAs with orthologues in rat (18%, Chi-square test, p-value  $3e^{-4}$ ) and for mouse lncRNAs with orthologues in the chicken (39%, Chi-square test, p-value 1.5e<sup>-7</sup>). Likewise, 347 348 while only 9.9% of mouse-specific lncRNAs have their highest level of expression in one of the two 349 embryonic stages, this proportion is significantly higher for IncRNAs with orthologues in the chicken 350 (27%, Chi-square test, p-value 0.002). We note however that these results may be affected by our 351 narrower sampling for the chicken, which is biased towards embryonic organs, although we did include 352 data from adult organs for this species (Supplementary Table 2).

353

# Patterns of IncRNA expression variation across species, organs and developmental stages

354 We next assessed the global patterns of expression variation across species, organs and 355 developmental stages, for predicted mouse - rat lncRNA orthologues (Supplementary Dataset 6). As 356 for protein-coding genes, the main source of variability in a PCA performed on IncRNA expression levels 357 is the difference between adult and aged testes and the other samples (Figure 6A). However, for 358 IncRNAs, samples cluster according to the species of origin already on the second factorial axis (11.6% 359 explained variance), confirming that IncRNA expression patterns evolve rapidly. Overall, differences between organs and developmental stages are less striking for IncRNAs, compared to differences 360 between species (Figure 6A). This pattern is also visible on a hierarchical clustering analysis (performed 361 362 on distances derived from Spearman's correlation coefficient): in contrast with what is observed for 363 protein-coding genes, for lncRNAs samples generally cluster by species, with the exception of young 364 and aged adult testes, which are robustly grouped (Figure 6B).

The higher rates of IncRNA expression evolution are also visible when analyzing within-species variations, through comparisons across biological replicates (Figure 7A). We sought to measure the selective pressures acting on expression patterns by contrasting between-species and within-species variations, in the spirit of a classical approach for coding sequences (McDonald and Kreitman 1991). 369 We constructed an expression conservation index by dividing the between-species and the within-370 species Spearman's correlation coefficient, computed on all genes from a category, for a given 371 organ/developmental stage combination (Supplementary Methods). The resulting values are very high 372 for protein-coding genes, in particular for the brain and the mid-stage embryonic kidney, where the 373 expression conservation scores are above 0.95. However, there is significant less conservation 374 between species for the adult and aged testes (expression conservation score ~0.88, bootstrap p-value 375 <0.01, Figure 7B). For IncRNAs, expression conservation values vary between 0.5 and 0.7, significantly 376 lower than for protein-coding genes (bootstrap p-value <0.01). The lowest conservation scores are 377 observed for young and aged adult testes (Figure 7C).

378

#### Parallel patterns of temporal expression variation for mouse and rat IncRNAs

379 We delved deeper into the evolutionary comparison of protein-coding genes and IncRNA 380 expression patterns, by asking whether temporal expression variations are shared between species. 381 Several hundred orthologous IncRNAs are DE (FDR<0.01) in both mouse and rat, in each organ (minimum 150 in liver, maximum 1583 in testes, Figure 8A). Likewise, between 6775 (in liver) and 382 383 10608 (in testes) protein-coding genes are DE in both species (Supplementary Figure 13). Overall, 384 shared DE IncRNAs show similar patterns of variation among developmental stages in mouse and rat, 385 reaching their maximum expression in the same (or close) developmental stages (Figure 8A). For 386 example, out of 42 IncRNAs that are DE in mouse brain and reach their maximum expression in the 387 mid-stage embryo, 31 (74%) reach their maximum expression in the corresponding stage in the rat 388 (Figure 8A). We clustered the relative expression profiles using the K-means algorithm (Supplementary 389 Methods). Although individual gene trajectories show variations between species, the average 390 expression profiles within each K-means cluster are generally similar between mouse and rat (Figure 391 8B-E, Supplementary Figure 13). This is particularly striking for the brain, where all 5 IncRNA clusters 392 show similar average expression profiles for the two species (Figure 8B). Greater differences between 393 species are observed in other organs, such as the kidney, where 2 out of 5 clusters (120 genes in total, 394 that is 31% of shared DE IncRNAs in kidney) have average expression profiles that reach their maximum in different stages in mouse and rat (Figure 8C). The promoters of shared DE IncRNAs in each cluster 395 396 contain transcription factor binding sites that are coherent with the expression profile of the cluster, 397 such as brain homeobox POU3F2/BRN2 binding sites for the first K-means cluster in the brain, which 398 has maximum expression in the mid-stage embryo (Supplementary Table 8). We note that 399 transcription factor binding site enrichments are generally not statistically significant for lncRNAs, 400 partly due to the low gene counts in each cluster.

401 The testis is the only organ where opposite K-mean cluster expression profiles are observed in 402 the two species (increasing with time in mouse and decreasing in rat, or vice-versa). For lncRNAs, this

403 occurs for 1 of the 4 detected clusters, containing 56 IncRNAs (3.5% of all shared DE IncRNAs in this 404 organ, Figure 8E). For protein-coding genes, opposite average profiles are observed for 2 out of 4 405 clusters, comprising 1182 and 1509 genes, i.e. 25% of all shared testes-DE protein-coding genes 406 (Supplementary Figure 13). These clusters do not stand out in terms of transcription factor binding site 407 (Supplementary Table 8) or gene ontology enrichment (Supplementary Dataset 4). This pattern 408 confirms previous reports of rapid expression evolution in the adult testes (Brawand et al. 2011), and 409 extends them by showing that patterns of variations among developmental stages are often species-410 specific in the testes, for protein-coding genes.

411

# Evolutionary divergence of individual IncRNA expression profiles

412 To further quantify IncRNA expression differences between species, we measured the 413 Euclidean distance between relative expression profiles (average TPM values across biological replicates, normalized by dividing by the sum of all values for a gene, for each species), for mouse and 414 415 rat orthologues (Supplementary Methods, Supplementary Dataset 7, Supplementary Table 9). The 416 resulting expression divergence values correlate negatively with the average expression level (Rsquared 0.13, t-test p-value <  $1e^{-10}$ , Figure 9A), as expected given that abundance estimation is less 417 reliable for weakly expressed genes. While the raw expression divergence values are significantly 418 419 higher for IncRNAs (median 0.18) than for protein-coding genes (median 0.11, Wilcoxon test p-value < 1e<sup>-10</sup>, Figure 9B), this is largely due to the low lncRNA expression levels. Indeed, the effect disappears 420 when analyzing the residual expression divergence after regressing the expression level (median value 421 -0.03 for protein-coding genes, -0.06 for lncRNAs, Wilcoxon test < 1e<sup>-10</sup>, Figure 9C). These patterns 422 423 remain true when analyzing separately protein-coding and IncRNAs with different types of promoters, 424 bidirectional or unidirectional (Supplementary Figure 14A). For IncRNAs, we also observe a weak 425 negative correlation between expression divergence and the extent of gene structure conservation (Rsquared 0.04, t-test p-value <  $1e^{-10}$ , Figure 9D). We measured the relative contribution of each 426 427 organ/developmental stage to the expression divergence estimate (Figure 9E). For both protein-coding 428 genes and IncRNAs, by far the highest contributors are the young adult and aged testes samples, which are responsible for almost 30% of the IncRNA expression divergence (Figure 9E). This is visible in the 429 430 expression patterns of the 2 protein-coding and IncRNA genes with the highest residual expression 431 divergence: the IncRNA expression divergence is mostly due to changes in adult testes, while more 432 complex expression pattern changes seem to have occurred for the protein-coding genes (Supplementary Figure 14). The most divergent protein-coding genes are enriched in functions related 433 434 to immunity (Supplementary Dataset 7), suggesting that differences in immune cell infiltrates among 435 species could be responsible for these extreme cases of expression pattern divergence.

436

#### Candidate species-specific IncRNAs

We next investigated the most extreme cases of expression divergence: situations where 437 438 expression can be robustly detected in one species, but not in the other one, despite almost perfect 439 sequence alignment (Supplementary Methods). We selected IncRNA loci that were supported by at 440 least 100 uniquely mapped reads in one species, with no reads detected in the predicted homologous 441 region in the other species. With this convention, we obtain 1041 candidate mouse-specific and 1646 442 candidate rat-specific loci (Supplementary Dataset 8). These lists include striking examples, such as the 443 region downstream of the Fzd4 protein-coding gene, which contains a mouse-specific and a rat-specific 444 IncRNA candidate, each perfectly aligned in the other species (Supplementary Figure 15). Candidate 445 species-specific IncRNAs are more frequently associated with predicted enhancers than orthologous IncRNAs (52% vs. 33%, Chi-square test, p-value <1e<sup>-10</sup>), are less often spliced (56% vs. 61%, Chi-square 446 test p-value 1.6e<sup>-3</sup>) and associated with bidirectional promoters (24% vs. 61%, Chi-square test, p-value 447 <1e<sup>-10</sup>, Supplementary Figure 16). Moreover, we could confirm that their presence is associated with 448 449 increased expression divergence in the neighboring genes. To test this, we selected species-specific 450 and orthologous IncRNAs that are transcribed from bidirectional promoters shared with protein-451 coding genes, and evaluated the expression divergence of their protein-coding neighbors 452 (Supplementary Figure 16D, E). Though the difference is subtle, genes that are close to species-specific 453 IncRNAs have significantly higher expression divergence than the ones that have conserved IncRNA neighbors, even after correcting for expression levels (Wilcoxon test, p-value < 1e<sup>-3</sup>). It thus seems that 454 455 expression changes that led to the species-specific IncRNA transcription extend beyond the IncRNA 456 locus and affect neighboring genes, as previously proposed (Kutter et al. 2012).

#### 457 Discussion

458

#### Comparative transcriptomics across species, organs and developmental stages

459 More than a decade after the publication of the first genome-wide IncRNA datasets (Guttman 460 et al. 2009; Khalil et al. 2009), the debate regarding their functionality is still not settled. Evolutionary 461 approaches provide important tools to assess biological functionality (Haerty and Ponting 2014), and 462 they have been already successfully applied to IncRNAs. However, most large-scale comparative 463 transcriptomics studies available so far (Kutter et al. 2012; Washietl et al. 2014; Hezroni et al. 2015; 464 Necsulea et al. 2014), with one recent exception (Sarropoulos et al. 2019), have focused on IncRNAs 465 detected in adult organs. We hypothesized that IncRNAs expressed during development may be 466 enriched in functional loci, as suggested by the increasing number of IncRNAs with proposed 467 developmental roles (Rinn et al. 2007; Sauvageau et al. 2013; Grote et al. 2013; Grote and Herrmann 468 2015). To test this hypothesis, we performed a multi-dimensional comparative transcriptomics 469 analysis, following IncRNA and protein-coding genes across species, organs and developmental stages. 470 We ensured that our transcriptome collection reflects the changes in cellular composition and 471 physiological functions that occur during major organ development, by analyzing cell-type specific 472 gene markers derived from single-cell analyses (Tabula Muris Consortium 2018; Green et al. 2018). We 473 showed that protein-coding gene expression profiles across major organs and developmental stages 474 are well conserved among species, even at large evolutionary distances. Although differences among 475 rodents and chicken are considerable when analyzing the full set of orthologous protein-coding genes 476 (Figure 1), the expression profiles of genes that are known to be implicated in embryonic development 477 and in gene expression regulation processes are highly conserved among species (Supplementary 478 Figure 6). Our transcriptome collection thus enables detecting temporal expression patterns shared 479 across divergent species, for key players in developmental regulatory networks. These observations 480 are consistent with findings from a recent publication, which studied protein-coding gene expression 481 patterns during major organ development in amniote species (Cardoso-Moreira et al. 2019). Our 482 transcriptome dataset covers a narrower range of species and developmental stages than this 483 comprehensive resource (Cardoso-Moreira et al. 2019), but uniquely includes aged individuals, thus 484 completing the overview of temporal expression patterns. Thus, our work represents an additional 485 resource for evolutionary studies of gene expression.

To our knowledge, together with a recent publication (Sarropoulos et al. 2019), our work is one of the first large-scale lncRNA evolutionary studies that include a temporal dimension, by sampling different developmental stages. Our manuscript and this recent work concur to reveal an enrichment for functional lncRNAs early in development (Sarropoulos et al. 2019). Here, we perform in-depth analyses of expression pattern evolution, short-term and long-term sequence evolution for different regions of lncRNAs loci, in conjunction with their expression patterns. We thus bring new insights into the evolution and functionality of lncRNAs.

493

## Spatio-temporal IncRNA expression patterns

494 Our first major observation is that IncRNAs are overwhelmingly expressed in the young and 495 aged adult testes (Figure 3), in agreement with previous data (Soumillon et al. 2013). Their relative 496 depletion in embryonic and newborn testes reinforces the association between lncRNA transcription 497 and spermatogenesis, in accord with the hypothesis that the particular chromatin environment during 498 spermatogenesis is a driver for promiscuous, non-functional transcription (Kaessmann 2010; 499 Soumillon et al. 2013). Interestingly, we show that IncRNAs are significantly differentially expressed 500 among developmental stages, at least as frequently as protein-coding genes, after correcting for their 501 lower expression levels. However, in contrast with protein-coding genes, the majority of IncRNAs reach 502 their highest expression levels in adult rather than in developing organs (Figure 3). As requirements 503 for tight gene expression control are higher during embryonic development (Ben-Tabou de-Leon and

Davidson 2007), an explanation for the relative lncRNA depletion in embryonic and newborn transcriptomes is that transcriptional noise is deleterious and thus more efficiently blocked during the early stages of development. Differences in cellular composition heterogeneity may also be part of the explanation. Expression analyses of cell-type specific markers suggest that adult organ transcriptomes may be a mix of more diverse cell types, including substantial immune cell infiltrates (Supplementary Figure 1). A higher cell type diversity may explain the increased abundance of lncRNAs in young and aged adult organs, especially given that lncRNAs are thought to be cell-type specific (Liu et al. 2016).

511 We found that IncRNA expression patterns are generally similar between young and aged adult 512 individuals: less than 50 IncRNAs are significantly DE between these two stages, for most organs 513 (Supplementary Figure 9). Moreover, the levels of sequence and expression conservation are globally 514 similar between young and aged adults, for both protein-coding and IncRNA genes (Figures 4,7). 515 Overall, our analyses indicate that, with our sampling (Supplementary Table 1), the physiological 516 processes at work in aged organs are highly similar to those acting in juvenile organs, suggesting that 517 developmental stage sampling should be further extended for in-depth analyses of the aging process.

518

# Functionally constrained IncRNAs are enriched in developmental transcriptomes

519 Our long-term sequence conservation analyses confirm that IncRNAs are overall under weak, 520 but significant selective constraint (Ponjavic et al. 2007): IncRNA sequence conservation scores are 521 much lower than those of protein-coding genes, but above those of intergenic regions (Figure 4, 522 Supplementary Figures 10-11). Interestingly, intergenic regions flanking IncRNAs are on average less 523 conserved than the genomic intergenic average (Figure 4), suggesting that the rapid lncRNA evolution 524 may be a general feature of the genomic regions in which they reside. The underlying mechanisms are 525 unclear, but may reflect a lower density of constrained expression regulatory elements in the vicinity 526 of IncRNAs, or a higher accumulation of lineage-specific transposable elements (Kapusta et al. 2013).

527 We show that, for those lncRNAs that are expressed above noise levels (TPM>=1) in somatic 528 organs and in the embryonic and newborn developmental stages, there is a higher proportion of 529 evolutionarily constrained loci than in testes-expressed IncRNAs (Figure 4). Strikingly, we find that the 530 level of long-term sequence conservation for IncRNA promoter regions is similar to the one observed 531 for protein-coding promoters, when we analyze genes that are robustly expressed in embryonic brain 532 and kidney. Furthermore, we show that IncRNAs expressed in somatic organs and in the developing 533 testes differ from those expressed in the adult testes not only in terms of overall levels of sequence conservation, but also with respect to the regions of the lncRNA loci that are under selective constraint. 534 535 Thus, for IncRNAs expressed in somatic organs and in the developing testes, there is significantly more 536 evolutionary constraint on promoters and splice sites than on exons, while these patterns are not seen 537 for the bulk of IncRNAs, expressed in adult and aged testes (Supplementary Figure 11). We are thus

able to modulate previous reports of increased constraint on splicing regulatory regions in mammalian
IncRNAs (Schüler et al. 2014; Haerty and Ponting 2015), by showing that this pattern is specific to
IncRNAs that are expressed in somatic organs and in the developing testes.

541 These results are also in agreement with recent findings suggesting that biological function 542 may reside in the presence of additional non-coding regulatory elements at the lncRNA promoter 543 rather than in the production of a specific transcript (Engreitz et al. 2016; Groff et al. 2016). While the 544 elevated sequence conservation at splicing regulatory signals could indicate that the production of a 545 specific mature lncRNA is required, splicing of lncRNA transcripts was recently proposed to affect the expression of neighboring protein-coding genes (Engreitz et al. 2016). Thus, while there is evidence for 546 547 increased functionality for IncRNA loci that are detected in developmental transcriptomes or in adult 548 somatic organs, in agreement with a recent report (Sarropoulos et al. 2019), our sequence 549 conservation analyses are compatible with the hypothesis that their biological functions may be 550 carried out in an RNA-independent manner, as exons are under less constraint than promoters or 551 splice sites. Alternatively, their function may be carried out by small conserved elements, such that 552 the sequence conservation on the entire lncRNA exonic sequence is weak (Ulitsky 2016).

553

#### Evolutionary divergence of spatio-temporal expression profiles for IncRNAs

554 We previously showed that IncRNA expression patterns evolve rapidly across species in adult 555 organs (Necsulea et al. 2014). Here, we show that this rapid evolution of IncRNA expression is also true 556 for embryonic and newborn developmental stages. Expression comparisons across species, organs and 557 developmental stages are dominated by differences between species for IncRNAs (Figure 6), while 558 similarities between organs and developmental stages are predominant for protein-coding genes, 559 even across distantly-related species (Figure 1). We assessed the extent of expression conservation by 560 contrasting between-species and within-species expression variations and we showed that IncRNAs 561 have significantly lower levels of conservation than protein-coding genes, for all organs and 562 developmental stages (Figure 7). However, IncRNA expression is more conserved in somatic organs 563 and in early embryonic stages than in the adult testes. Moreover, when orthologous lncRNAs are 564 differentially expressed among developmental stages in both mouse and rat, they generally show 565 parallel profiles of expression variation in both species (Figure 8). This result is in agreement with a 566 recent publication, which showed that temporal patterns of expression variation tend to be 567 evolutionarily conserved for developmentally dynamic IncRNAs (Sarropoulos et al. 2019). We note that these temporal patterns of variation may in fact be caused by spatially-restricted lncRNA expression. 568 569 Previous reports indicated that lncRNA expression may be cell type-specific (Liu et al. 2016). The 570 differentially expressed lncRNAs, shared across mouse and rat, could be specific of cell types that 571 change their relative abundance in whole-organ transcriptomes with developmental time.

572 Interestingly, when we evaluate expression divergence individually for each orthologous gene 573 pair, correcting for the lower lncRNA expression levels, we find that lncRNAs are not more divergent 574 than protein-coding genes (Figure 8). This observation indicates that much of the between-species 575 differences in IncRNA expression patterns is tightly linked with the low expression levels of IncRNAs. It 576 is not clear however whether this is purely an indication of technical biases, that hamper expression 577 estimation for lowly expressed lncRNAs, or whether the low lncRNA expression levels are a sign that 578 these transcripts are non-functional. For cell type-specific lncRNAs, low expression in whole-organ 579 transcriptomes are expected. This question may soon be directly addressed, as single-cell assays 580 become more sensitive and allow investigation of lncRNAs (Liu et al. 2016).

# 581

596

#### Candidate species-specific IncRNAs

582 Finally, we analyzed extreme cases of expression divergence between species, where transcription can be robustly detected in one species but not in the other, despite the presence of 583 584 good sequence conservation. We identify more than a thousand candidate species-specific lncRNAs, 585 in both mouse and rat. Interestingly, we observe that candidate mouse-specific lncRNAs are more 586 frequently transcribed from enhancers than IncRNAs conserved between mouse and rat 587 (Supplementary Figure 11). This observation is consistent with previous reports that enhancers and 588 enhancer-associated lncRNAs evolve rapidly (Villar et al. 2015; Marques et al. 2013). Moreover, we 589 show that these lncRNA expression changes do not occur in an isolated manner. When species-specific 590 IncRNA transcription was inferred at protein-coding genes bidirectional promoters, the neighboring 591 protein-coding genes also showed increased expression divergence, compared to genes that are 592 transcribed from conserved IncRNA promoters. We thus confirm that IncRNA turnover is associated 593 with changes in neighboring gene expression (Kutter et al. 2012). While IncRNAs changes may be 594 directly affecting gene expression, another probable hypothesis is that a common mechanism affects 595 both IncRNAs and protein-coding genes transcribed from bidirectional promoters.

#### **Conclusions**

597 Our comparative transcriptomics approach confirms that IncRNAs repertoires, sequences and 598 expression patterns evolve rapidly across species, and shows that accelerated rates of IncRNA 599 evolution are also seen in developmental transcriptomes, albeit less frequently. These observations are consistent with the hypothesis that the majority of IncRNAs (or at least of those detected with 600 601 sensitive transcriptome sequencing approaches, in particular in the adult testes) may be non-602 functional. However, we are able to modulate this conclusion, by showing that there are increased 603 levels of functional constraint on IncRNAs expressed during embryonic development, in particular in 604 the developing brain and kidney. These increased levels of constraint apply to all analyzed aspects of 605 IncRNAs, including sequence conservation for exons, promoter and splice sites, but also expression

606 pattern conservation. For many of these loci, biological function may be RNA-independent, as the 607 highest levels of selective constraint are observed on promoter regions and on splice signals, rather 608 than on lncRNA exonic sequences. Our results are thus compatible with unconventional, RNA-609 independent functions for lncRNAs expressed during embryonic development.

#### 610 Materials and methods

# 611 Biological sample collection

612 We collected samples from three species (mouse C57BL/6J strain, rat Wistar strain and chicken 613 White Leghorn strain), four organs (brain, kidney, liver and testes) and five developmental stages 614 (including two embryonic stages, newborn, young and aged adult individuals). We sampled the 615 following stages in the mouse: embryonic day post-conception (dpc) 13.5 (E13.5 dpc, hereafter mid-616 stage embryo); E17 to E17.5 dpc (late embryo); post-natal day 1 to 2 (newborn); young adult (8-10 617 weeks old); aged adult (24 months old). For the rat, we sampled the following stages: E15 dpc (mid-618 stage embryo); E18.5 to E19 dpc (late embryo); post-natal day 1 to 2 (newborn); young adult (8-10 619 weeks old); aged adult (24 months, with the exception of kidney samples and two of four liver samples, 620 derived from 12 months old individuals). The embryonic and neonatal developmental stages were 621 selected for maximum comparability based on Carnegie stage criteria (Theiler 1989). For chicken, we 622 collected samples from Hamburger-Hamilton stages 31 and 36 (hereafter termed mid-stage and late 623 embryo), selected for comparability with the two embryonic stages in mouse and rat (Hamburger and 624 Hamilton 1951). Each sample corresponds to one individual, except for mouse and rat mid-stage 625 embryonic kidney, for which tissue from several embryos was pooled prior to RNA extraction. For adult 626 and aged organs, multiple tissue pieces from the same individual were pooled and homogenized prior 627 to RNA extraction. For brain dissection, we sampled the cerebral cortex. For mouse and rat, with the 628 exception of the mid-stage embryonic kidney, individuals were genotyped and males were selected 629 for RNA extraction. Between two and four biological replicates were obtained for each 630 species/organ/stage combination, amounting to 97 samples in total (Supplementary Table 1).

631

# RNA-seq library preparation and sequencing

We performed RNA extractions using RNeasy Plus Mini kit from Qiagen. We assessed RNA quality with the Agilent 2100 Bioanalyzer. RNA integrity numbers (RIN) are available in Supplementary Table 1; see Supplementary Methods for additional RNA integrity analyses. Sequencing libraries were produced with the Illumina TruSeq stranded mRNA protocol with polyA selection, and sequenced as 101 base pairs (bp) single-end reads, at the Genomics Platform of iGE3 and the University of Geneva. Libraries are strand-specific and the sequenced strand is complementary to the RNA molecule.

#### 638 <u>Additional RNA-seq data</u>

To improve detection power for lowly expressed IncRNAs, we complemented our RNA-seq collection with samples generated with the same technology for Brown Norway rat adult organs (Cortez et al. 2014). We added published data for adult chicken (red jungle fowl strain UCD001) organs (McCarthy et al. 2019), as well as for embryonic chicken (White Leghorn) organs (Uebbing et al. 2015; Ayers et al. 2013). As the data were not comparable with our own in terms of library preparation and animal strains, these samples were only used to increase IncRNA detection sensitivity.

# 645 <u>RNA-seq data processing</u>

646 We used HISAT2 (Kim et al. 2015) release 2.0.5 to align the RNA-seq data on reference 647 genomes. The genome sequences (assembly versions mm10/GRCm38, rn6/Rnor\_6.0 and 648 galGal5/Gallus\_gallus-5.0) were downloaded from the Ensembl database (Cunningham et al. 2019). Genome indexes were built using only genome sequence information. To improve detection 649 650 sensitivity, at the alignment step we provided known splice junction coordinates extracted from 651 Ensembl. We set the maximum intron length for splice junction detection at 1 million base pairs (Mb). 652 The following command-line arguments were used: --rna-strandness R --known-splicesiteinfile=SpliceSites Ensembl.txt --max-intronlen 1000000 --dta-cufflinks, where SpliceSites Ensembl.txt 653 corresponds to the exon junction coordinates extracted with hisat2\_extract\_splice\_sites.py. See also 654 655 Supplementary Methods for additional RNA-seq data quality analyses.

#### 656 Trans

# Transcript assembly and filtering

657 We assembled transcripts for each sample using StringTie (Pertea et al. 2015), release 1.3.5, 658 based on read alignments obtained with HISAT2. We provided genome annotations from Ensembl 659 release 94 as a guide for transcript assembly. We filtered Ensembl annotations to remove transcripts 660 that spanned a genomic length above 2.5 Mb. For protein-coding genes, we kept only protein-coding transcripts, discarding isoforms annotated as "retained\_intron", "processed\_transcript" etc. We set 661 662 the minimum exonic length at 150 bp, the minimum anchor length for splice junctions at 8bp and the 663 minimum isoform fraction at 0.05. The following StringTie command-line arguments were used: -G 664 Ensembl annotations.gtf -m 150 -a 8 -f 0.05 -p 8 -rf, where Ensembl annotations.gtf correspond to 665 the Ensembl annotations filtered as described above. We compared the resulting assembled 666 transcripts with Ensembl annotations and we discarded read-through transcripts, overlapping with 667 multiple multi-exonic Ensembl-annotated genes. For strand-specific samples, we discarded transcripts 668 for which the ratio of sense to antisense unique read coverage was below 0.01. We discarded multi-669 exonic transcripts that were not supported by splice junctions with correctly assigned strands. The 670 filtered transcripts obtained for each sample were assembled into a single dataset per species using 671 the merge option in StringTie. For increased sensitivity, we removed the minimum FPKM and TPM

thresholds, but required a minimum isoform fraction of 0.05 for transcript inclusion. The following
StringTie command-line arguments were used: stringtie -v --merge -G Ensembl\_annotations.gtf -m 150
-a 8 -p 8 -F 0 -T 0 -f 0.05. We constructed a combined annotation dataset, starting with Ensembl
annotations, to which we added newly-assembled transcripts that had no exonic overlap with Ensembl
genes. We also included newly-annotated isoforms for known genes if they had exonic overlap with
exactly one Ensembl gene, thus discarding potential read-through transcripts or gene fusions.

#### 678

# Protein-coding potential of assembled transcripts

679 To determine whether the newly assembled transcripts were protein-coding or non-coding, 680 we mainly relied on the codon substitution frequency (CSF) score (Lin et al. 2007). As in a previous 681 publication (Necsulea et al. 2014) we scanned whole genome alignments and computed CSF scores in 682 75 bp sliding windows moving with a 3 bp step. We used pre-computed alignments downloaded from the UCSC Genome Browser (Casper et al. 2018), including the alignment between the mouse genome 683 684 and 59 other vertebrates (for mouse classification), between the human genome and 99 other 685 vertebrates (for rat and chicken classification) and between the rat genome and 19 other vertebrates 686 (for rat classification). For each window, we computed the score in each of the 6 possible reading 687 frames and extracted the maximum score for each strand. We considered that transcripts are protein-688 coding if they overlapped with positive CSF scores on at least 150 bp. As positive CSF scores may also 689 appear on the antisense strand of protein-coding regions due to the partial strand-symmetry of the 690 genetic code, in this analysis we considered only exonic regions that did not overlap with other genes. 691 In addition, we searched for sequence similarity between assembled transcripts and known protein 692 sequences from the SwissProt 2017\_04 (The UniProt Consortium 2017) and Pfam 31.0 (El-Gebali et al. 693 2019) databases. We kept only SwissProt entries with confidence scores 1, 2 or 3 and we used the 694 Pfam-A curated section of Pfam. We searched for sequence similarity using the blastx utility in the 695 BLAST+ 2.8.1 package (Camacho et al. 2009; Altschul et al. 1990), keeping hits with maximum e-value 1e<sup>-3</sup> and minimum protein sequence identity 40%, on repeat-masked cDNA sequences. We considered 696 697 that transcripts were protein-coding if they overlapped with blastx hits over at least 150 bp. Genes 698 were said to be protein-coding if at least one of their isoforms was classified as protein-coding, based 699 on either the CSF score or on sequence similarity with known proteins.

700

# Long non-coding RNA selection

To construct a reliable lncRNA dataset, we selected newly-annotated genes classified as noncoding based on both the CSF score and on sequence similarity with known proteins and protein domains, as well as Ensembl-annotated genes with non-coding biotypes ("lincRNA", "processed\_transcript", "antisense", "TEC", "macro\_lncRNA", "bidirectional\_promoter\_lncRNA", "sense\_intronic"). For newly detected genes, we applied several additional filters: we required a 706 minimum exonic length (corresponding to the union of all annotated isoforms) of at least 200 bp for 707 multi-exonic loci and of at least 500 bp for mono-exonic loci; we eliminated genes that overlapped for 708 more than 5% of their exonic length with unmappable regions; we kept only loci that were classified 709 as intergenic and at least 5 kb away from Ensembl-annotated protein-coding genes on the same strand; 710 for multi-exonic loci, we required that all splice junctions be supported by reads with correct strand 711 assignment (cf. above). For both *de novo* and Ensembl annotations, we removed transcribed loci that 712 overlapped on at least 50% of their length with retrotransposed gene copies, annotated by the UCSC 713 Genome Browser and from a previous publication (Carelli et al. 2016); we discarded loci that 714 overlapped with UCSC-annotated tRNA genes and with RNA-type elements from RepeatMasker (Smit 715 et al. 2003) on at least 25% of their length. We kept loci supported by at least 10 uniquely mapped 716 RNA-seq reads and for which a ratio of sense to antisense transcription of at least 1% was observed in 717 at least one sample. Although the fraction of reads stemming from the wrong strand due to errors in 718 library preparations is very low in our samples (Supplementary Table 1), loci situated on the antisense 719 strand of highly expressed genes can have unreliable expression estimates. Thus, for loci that had 720 sense/antisense exonic overlap with other genes, we computed expression levels either on complete 721 gene annotations, or only on exonic regions that had no overlap with other genes, and computed 722 Spearman's correlation coefficient between the two expression estimates, across all samples. We 723 discarded loci for which the correlation coefficient was below 0.9. Full gene annotations and IncRNA 724 selection criteria are provided in Supplementary Dataset 1 online.

#### 725

## Gene expression estimation

726 We computed the number of uniquely mapping reads unambiguously attributed to each gene 727 using the Rsubread package in R (Liao et al. 2019), discarding reads that overlapped with multiple 728 genes. We also estimated read counts and TPM (transcript per million) values per gene using Kallisto 729 (Bray et al. 2016). To approach absolute expression levels estimates, for better comparisons across 730 samples, we further normalized TPM values using a scaling approach (Brawand et al. 2011). Briefly, we 731 ranked the genes in each sample according to their TPM values, we computed the variance of the ranks across all samples for each gene, and we identified the 100 least-varying genes, found within the inter-732 733 quartile range (25%-75%) in terms of average expression levels across samples. We derived 734 normalization coefficients for each sample such that the median of the 100 least-varying genes be 735 identical across samples. We then used these coefficients to normalize TPM values for each sample. We excluded mitochondrial genes from expression estimations and analyses, as these genes are highly 736 737 expressed and can be variable across samples. For differential expression analyses, we used per-gene 738 unique read counts computed with Rsubread. For all downstream analyses we used normalized TPM

values. When indicated, we transformed TPM values with the following formula: x->log2(x+1). Gene
expression data is available in Supplementary Dataset 2 online.

741 Differential

# Differential expression analyses

742 We used the DESeq2 (Love et al. 2014) package release 1.22.2 in R release 3.5.0 (R Core Team 743 2018) to test for differential expression across developmental stages, separately for each organ and 744 species. We analyzed both protein-coding genes and IncRNAs, selected according to the criteria 745 described above. We first performed a global differential expression analysis, using the likelihood ratio 746 test to contrast a model including an effect of the developmental stage against the null hypothesis of 747 homogeneous expression across all developmental stages. This analysis was performed on all protein-748 coding and lncRNA genes for each species, as well as on 1-to-1 orthologues for mouse and rat. In 749 addition, we down-sampled the numbers of reads assigned to protein-coding genes to obtain identical 750 average numbers of reads for protein-coding genes and IncRNAs. The resampled read counts were 751 directly proportional to the original counts for each protein-coding gene. We also contrasted 752 consecutive developmental stages, for each species and organ, using the Wald test implemented in 753 DESeq2. Differential expression results are available in Supplementary Dataset 4 online.

754

#### Homologous IncRNA family prediction

755 We used existing whole-genome alignments as a guide to predict homologous lncRNAs across 756 species, as previously proposed (Washietl et al. 2014). We first constructed for each gene the union of 757 its exon coordinates across all isoforms, hereafter termed "exon blocks". We projected exon block 758 coordinates between pairs of species using the liftOver utility and whole-genome alignments 759 generated with blastz (http://www.bx.psu.edu/miller\_lab/), available through the UCSC Genome 760 Browser (Casper et al. 2018). To increase detection sensitivity, for the initial liftOver projection we 761 required only that 10% of the reference bases remap on the target genome. Projections were then 762 filtered, retaining only cases where the size ratio between the projected and the reference region was 763 between 0.33 and 3 for mouse and rat (0.2 and 5 for comparisons involving chicken). To exclude recent 764 lineage-specific duplications, regions with ambiguous or split liftOver projections were discarded. For 765 genes where multiple exon blocks could be projected across species, we defined the consensus 766 chromosome and strand in the target genome and discarded projected exon blocks that did not match 767 this consensus. We then evaluated the order of the projected exon blocks on the target genes, to 768 identify potential internal rearrangements. If internal rearrangements were due to the position of a 769 single projected exon block, the conflicting exon block was discarded; otherwise, the entire projected 770 gene was eliminated. As the projected reference gene coordinates could overlap with multiple genes 771 in the target genome, we constructed gene clusters based on the overlap between projected exon 772 block coordinates and target annotations, using a single-link clustering approach. We then realigned

773 entire genomic loci for each pair of reference-target genes found within a cluster, using lastz 774 (http://www.bx.psu.edu/miller\_lab/) and the threaded blockset aligner TBA (Blanchette et al. 2004). 775 Using this alignment, we computed the percentage of exonic sequences aligned without gaps and the 776 percentage of identical exonic sequence, for each pair of reference-target genes. We then extracted 777 the best hit in the target genome for each gene in the reference genome based on the percentage of 778 identical exonic sequence, requiring that the ratio between the maximum percent identity and the 779 percent identity of the second-best hit be above 1.1. Reciprocal best hits were considered to be 1-to-780 1 orthologous loci between pairs of species. For analyses across all three species, we constructed 781 clusters of reciprocal best hits from pairwise species comparisons, using a single-link clustering 782 approach. Resulting clusters with more than 1 representative *per* species were discarded. The results 783 of the homology prediction pipeline, sequence alignment statistics and Ensembl orthology 784 relationships for protein-coding genes are available in Supplementary Dataset 5 online.

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#### Availability of data and materials

786 The RNA-seq data were submitted to the NCBI Gene Expression Omnibus (GEO), under

787 accession number GSE108348. Supplementary Datasets containing additional processed data are

788 available at the address: <u>ftp://pbil.univ-lyon1.fr/pub/datasets/Darbellay\_LncEvoDevo</u>

789 The scripts used to analyse the data are available through a GitHub repository:

790 <u>https://github.com/anecsulea/LncEvoDevo</u>. See also Supplementary Methods.

# 791 Author contributions

FD performed organ dissections, RNA extractions, quality control, prepared samples for sequencing and contributed to study design and manuscript preparation. AN designed the study, performed computational analyses and wrote the manuscript.

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#### Figure legends

Figure 1. Comparative transcriptomics across species, organs and developmental stages.

A. Experimental design. The developmental stages selected for mouse, rat and chicken are
 marked on a horizontal axis. Organs sampled for each species and developmental stage are shown
 below. Abbreviations: br, brain; kd, kidney; lv, liver; ts, testes.

- B. First factorial map of a principal component analysis, performed on log2-transformed TPM
   values, for 10,363 protein-coding genes with orthologues in mouse, rat and chicken. Colors represent
   different organs and developmental stages, point shapes represent different species.
- C. Hierarchical clustering, performed on a distance matrix derived from Spearman correlations
   between pairs of samples, for 10,363 protein-coding genes with orthologues in mouse, rat and chicken.
   Organ and developmental stages are color-coded, shown below the heatmap. Species of origin is color coded, shown on the right. Sample clustering is shown on the left.
- 820

#### Figure 2. Transcriptome complexity in different species, organs and developmental stages.

A. Number of protein-coding genes supported by at least 10 uniquely mapped reads in each sample, after read resampling to homogenize coverage (Supplementary Methods). Colors represent different organs, point shapes represent different species. Developmental stages are indicated by numeric labels, 1 to 5, on the X-axis. We analyzed a total of 19,356 protein-coding genes in the mouse, 19,274 in the rat and 15,509 in the chicken.

D. Number of IncRNAs supported by at least 10 uniquely mapped reads in each sample, after
 read resampling to homogenize coverage. We analyzed a total of 18,858 candidate IncRNAs in the
 mouse, 20,159 in the rat and 5,496 in the chicken.

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#### Figure 3. Different expression patterns for protein-coding genes and lncRNAs.

A. Distribution of the organ in which maximum expression is observed, for protein-coding genes (pc) and IncRNAs (Inc), for mouse, rat and chicken. Organs are color-coded, shown above the plot. We defined the sample in which maximum expression is observed based on average expression values across replicates, for each organ / developmental stage combination (Supplementary Methods).

B. Distribution of the developmental stage in which maximum expression is observed, for
protein-coding genes and lncRNAs, for mouse, rat and chicken. Developmental stages are color-coded,
shown above the plot.

C. Percentage of protein-coding and IncRNA genes that are significantly (FDR<0.01)</li>
 differentially expressed (DE) among developmental stages, with respect to the total number of genes
 tested for each organ. Left panel: differential expression analysis performed with all RNA-seq reads.

Right panel: differential expression analysis performed after down-sampling read counts for protein-coding genes, to match those of lncRNAs (Materials and methods).

D. Distribution of the developmental stage in which maximum expression is observed, for
 protein-coding genes and IncRNAs that are significantly DE (FDR<0.01) in each organ, for the mouse.</li>
 Percentages are computed with respect to the total number of DE genes in each organ and gene class.

Figure 4. Increased levels of long-term sequence conservation for IncRNAs expressed early
 in development.

A. Sequence conservation scores (PhastCons scores, placental mammals) for protein-coding and lncRNAs exonic regions, for subsets of genes expressed above noise levels (TPM>=1) in each organ and developmental stage. Dots represent medians, vertical bars represent 95% confidence intervals. Numbers of analyzed genes are provided in Supplementary Table 5. Organs are color-coded; developmental stages are indicated (numbers 1 to 5) on the X-axis. The gray dots and vertical bars represent the median value and 95% confidence interval for all intergenic regions, genome-wide, or for intergenic regions flanking lncRNA loci (Supplementary Methods).

857 B. Same as A, for promoter regions (400 bp upstream of transcription start sites). Exonic
858 sequences were masked before assessing conservation.

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**C.** Same as B, for splice sites (first and last two bases of each intron).

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# Figure 5. Orthologous IncRNA families for mouse, rat and chicken.

862 A. Number of mouse protein-coding genes and IncRNAs in different classes of evolutionary 863 conservation. From left to right: all loci, loci with conserved sequence in the rat, loci for which 864 transcription could be detected (at least 10 unique reads) in predicted orthologous locus in the rat, 865 loci with predicted 1-to-1 orthologues, loci for which the predicted orthologue belonged to the same 866 class (protein-coding or IncRNA) in the rat, loci with conserved sequence in the chicken, loci for which transcription could be detected (at least 10 unique reads) in predicted orthologous locus in the 867 868 chicken, loci with predicted 1-to-1 orthologues, loci for which the predicted orthologue belonged to 869 the same class (protein-coding or lncRNA) in the chicken. We analyze 19,356 protein-coding genes and 870 18,858 candidate IncRNAs in the mouse.

**B.** Distribution of the organ in which maximum expression is observed, for mouse proteincoding and lncRNA genes that have no orthologues in the rat or chicken, for genes with orthologues in the rat and for genes with orthologues in chicken. The sample in which maximum expression is observed is computed based on average expression values across biological replicates, for each organ / developmental stage combination (Supplementary Methods).

876 C. Same as B, for the distribution of the developmental stage in which maximum expression is877 observed.

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#### Figure 6. Global comparison of IncRNA expression patterns across species.

A. First factorial map of a principal component analysis, performed on log2-transformed TPM
 values, for 2,893 orthologous lncRNAs between mouse and rat. Colors represent different organs and
 developmental stages, point types represent species.

883 **B.** Hierarchical clustering, performed on a distance matrix derived from Spearman correlations 884 between pairs of samples, for 2,893 orthologous lncRNAs between mouse and rat. Organ and 885 developmental stages are shown below the heatmap. Species of origin is shown on the right. Sample 886 clustering is shown on the left.

887

Figure 7. Global estimates of expression conservation across organs and developmentalstages.

A. Example of between-species and within-species variation of expression levels, for proteincoding genes (left) and lncRNAs (right), for orthologous genes between mouse and rat, for the midstage embryonic brain. Spearman's correlation coefficients (rho) are shown above each plot. We show a smoothed color density representation of the scatterplots, obtained through a 2D kernel density estimate (smoothScatter function in R).

895 **B.** Expression conservation index, defined as the ratio of the between-species and the within-896 species expression level correlation coefficients, for protein-coding genes, for each organ and 897 developmental stage. The vertical segments represent minimum and maximum values obtained from 898 100 bootstrap replicates. We analyzed 15,931 pairs of orthologous protein-coding genes.

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**C.** Same as B, for IncRNAs. We analyzed 2,893 orthologous mouse and rat IncRNAs.

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Figure 8. Conservation of developmental expression patterns of differentially expressed
 IncRNAs.

903 A. Comparison of the developmental stage in which maximum expression is observed, for 904 orthologous IncRNAs that are significantly differentially expressed (FDR<0.01) among developmental 905 stages, for both mouse and rat. The sample in which maximum expression is observed is computed 906 based on average expression values across biological replicates, for each organ / developmental stage 907 combination (Supplementary Methods). Genes are divided into classes based on the developmental 908 stage where maximum expression is observed in mouse organs (X-axis). The Y axis represents the 909 percentage of orthologous genes that reach maximum expression in each developmental stage, in the 910 rat. Numbers of analyzed genes are shown below the plot.

B. Expression profiles of orthologous IncRNAs that are significantly differentially expressed
(FDR<0.01) among developmental stages, for both mouse and rat, in the brain. TPM values were</li>
averaged across replicates and normalized by dividing by the maximum, for each species. The resulting
relative expression profiles were combined across species and clustered with the K-means algorithm.
Dots represent the average profiles of the genes belonging to each cluster. Gray lines represent profiles
of individual genes from a cluster. Numbers of genes in each cluster are shown in the plot.

917 **C.** Same as B, for the kidney.

918 **D.** Same as B, for the liver.

919 E. Same as B, for the testes. For this organ, we searched for only 4 clusters with the K-means920 algorithm.

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# Figure 9. Per-gene estimates of expression pattern divergence between species.

A. Relationship between the per-gene expression divergence measure (Euclidean distance of relative expression profiles among organs/stages, between mouse and rat), and the average expression values (log2-transformed TPM) across all mouse and rat samples. We show a smoothed color density representation of the scatterplots, obtained through a 2D kernel density estimate (smoothScatter function in R). Red line: linear regression.

928 **B.** Distribution of the expression divergence value for all protein-coding and lncRNA genes with 929 predicted 1-to-1 orthologues in mouse and rat.

930 C. Distribution of the residual expression divergence values, after regressing the average931 expression level, for protein-coding genes and lncRNAs.

932 D. Relationship between expression divergence and exonic sequence conservation (% exonic
 933 sequence aligned without gaps between mouse and rat), for protein-coding genes and lncRNAs.

934 E. Average contribution of each organ/developmental stage combination to expression935 divergence, for protein-coding genes and lncRNAs.

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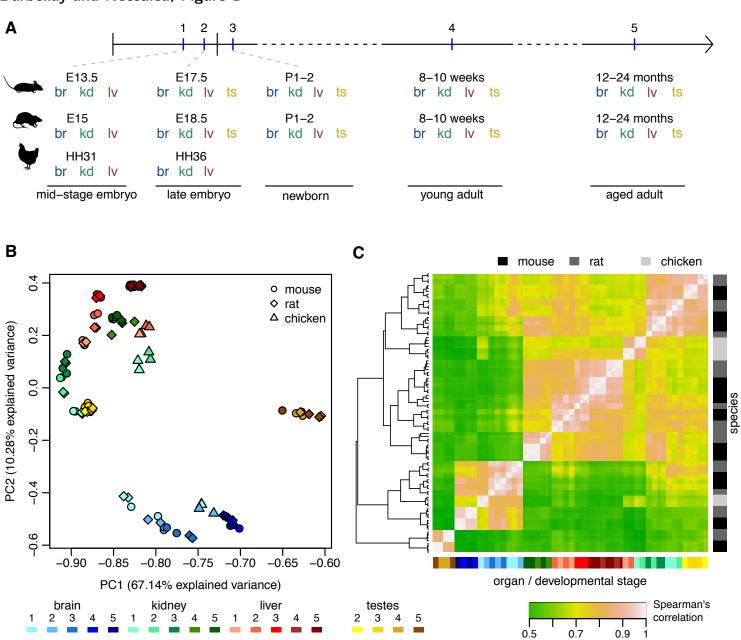
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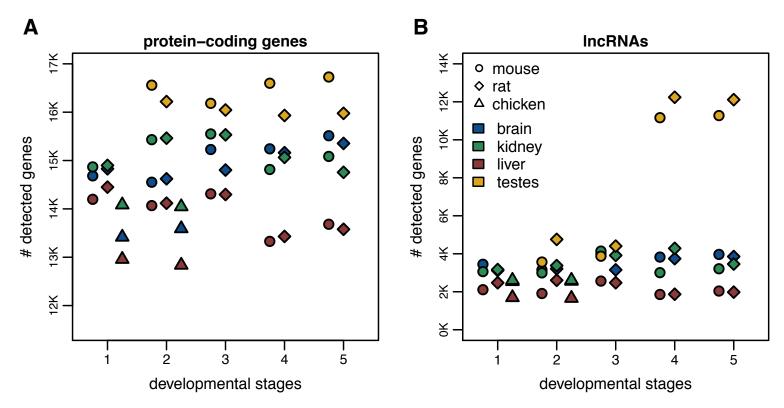
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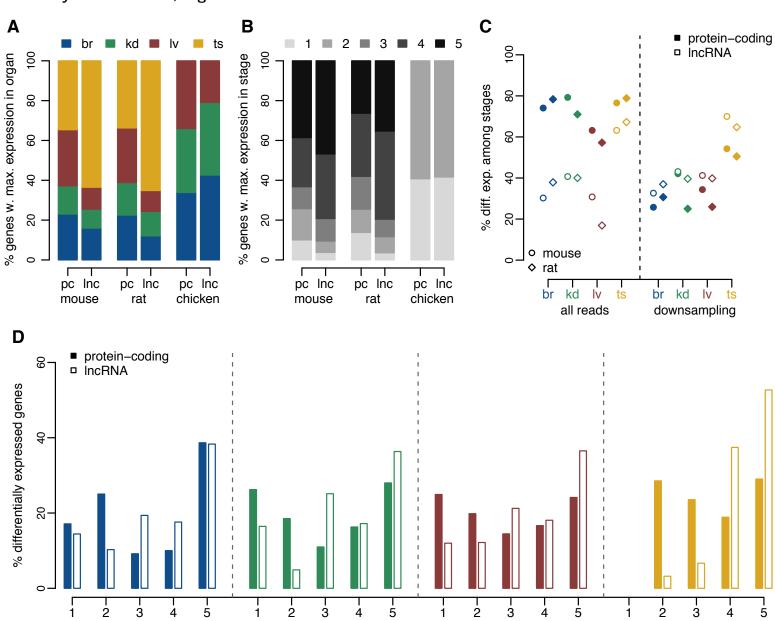
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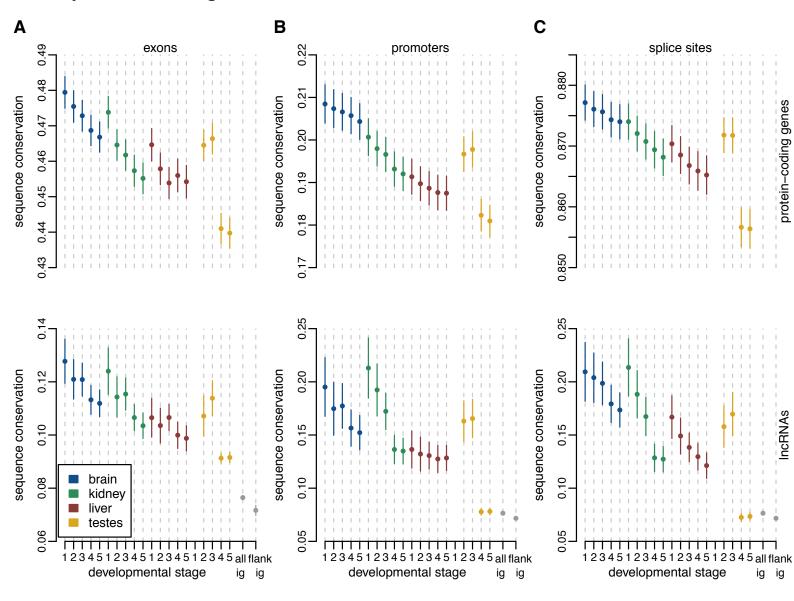
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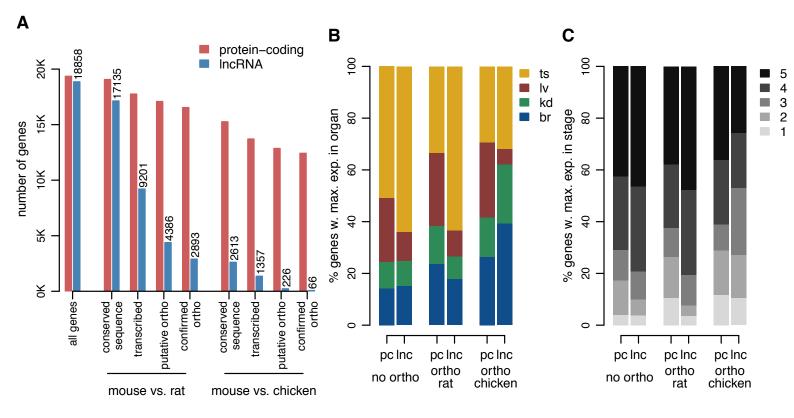


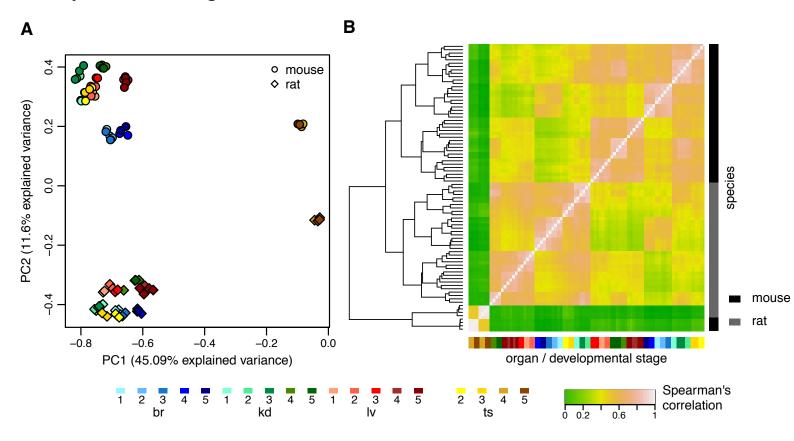




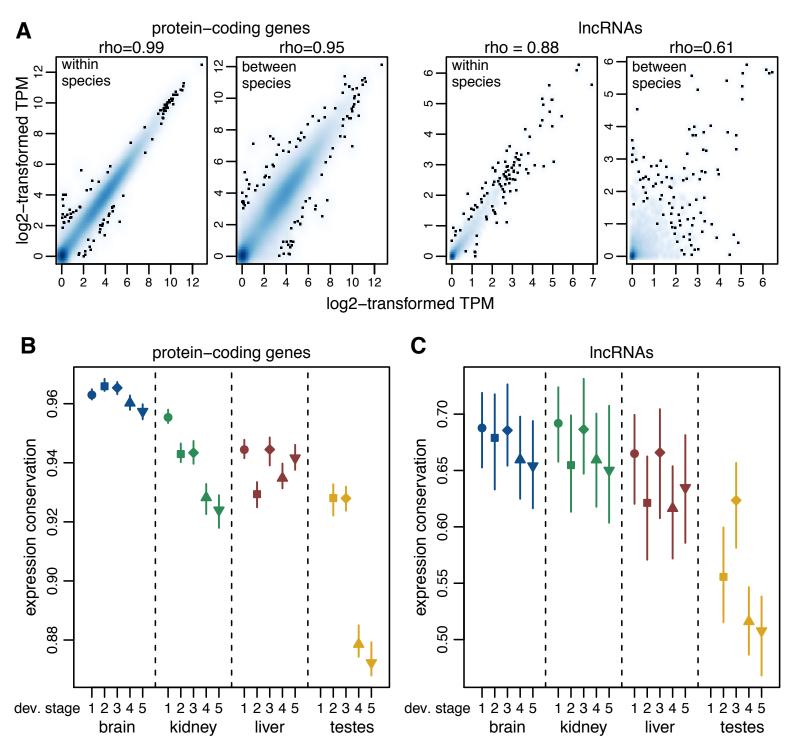
developmental stage with maximum expression, mouse genes



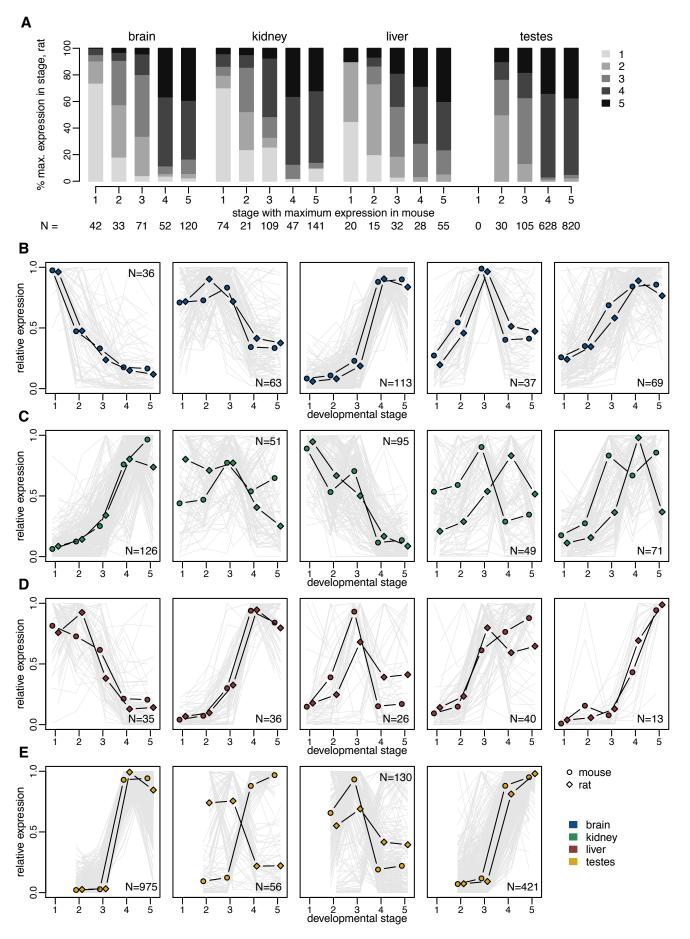


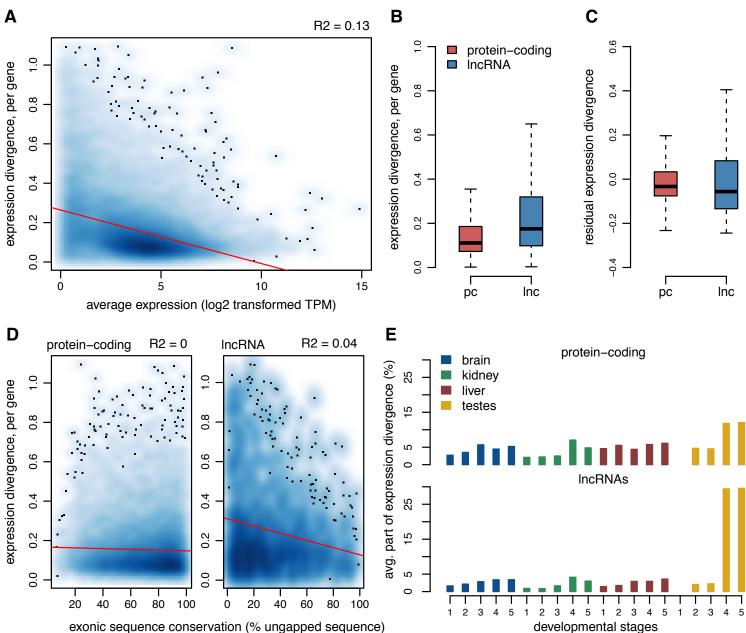


Darbellay and Necsulea, Figure 7









exonic sequence conservation (% ungapped sequence)