



An atlas of chicken long non-coding RNAs gathering multiple sources: gene models and expression across more than twenty tissues

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Rennes, France





Surday, January 13° 2019
Non-coding RNA workshop – Plant and Animal Genome XXVII
San Diego, CA





Fundamental goal #1: improve our knowledge of the chicken's genes at the genom

. What is the state of the knowledge of the chicken's genes vs. human's ?

Chicken:
 → ~ 20 000 protein coding genes
 → ~ 2 000 small non-coding genes
 → ~ 4 600 long non-coding genes

2 v94. December 2018

⇒ our knowledge of chicken (long)ncRNA is still incomplete

in term of number (low expression ⇒ high sequencing depth in numerous tissues)
 in term of isoforms (similarly to coding genes, requires technologies ⇒ long reads)

Human :

→ ~ 20 000 protein coding genes

→ ~ 5 000 small non-coding genes

→ ~ 15 000 long non-coding genes

⇒ first goal: improve our knowledge of the chicken's long non-coding genes at the genome scale in term of number of loci, using a high number of short reads (364 samples × 90M reads)

Context: IncRNAs, from cellular processes to complex traits variation

Long non-coding RNAs (lncRNAs) - transcripts longer than 200 nucleotides that are not translated into proteins

- IncRNAs have been showed to act on numerous cellular processes:
 chromatin compaction (Xist)

 - Chloridati Curipaciani (Analy
 gene expression :
 → transcription (Fendir, Grote et al., 2013)
 → translation (lincRNA-p21, Yoon et al., 2012)
- Influence on complex traits & diseases ?
 General agreement that complex traits and diseases are influenced by numerous loci ⇒ 90% cut of coding regions (Manolio et al., 2009)
 These loci are located in regulatory regions (Maurano et al., 2012)

 - ⇒ effect on genes expression
 Some of them affect IncRNA expression (Kumar et al., 2013)
- Numerous, poor functional annotation:
 no function prediction rules (yet)
 lowly expressed (10-fold less than mRNAs): difficult to validate experimentally

Fundamental goal #2: annotate IncRNAs thanks to their expression profile in chicken focus on tissue-specificity

Problematic: knowing the existence of a gene (coding or non-coding) doesn't inform on its function(s)

- with coding genes:
 prediction of the associated amino-acid sequence,
 prediction of the protein function by motif conservation across species.
- o with long non-coding genes:
- poor sequence conservation, poor / no domains knowledge.
- ⇒ IncRNA are poorly annotated

⇒ second goal: annotate the lncRNAs applying the two following strategies

Strategies to annotate IncRNAs function:

expression profile: IncRNA expressed in tissue A and little or not in tissue B (= specific to tissue A) might have a role related to tissue A function



study their configuration with closest coding gene: e.g., close and divergent ⇒ common regulation ?



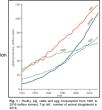
Context: the chicken (G. gallus), a specie of great economical and social important

- Meat : currently, one of the most produced and consumed meat in the world ⇒ almost 120 millions tonnes produced in 2016
- Eggs: good sources of proteins, fatty acids and micronutrients for human co

 Easy to produce and store, no cultural restriction for consumption

 In 2011, −200 per year per capita (IEC, 2011)

 ⇒ almost 70 millions tonnes produced in 2016



1. Improvement of our knowledge of chicken's genes at the genome scale

- Strategy:
 exploit the volume of expression data from INRA's lab to modelize yet-undiscovered genes, in complement to Ensembl genes
- extension of this catalogue using external sources



Data 🏥 INRA

- 364 RNA-seq samples from 3 tissues:
 adipose tissue (56 samples), blood (128 samples) and liver (180 samples)







- 90 millions reads per sample
 stranded reads
 2 × 150 pb paired-ends

- memonos

 Mapping on G. gallus reference genome (GalGalS) with STAR (Dobin et al., 2013),
 using Ensembl v94 annotation

 Transcriptione reconstruction using Cufflink2 (Trapnell et al., 2013)

 Elimination of modelized genes overlapping on the same strand any Ensembl v94 gene
 incRNA prediction using FEEInc (Wucher et al., 2017)

Results: addition of 25 214 genes to the 24 881 Ensembl genes

⇒ Total of 50 095 genes



1. Selection of the most robust modeled genes based on their expression pattern

Reminder: lncRNA = lowly expressed (1/10th mRNA). Necessity to separate background noise from real expression.

To improve reliability of the 25 214 newly modelized genes, we selected models according to their expression, y 2 expression criteria:

an normalized expression metric (TPM > 0.1)

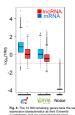
number of reads supporting the model

selection of 58.5% of the 25 214 genes (14 760 models)

Comparison with background noise:

"No genes" = set of genomic regions out of the Ensembl and our genes models: our models have expression levels supported to the background pains.

Comparison with mRNAs and Ensembl genes:



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Gene type	Number of genes	% of total	Ensembl	
IncRNAs	13 009	88.14%	4 641	
mRNAs	1 199	8.12%	18 346	
Others	552	3.74%	1 894	
TOTAL	14 760	100%	24 881	

1. Sequential building of an extended annotation using 6 external data source

- From this set of 14 760 INRA genes (88% IncRNA) + 24 881 Ensembl genes, ⇒ extension of the chicken IncRNA annotation using external sources: 1. NONCODE (Fang et al., 2018) + 93 22 IncRNAs 2. NGBI: 5 738 IncRNAs ALBE (Lit et al., 2015): 5 752 IncRNAs

- 3. ALDB (Li et al., 2015): 5 /52 Incrinas 4. FR-AGENCODE (Foissac et al., submitted): 6 089 IncRNAs

For each source, we kept only the loci with no overlap with the extended annotation:



- As a result, we increased gene numbers from 14 760 + 24 881 to →→ →→ 52 075 genes including
 30 084 IncRNAs: ×8.5 compared to Ensembl alone
 19 545 RRNAs: ×1.1 compared to Ensembl alone
- This work was done at the locus level since isoforms are poorly known and expression analysis are done at the gene level

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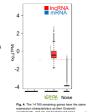
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2. Annotation of the IncRNAs thanks to their expression profile in chicken, focus on



Thanks to the previous step:

☑ extended IncRNA catalogue (from 4 640 to 30 084)

☑ role or function of the IncRNAs?

objective of this second part: provide an annotation of the potential IncRNA functions using their expression profiles

2. Annotation of the IncRNAs thanks to their expression profile in chicken, focus or

Hypothesis: if a gene is expressed in one or a few tissues, its function is likely to be related to the function of the

To study tissue-specificity, necessity to have numerous tissues ⇒ public data from other teams.

3 datasets were analyzed:

- INRA: 5 tissues (Jehl et al, in preparation)
 Sishuan University: 6 tissues (Inga et al., 2017)
 Roslin Institute: 21 tissues (used by Ensembl for the chicken genome annotation)

▲ IncRNAs are lowly expressed ⇒ sensitivity to genotype, physiological status, experimental conditions, ...

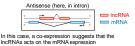
Preliminary question: can these 3 datasets be studied together after expression normalization, or is there a batch

Long non-coding genes are classified by position relatively to the nearest coding genes using FEELnc (Wucher et al., 2017)

as some configurations suggest potential (co)-regulations

Divergent

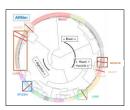
In this case, if the genes are close, a co-expression suggests a common regulation, and therefore a common function

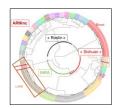


In both case, we infer the function of the IncRNA using the function of

	Divergent	Antisense of intron
In the extended catalogue	4 895	2 105
with $\tau \ge 0.95$	977	418

- The mRNAs expression clusters the sample based on tissues ar The IncRNAs expression clusters the sample mainly by projects ession clusters the sample based on tissues and functions





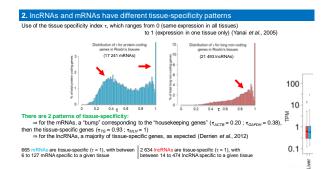
⇒ For the IncRNAs, combining different datasets is not pr
⇒ focus on the 21 tissues from the Roslin Institute data

- We provide a large catalogue of chicken lncRNAs at the gene level from 4 640 (Ensembl v94) to 30 084 lncRNAs
- We also provided a rough annotation of all these genes, based on a
- their expression pattern across 21 chicken tissues their position relative to the nearest coding gene
- ⇒ this will allow the scientific community to work on different scientific problematics related to IncRNAs in chicken

- Perspective:

 we will use these data to study the genetic component of feed efficiency in layer chicken
 => feed =60% of production cost in monogastrics
 => reduction of feed vs. food competition
 >> lower environmental impact











Thank you for your attention!

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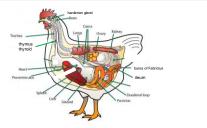
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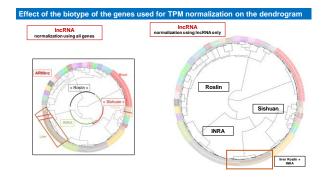
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Tissues present in Roslin Institute data

bursa of Fabric	ius thyroid	duodenum	proventriculus	trachea	breast muscle	kidney	
harderian glan	d caecal tonsils	ovary	skin	lung	optical lobe	liver	
thymus	ileum	pancreas	fat gizzard	heart	cerebellum	spleen	





1. Selection of the most robust modeled genes based on their expression pattern

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- To improve reliability of the 25 214 newly modelized genes, we selected models using two criteria:

 an expression metric (TPM)

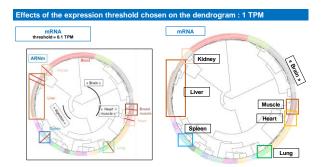
 number of supporting reads

5 paired-reads or more in 25% of the samples of a tissue: at least 75 supporting reads.
 ⇒ leaves 14 760 modelized genes from the 25 214 (= 58.5%).

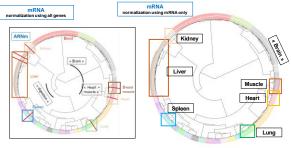
Number of IncRNAs and mRNAs in the remaining catalogue

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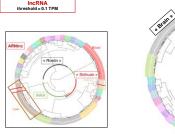
- INRA genes + Ensembl genes = 14 760 + 24 881 = **39 641 genes**

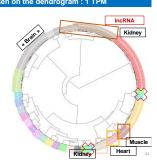


Effect of the biotype of the genes used for TPM normalization on the dendrogram



Effects of the expression threshold chosen on the dendrogram : 1 TPM





A tissue-specificity metrics: the tau (τ)

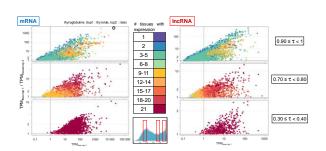
To study tissue-specificity, we used the tau metrics from Yanai et al., 2005:

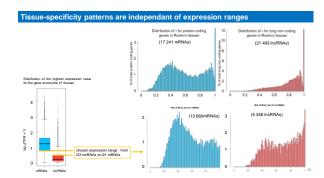
- associates a value (τ) to each gene
- of or each gene, it accounts for:

 the expression in each tissue, relatively to the tissue with the highest expression

 the number of tissues with expression
 - the number of tissues with expression
 the total number of tissues
- value from 0 to 1
- τ = 0 ⇒ ubiquist gene: expressed in all tissues at the same level
 τ = 1 ⇒ tissue-specific gene: expressed in one and only one tissue
- number of tissues with expression $\max_{1 \le i \le T} (x_i)$

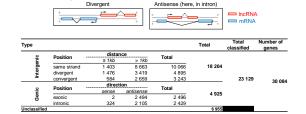
Gene expression at different τ values



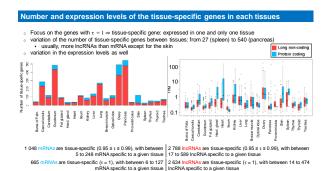


IncRNA classification using FEELnc

Long non-coding genes are classified relatively to the nearest coding genes using FEELnc (Wucher et al., 2017) ⇒ some configurations suggest potential (co)-regulations



Antisense (here, in intron)



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