

# Tissue-specific functional annotation for genetic variation

Dr. K vin Vervier, PhD

Department of Psychiatry  
University of Iowa Hospitals and Clinics

May 24, 2017



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# Context

# Functional annotation for genetic variation

- Whole-genome sequencing to understand genetic trait **architecture** in large cohorts.
- Given thousands of candidate variants, **prioritize** candidate variants.
- Especially difficult for **non-coding** variants.
- Recent scores (e.g. CADD/DANN/Eigen) predict **generic** deleteriousness annotation.
- Idea: using **tissue-related** data to derive a functional score.

We propose a new tool, **Tissue Specific Annotation** (TiSAn\*) that:

- measures how likely a position is **related to tissue functions**,
- returns high score for tissue-related positions,
- can easily be **adapted** to many tissues,
- is a predictive model, based on **machine learning**,

$$\mathbb{P}(Y = \textit{tissue} \mid X) = f(X; w).$$

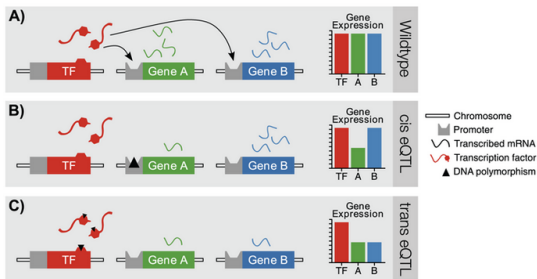
\*also the french word for herbal tea

## **Collect information at the tissue level**

# Gene-Tissue Expression (GTEx)



- study gene expression in  $\sim 50$  different tissues
- genotype available for most of the donors
- tissue-specific expression Quantitative Trait Loci (eQTL)

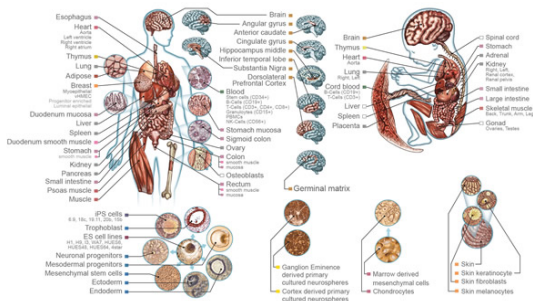


Source: Wolen and Miles, 2012

# ENCODE/RoadMap Epigenomics (RME)



- map DNA **methylation** in more than 80 cell types
- tissue-specific regulation mechanisms
- differentially methylated regions



Source: RME Consortium, 2015



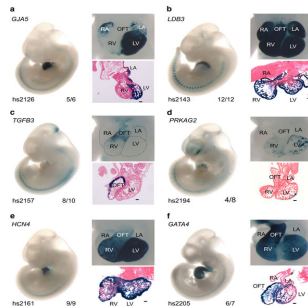


- gene2pubmed: curated database for articles citing genes
- **literature mining** for tissue-related genes (NCBI API)
- query on tissue-gene co-citations (title+abstract)

- ☐ [Superfluous role of mammalian septins 3 and 5 in neuronal development and synaptic transmission.](#)
4. Tsang CW, Fedchyshyn M, Harrison J, Xie H, Xue J, Robinson PJ, Wang LY, Trimble WS.  
Mol Cell Biol. 2008 Dec;28(23):7012-29. doi: 10.1128/MCB.00035-08. Epub 2008 Sep 22.  
PMID: 18809578 **Free PMC Article**  
[Similar articles](#)
- ☐ [Targeted disruption of Sept3, a heteromeric assembly partner of Sept5 and Sept7 in axons, has no effect on developing CNS neurons.](#)
5. Fujishima K, Kiyonari H, Kurisu J, Hirano T, Kengaku M.  
J Neurochem. 2007 Jul;102(1):77-92.  
PMID: 17564677 **Free Article**  
[Similar articles](#)
- ☐ [Septin 3 \(G-septin\) is a developmentally regulated phosphoprotein enriched in presynaptic nerve terminals.](#)
6. Xue J, Tsang CW, Gai WP, Malladi CS, Trimble WS, Rostas JA, Robinson PJ.  
J Neurochem. 2004 Nov;91(3):579-90.  
PMID: 15485489 **Free Article**

# Projects dedicated to one tissue

- previous databases work for a large set of tissues
- we also integrate data from **single-tissue** projects
- developmental brain methylation (Spiers et al., 2015)
- fetal heart enhancers (Dickel et al., 2016)
- especially relevant for **rare** tissues



Source: Dickel, 2016

# Machine-learning for functional annotation

# Features space description

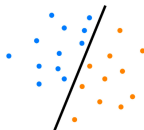
For each genomic position, we extract the following tissue-specific descriptors

- **transcriptomics**:
  - distance to the closest tissue eQTL (GTEx)
  - distance to the closest 'tissue gene' (PubMed)
- **epigenomics**:
  - distance to methylation regions (RME)
  - methylation level (RME)
- **genomics**:
  - $n$ -nucleotides composition in 1kb neighborhood ( $n = 1, 2, 3, 4$ )
- **single-tissue data**:
  - fetal brain methylation (Spiers *et al.*, 2015)
  - heart enhancers (Dickel *et al.*, 2016)

Currently, ~360 features are used in the data representation.

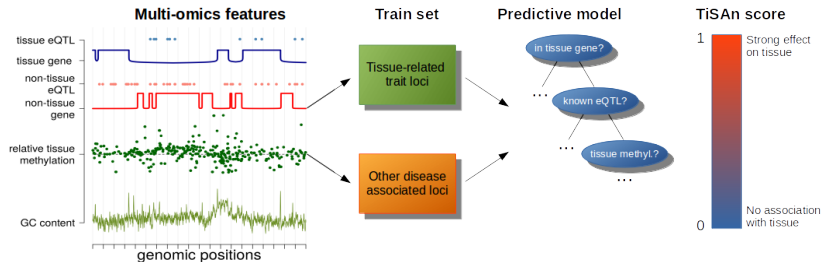
# Training set definition

- Supervised machine learning consists in **separating** positive and negative examples
- Decision rule optimization and **pattern** detection



- Association between a location and a tissue can be found in **disease-related** loci.
- Online databases like **GWAS Catalog**, genotype array probes (e.g., PsychArray and MetaboChip).

# Method overview



TiSA framework

- Applied on two human tissues: [brain](#) and [heart](#).

# Tutorial

# Tutorial: database + vignettes

- **R** vignettes are distributed as a package and provide [guidelines](#) for each model development step
- Genome-wide scores for heart and brain are [available](#) at <http://flamingo.psychiatry.uiowa.edu/TiSAn>
- .bed format makes TiSAn easy to [integrate](#) in most bioinformatics pipelines

[Github](#): <http://github.com/kevinVervier/TiSAn>



# Tutorial: get TiSAn scores for candidate loci

- Input: VCF file with loci of interest.
- Plug TiSAn databases in, for instance, *vcfanno* tool.

```
kvervier@luxor:~/git_repos/TiSAn$ vcfanno ~/varann/aim1/data/TiSAn.conf data/example1.vcf
=====
vcfanno version 0.2.4 [built with go1.8]

see: https://github.com/brentp/vcfanno
=====
vcfanno.go:115: found 2 sources from 2 files
##fileformat=VCFv4.2
##INFO=<ID=TiSB,Number=1,Type=Float,Description="calculated by max of overlapping values in column 4 from /sdata/vcfannotations/TiSAn_Brain.bed.gz">
##INFO=<ID=TiSH,Number=1,Type=Float,Description="calculated by max of overlapping values in column 4 from /sdata/vcfannotations/TiSAn_Heart.bed.gz">
#CHROM POS ID REF ALT QUAL FILTER INFO FORMAT
1 1005806 rs3934834 C T 0.0 PASS TiSB=0;TiSH=0.4307
1 243943084 rs4132509 C A 0.0 PASS TiSB=0;TiSH=0.5663
13 81753314 rs12584499 C G 0.0 PASS TiSB=0.718;TiSH=0
14 62763347 rs2354331 C T 0.0 PASS TiSB=0.358;TiSH=0.2904
21 37417489 rs2835248 A G 0.0 PASS TiSB=0;TiSH=0
4 154746806 rs10031057 A G 0.0 PASS TiSB=0;TiSH=0.8565
vcfanno.go:241: annotated 6 variants in 0.13 seconds (47.8 / second)
```

# Tutorial: visualization in UCSC Genome Browser

- Step 1: Access the UCSC Genome Browser custom track page
- For both TiSAn scores, paste instructions in "Paste URLs or data" box
- Submit your query.

Genomes Genome Browser Tools Mirrors Downloads My Data Help About Us

**Add Custom Tracks**

clade  genome  assembly

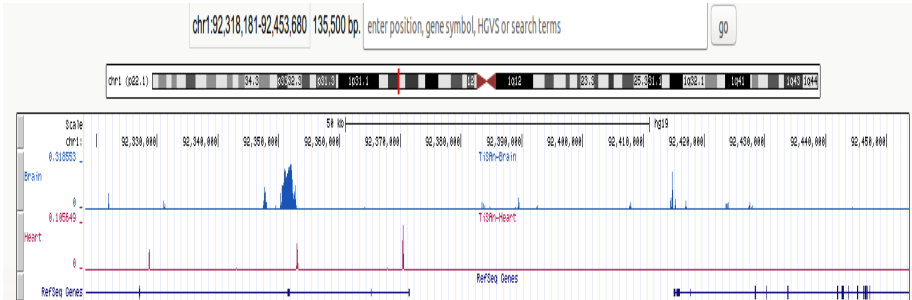
Display your own data as custom annotation tracks in the browser. Data must be formatted in [bigBed](#), [bigChain](#), [bigWig](#), [Genome SNP](#), [PSL](#), or [WIG](#) formats. To configure the display, set [track](#) and [browser](#) line attributes as described in the box below. Examples are [here](#).

Paste URLs or data: Or upload:  No file selected.

```
track type=bigwig name="Brain" description="TiSAn-Brain" visibility=full
autoScale=on alwaysZero=on maxHeightPixels=100:30:10 color=24,84,181
bigDataUrl=http://flamingo.psychiatry.uiowa.edu/TiSAn/TiSAn_Brain.bw
track type=bigwig name="Heart" description="TiSAn-Heart" visibility=full
autoScale=on alwaysZero=on maxHeightPixels=100:30:10 color=181,24,84
bigDataUrl=http://flamingo.psychiatry.uiowa.edu/TiSAn/TiSAn_Heart.bw
```

Optional track documentation: Or upload:  No file selected.

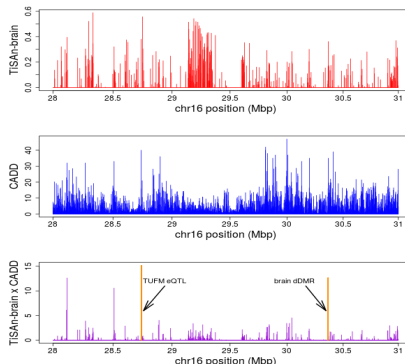
## Tutorial: visualization in UCSC Genome Browser



# Applications

# Application: region-based analysis: 16p11

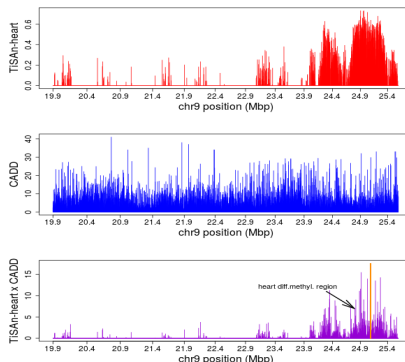
- Known to be related to autism and schizophrenia.
- On Chromosome 16, from  $\sim 28.3\text{Mb}$  to  $30.3\text{Mb}$ .
- TiSAn **combined** with pathogenicity scores, like CADD.



TiSAn-brain (top), CADD (mid), TiSAn  $\times$  CADD (bot)

# Application: region-based analysis: 9p21

- Known to be related to cardiovascular disease.
- On Chromosome 9, from  $\sim 19.9\text{Mb}$  to  $25.5\text{Mb}$ .
- TiSAn **combined** with pathogenicity scores, like CADD.



TiSAn-heart (top), CADD (mid), TiSAn  $\times$  CADD (bot)

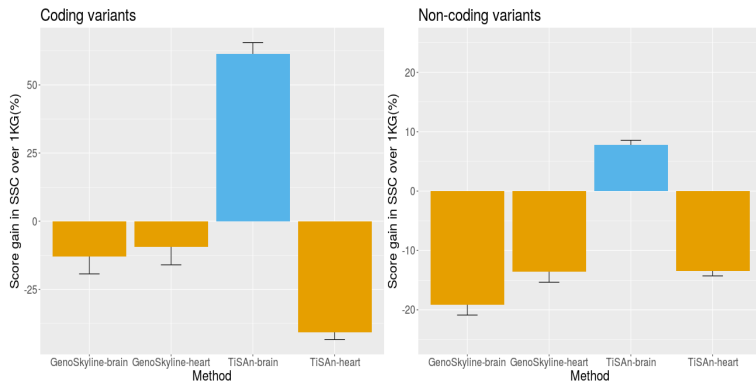
# Autism genes set enrichment



- Simons Simplex Collection (SSC): genetic repository of 2,600 simplex families with **autistic proband**
- 1,000 Genomes (1KG): genetic variation in **unaffected** population
- Expected enrichment in **brain-related** genetic burden in SSC cohort, even in unaffected family members.

# Autism genes set enrichment

- Pathogen variants found around strong **autism candidate genes**
- Comparison in average score between SSC and 1KG variants
- Comparison with GenoSkyLine (Lu *et al.*, 2016)

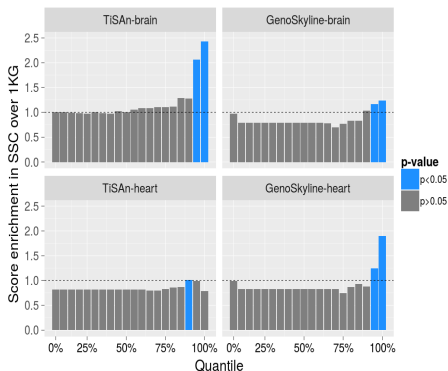


Relative score gain between 1KG and SSC.



# Autism genes set enrichment

- Mix SSC and 1KG variants and rank them based on their scores.
- For each quantile, compute **relative enrichment** in SSC over 1KG.



Quantile-wise enrichment in SSC variants over 1KG.

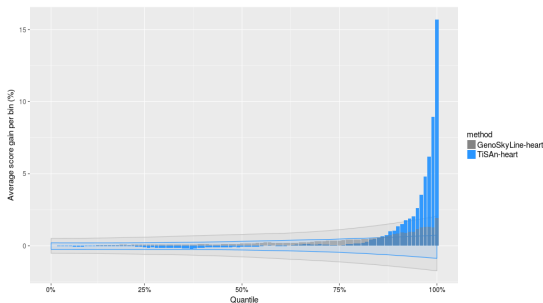
# Genome-wide association for coronary artery disease



- Combine multiple large scale genetic studies to identify risk loci for **coronary artery disease**
- Estimate trait association ( $p$ -value) for  $\sim 8,000,000$  SNPs.
- Hypothesis: TiSAn-heart score increases with association strength.

# Genome-wide association for coronary artery disease

- Loci binned based on their  $p$ -values into percentile groups
- Score gain between top percentile and remaining groups



Cumulative quantile-wise functional score enrichment.

# Conclusion

- **General** framework for tissue-related functional score
- Enrichment found for both brain and heart models in known loci
- Next steps:
  - Evaluate **deep-learning** based solutions
  - Discovery analysis in unpublished data (bipolar disorder, sudden death)
  - Combine with **SLINGER** for tissue-specific gene expression inference

# Thank you for your attention

Fundings: NIH MH105527 and DC014489

Collaborators: Dr Jacob Michaelson lab (UI)



Brew your own TiSAn!

Github: <http://github.com/kevinVervier/TiSAn>