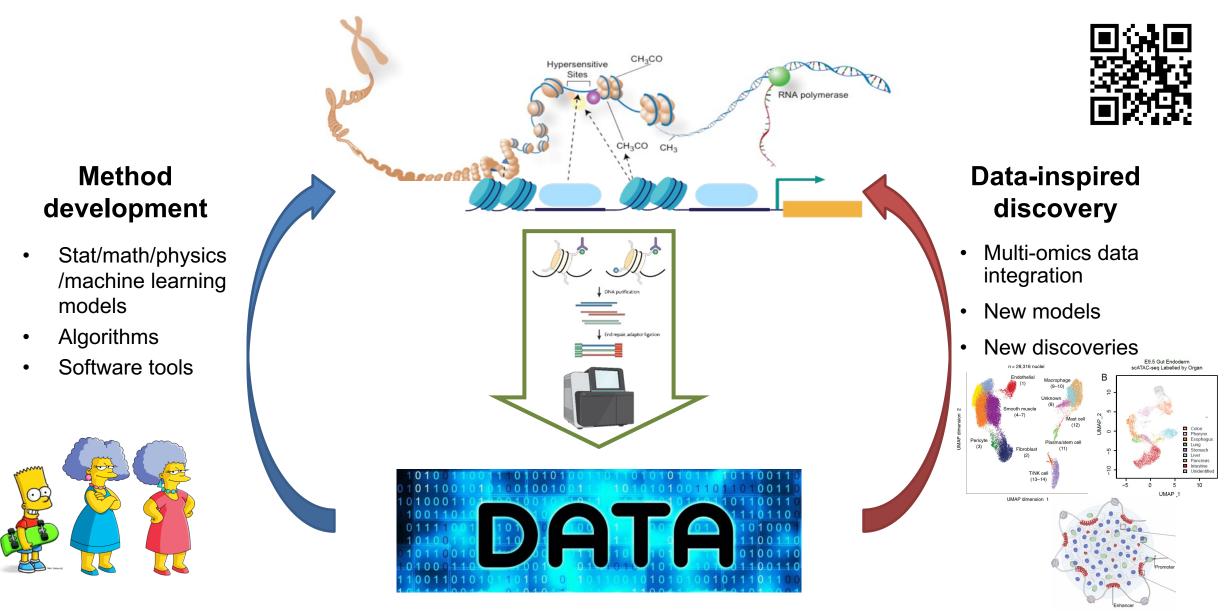
# Visualization and interpretation of high-throughput genomics data

Chongzhi Zang Associate Professor of Genome Sciences Director of Computational Genomics, UVA Cancer Center zang@virginia.edu zanglab.org

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My lab develops computational methods and uses computational approaches to study epigenetics and transcriptional regulation





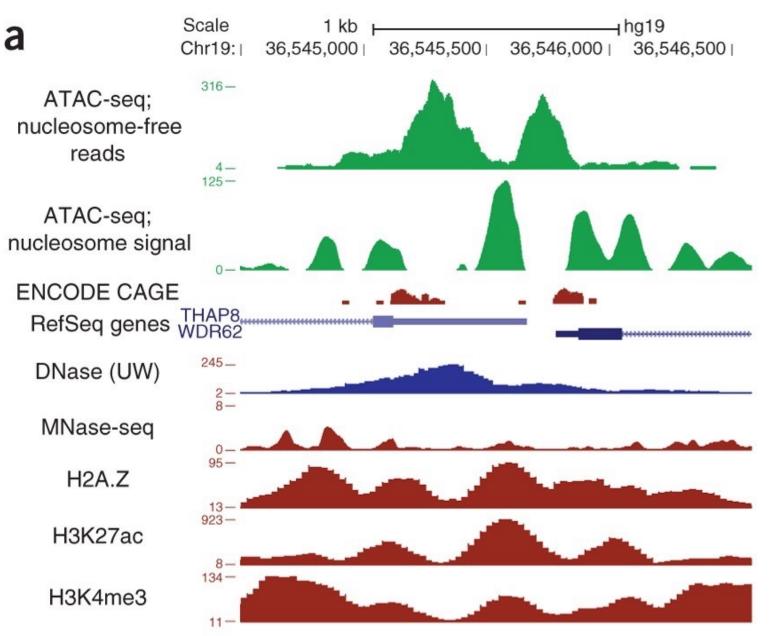
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# Learning Objectives

- Understand types of plots commonly used for presenting highthroughput genomics data
- Understand essential elements in genomics data visualization
- Get some tips for data presentation

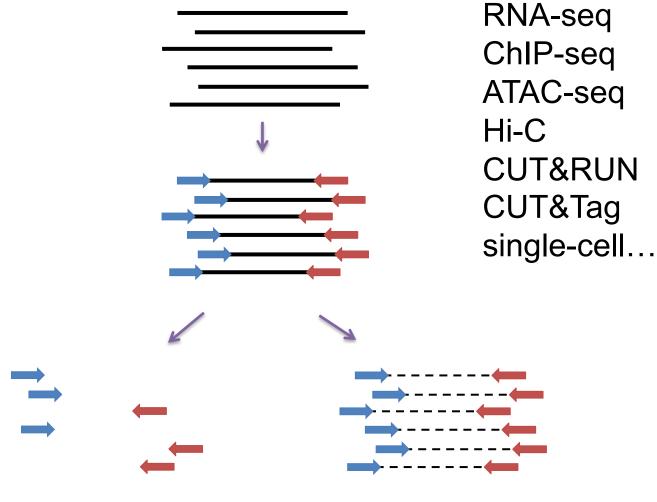
Chr19 (q13.12) С 19p13.3 19p13.2 13.12 19p13.11 19p12 19q12 q13.11 q 3.12 19q13.2 q13.32 q13.33 13.41 q13.42 q13.43 Scale 50 kb F -| hg19 Chr19: 36,200,0001 36,150,0001 36,250,0001 1-ATAC-seq (50,000 cells per replicate) 0 DNase HS ENCODE/Duke 0 1 -FAIRE-seq ENCODE/UNC 0 ATAC-seq (500 cells) a di 11 0 - 🚻 CTCF H3K4me1 1 H3K27ac H3K4me3 HAUS5 ZBTB32 UPK1A UPK1A-AS1 ■++++ ETV2 COX6B1++++++

Buenrostro et al. Nat Methods 2013



Buenrostro et al. Nat Methods 2013

# High-throughput short-read sequencing (Illumina)

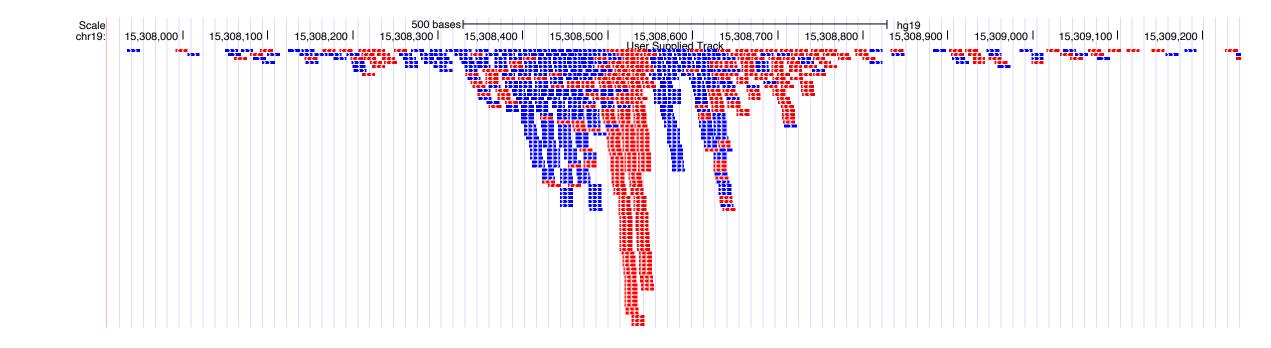




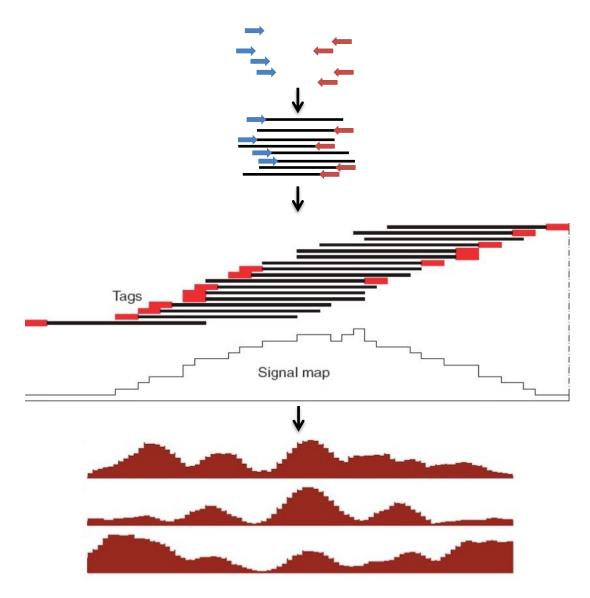
Single-end



#### Original sequence reads are not easy to visualize



#### Signal tracks are sequence fragments piled up



#### • bedGraph:

chr4	10344200	10344250	5
chr4	10344250	10344300	10
chr4	10344300	10344350	25
chr4	10344350	10344400	15
chr4	10344400	10344450	8

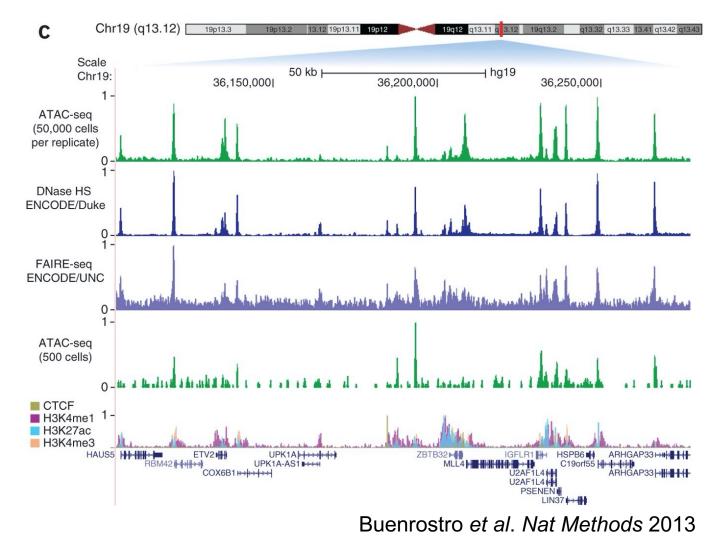
#### • wiggle:

track type=wiggle\_0 variableStep chrom=chr4 span=50 10344200 5 10344250 10 10344300 25 10344350 15 10344400 8

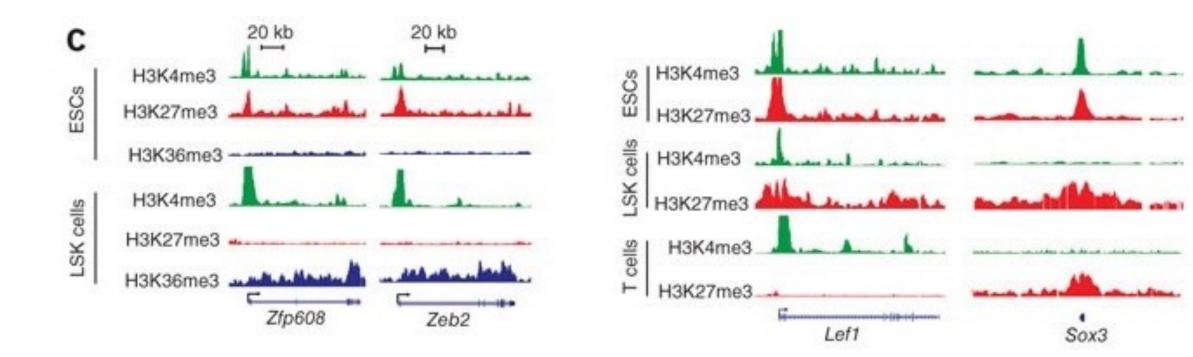
• bigWig: indexed binary format

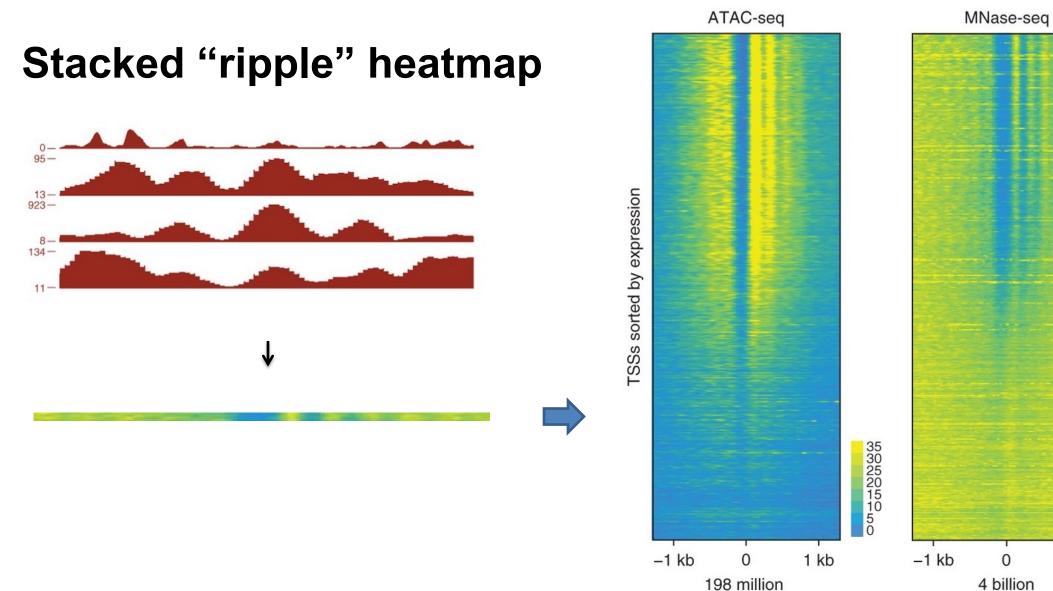
# **Essential elements in genome browser tracks**

- Chromosomal locations
- Track label
- Track scale
  x: resolution?
  y: normalization?



## How to integrate patterns observed on signal tracks?





4 billion single-end reads

paired reads

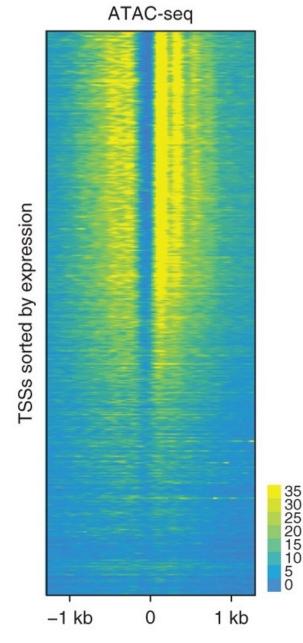
Buenrostro et al. Nat Methods 2013

1.20 1.04 0.88 0.72 0.56 0.40 0.24 0.08

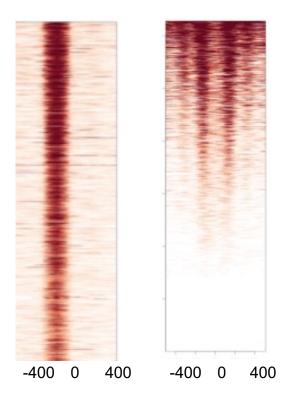
1 kb

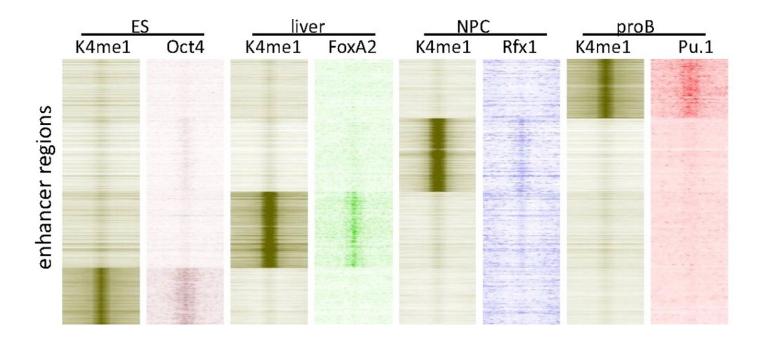
# **Essential elements in a ripple heatmap**

- Heatmap presents 3-dimensional data
- x: What loci/anchor is each row? Range?
- y: What are the rows? How many? How are they ranked?
- h: Data title/label (what signal?) color scale?



#### Multiple datasets visualization by ripple heatmap

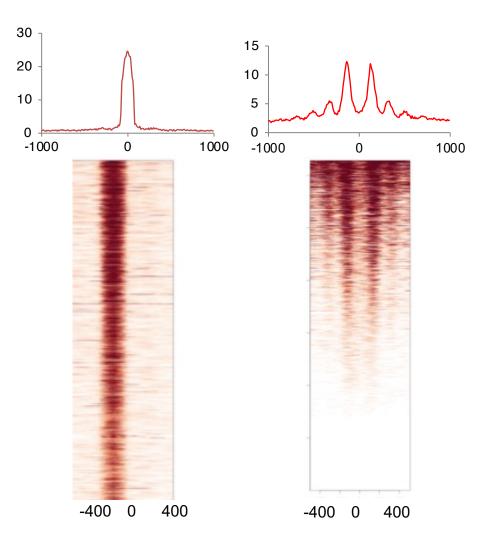




Creyghton et al. PNAS 2010

Luyten et al. Genes Dev 2014

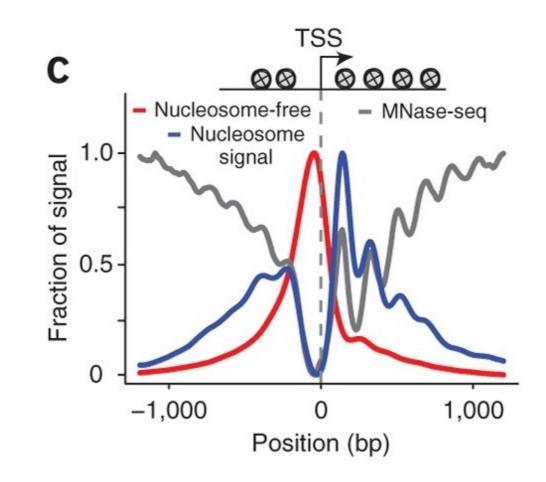
# **Composite plots**



Luyten et al. Genes Dev 2014

## **Essential elements in a composite plot**

- Data title/label/legend
- Data source (average of what?)
- x: anchor, scale
- y: scale, normalization

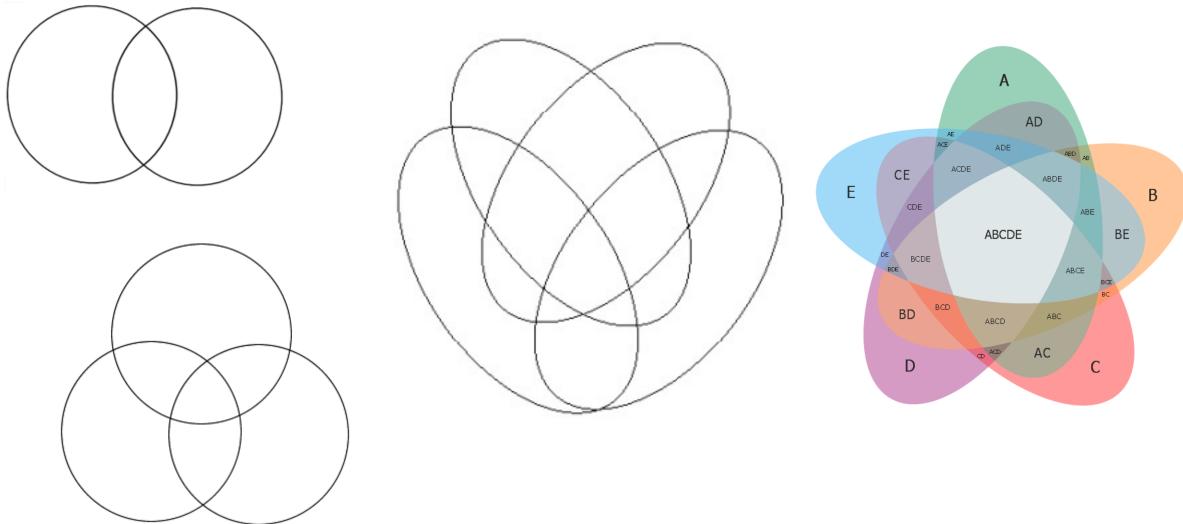


Buenrostro et al. Nat Methods 2013

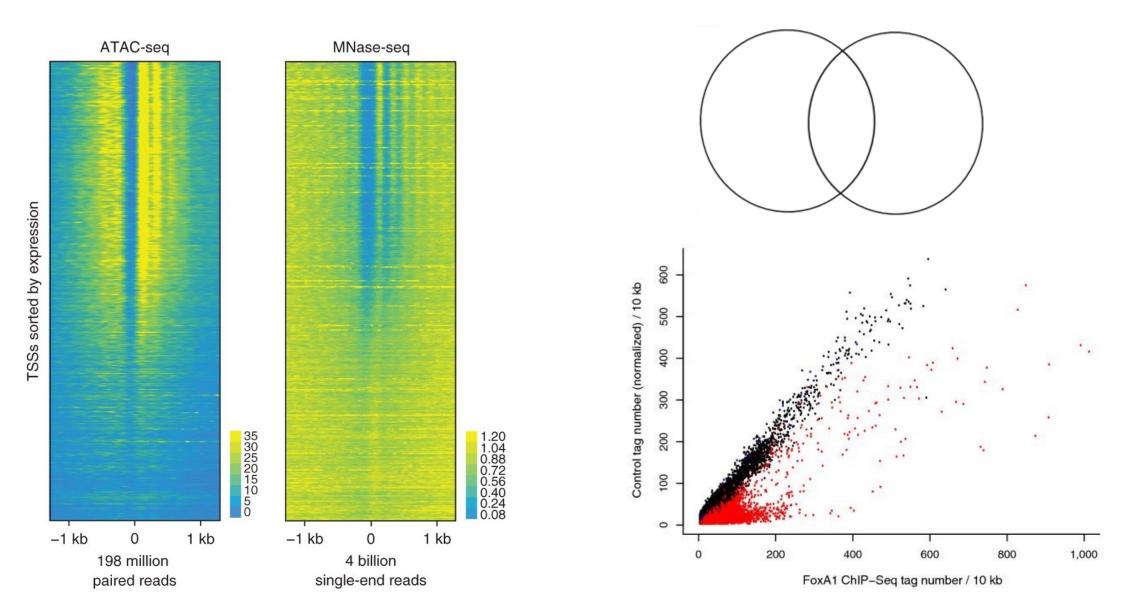
# **Common misinterpretations of composite plots**

- Caveat 1: A peak in a composite plot may be contribute by only a tiny fraction of regions (not representative of global picture)
- Caveat 2: A higher peak does not necessarily mean stronger signal or more region coverage

#### Venn diagram presents yes/no relations between sets



#### Heatmaps and scatter plots are more informative than a Venn diagram

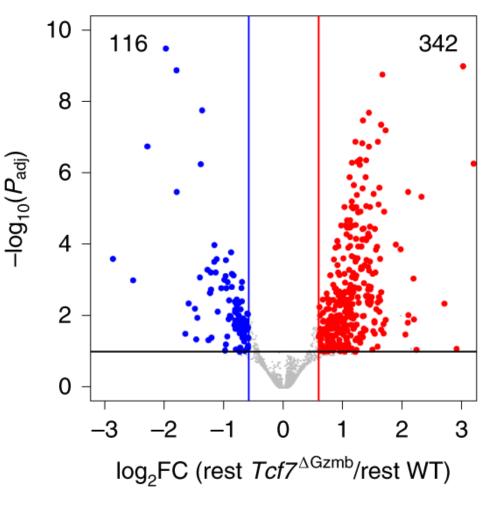


# Volcano plot for differential gene expression

- Scatter plot
- 2 dimensions:

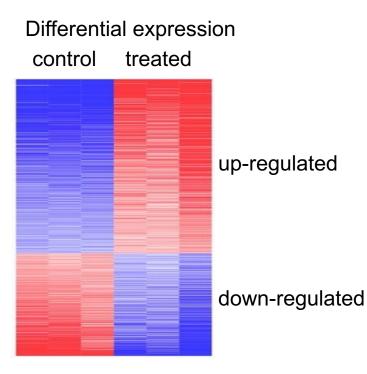
x: signal strength (e.g., log2 fold change)y: statistical significance (e.g., -log10P)

Set cutoffs on 2 axes

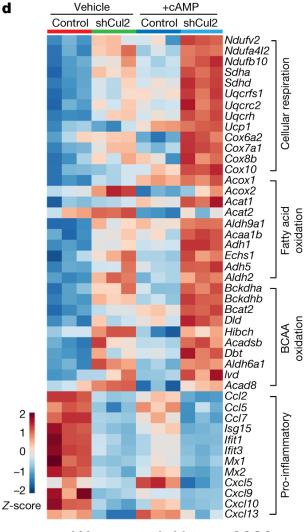


Shan et al. Nature Immunology 2022

# Differential gene expression visualization by heatmap

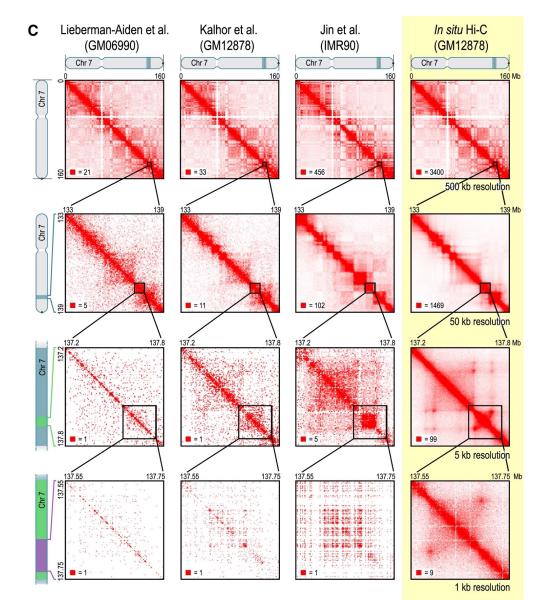


# Co-expressed gene clusters



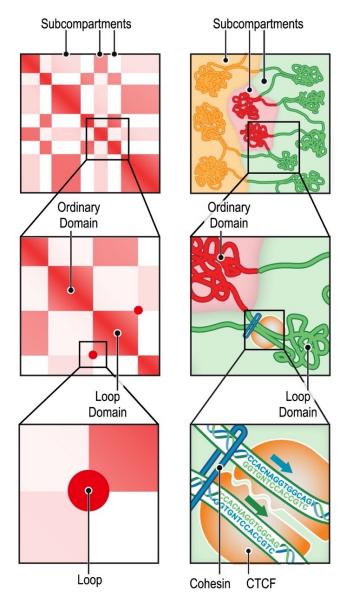
Wang et al. Nature 2022

# **Hi-C contact heatmap for 3D genome interactions**



Rao et al. Cell 2014

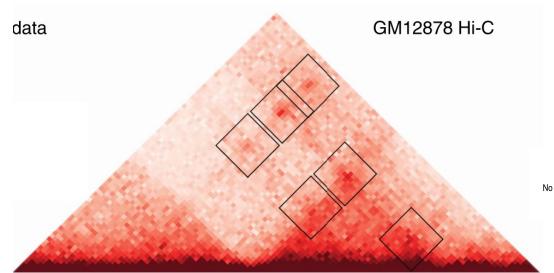
# Hi-C contact heatmap for 3D genome interactions

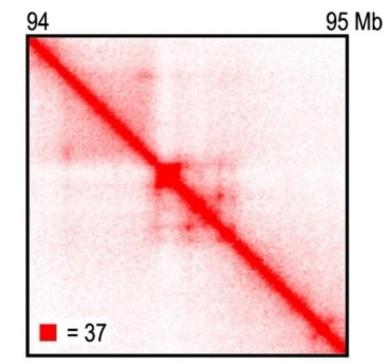


Rao et al. Cell 2014

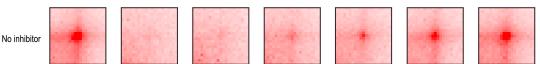
# **Essential elements in a Hi-C contact heatmap**

- Scale, scale, scale
- Resolution
- Normalization
- Blocks, stripes, loops (2d peaks)

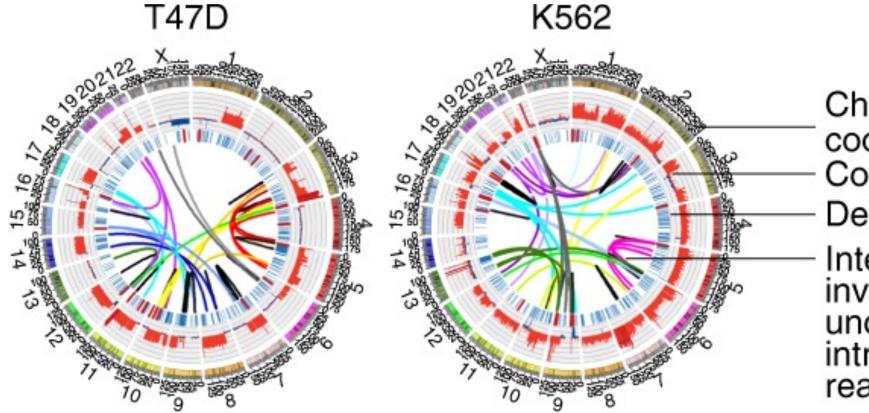




5 kb resolution

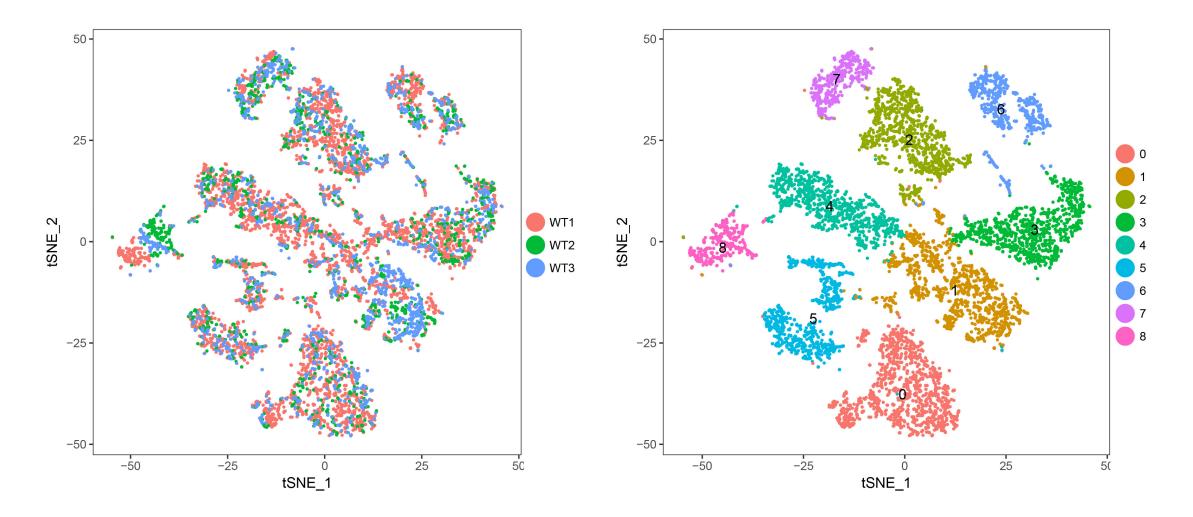


# Circos plot integrates multiple types of genomics data

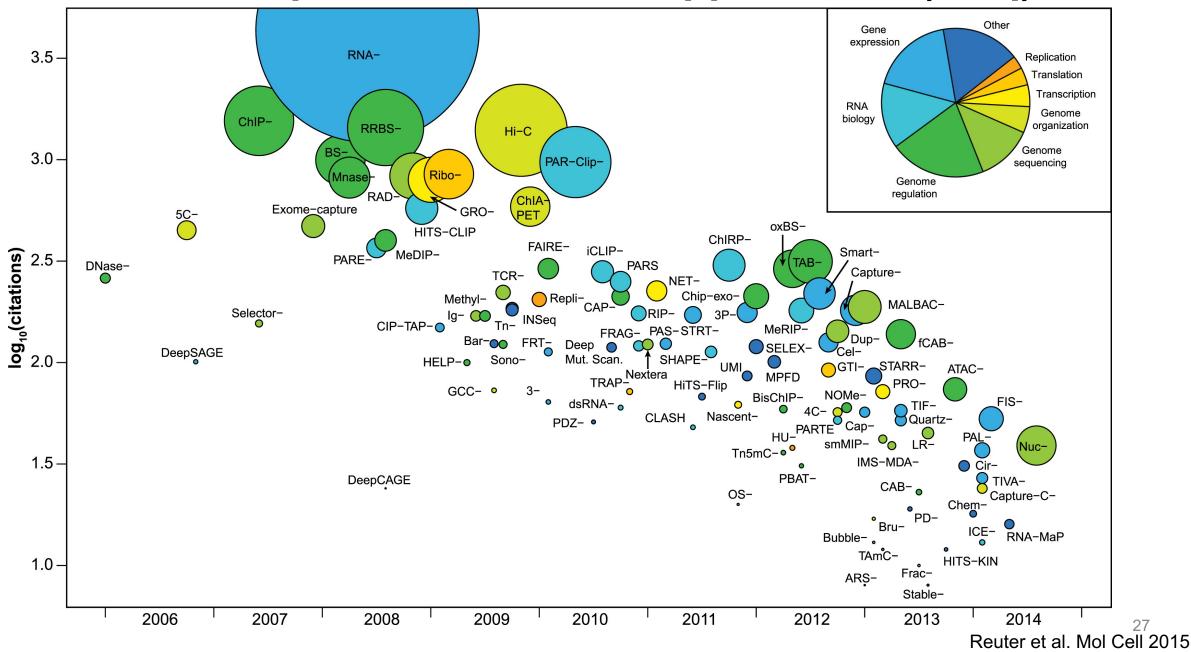


 Chromosome
 coordinates
 Copy number
 Deletion, duplication
 Interchromosomal TLs inversions, and unclassified intrachromosomal rearrangements (>1Mb)

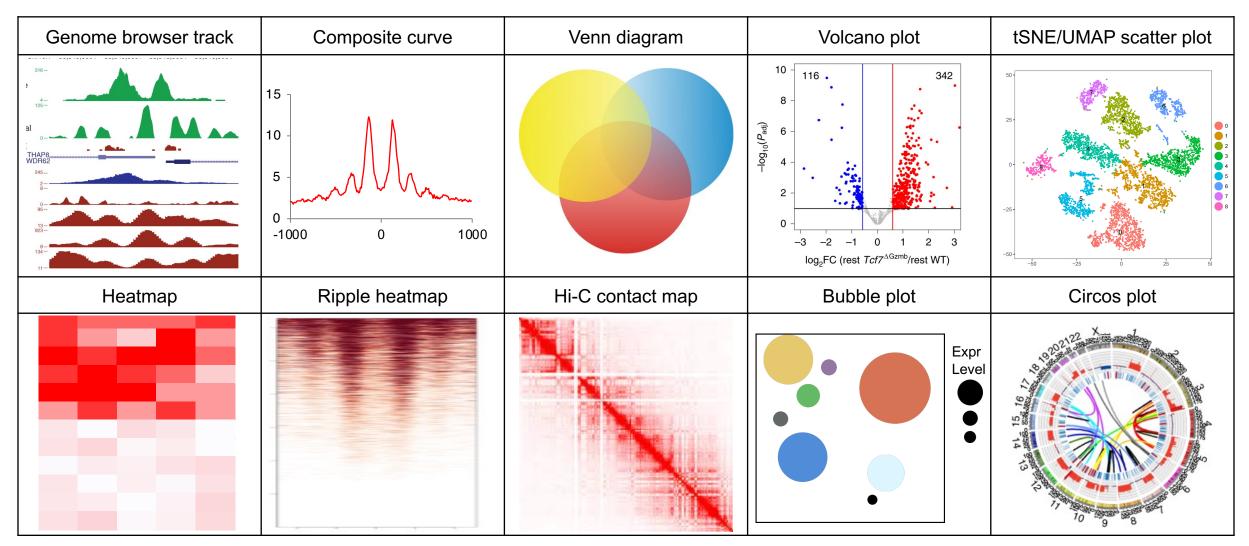
#### Single-cell data: clustering vs. t-SNE/UMAP visualization



#### **Bubble plots: NGS-based applications (-seq)**



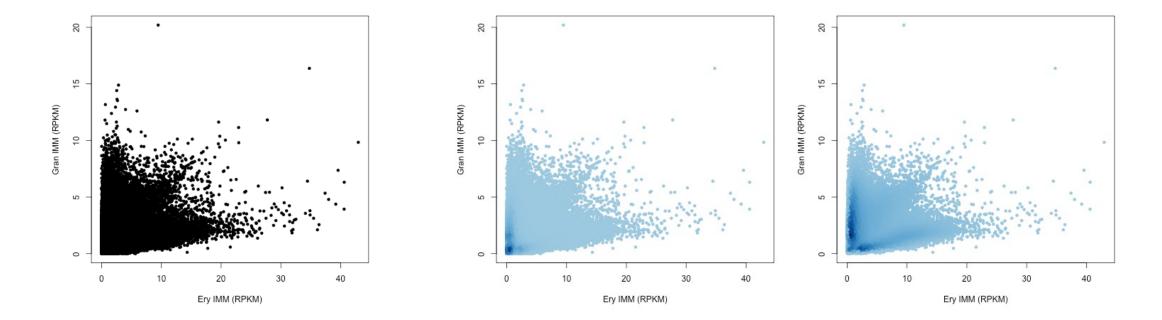
# Summary



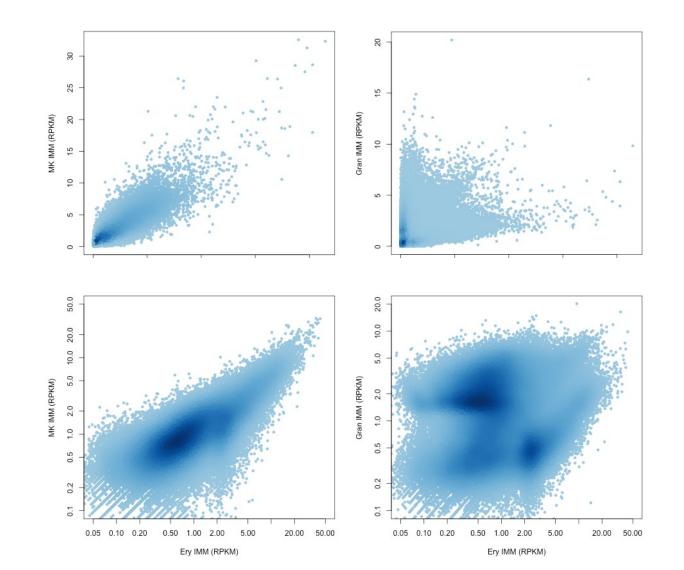
# Some tips

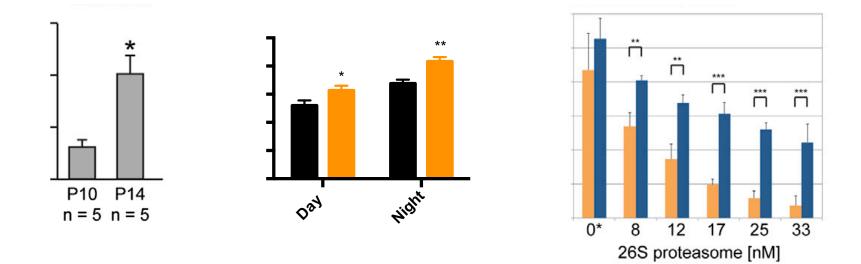
# **Scatter plot**

• The density of the data points matters!



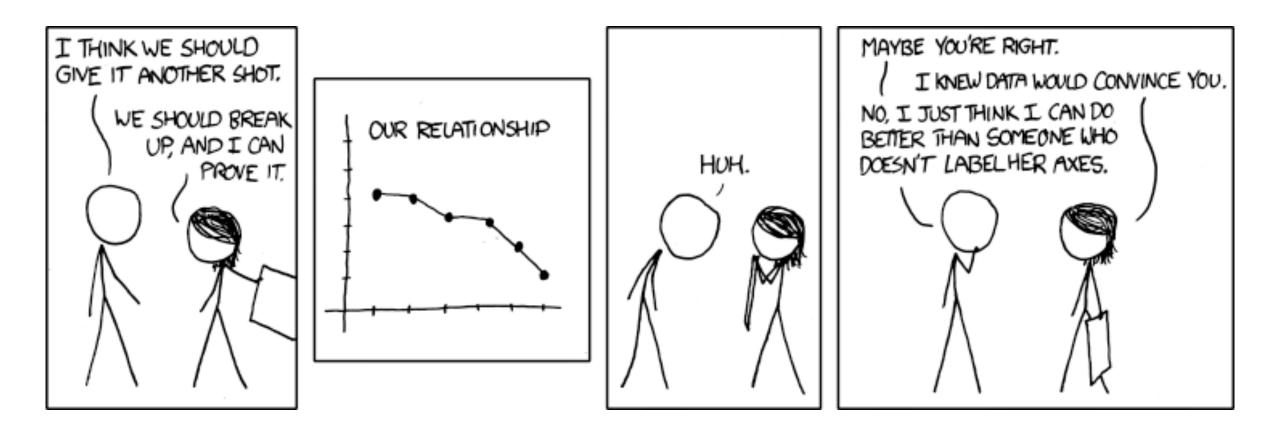
# Choose the appropriate scale for plotting (linear or logarithm)

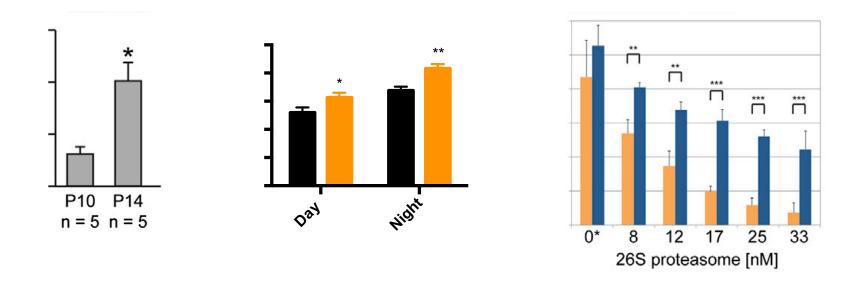




?

Plots from Cell 2014

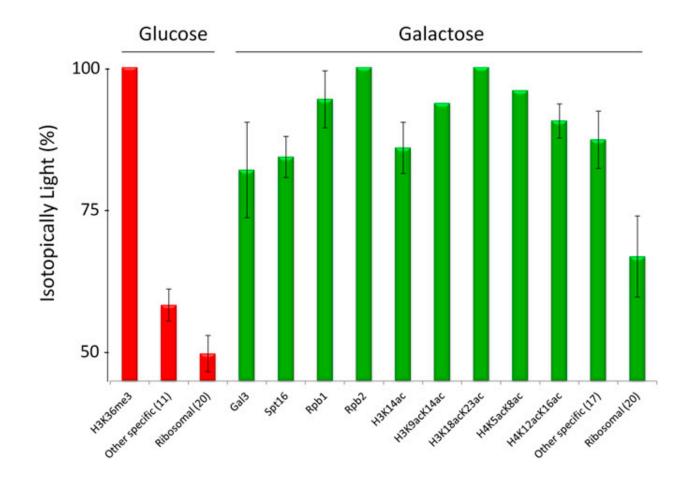




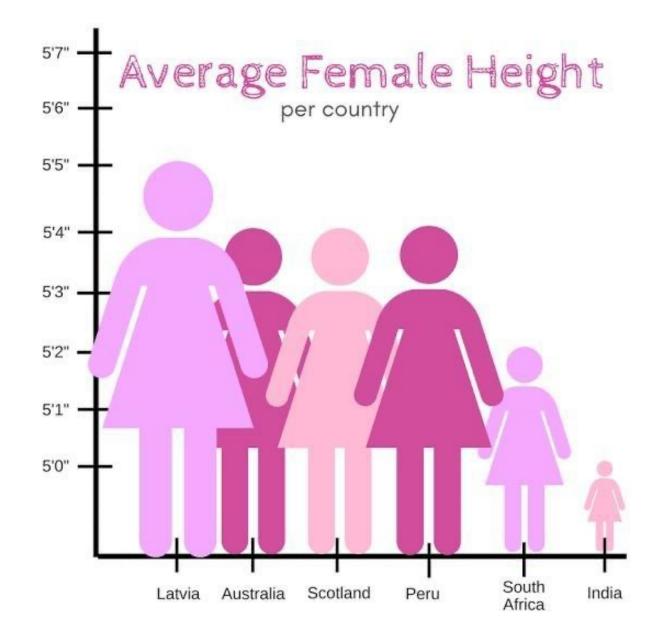
?

Bar charts should always start from 0, and on the linear scale. If the difference is small, box plots or original data points are better.

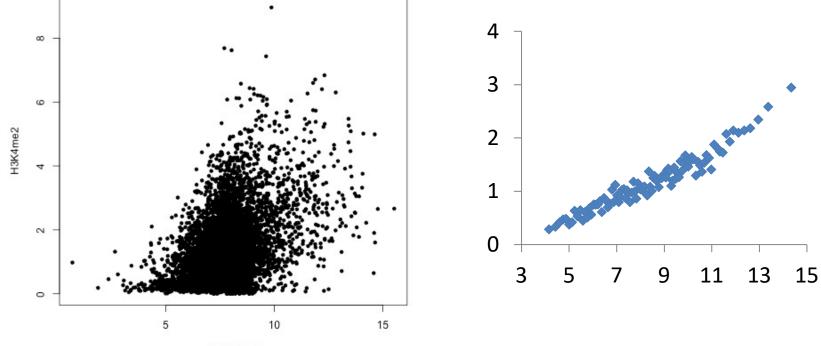
# **Another example**



Byrum et al. Cell Reports 2012

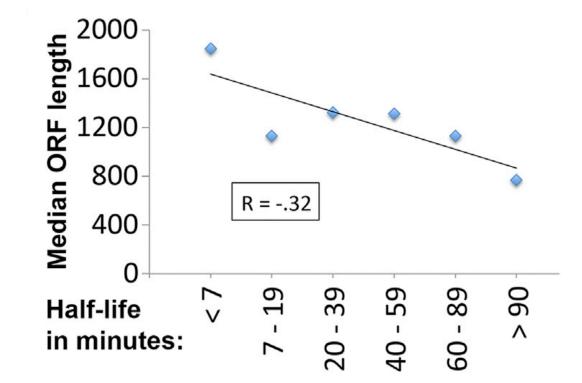


# Scatter plot? Group them if needed

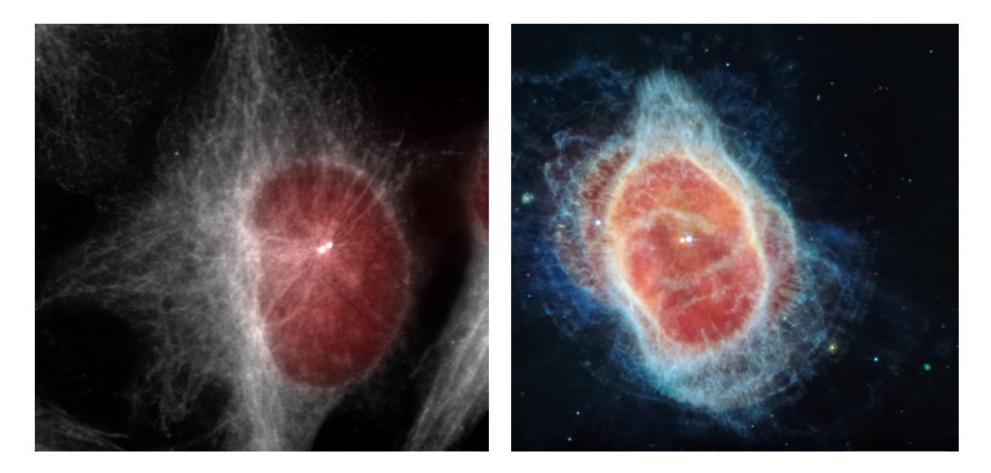


expression

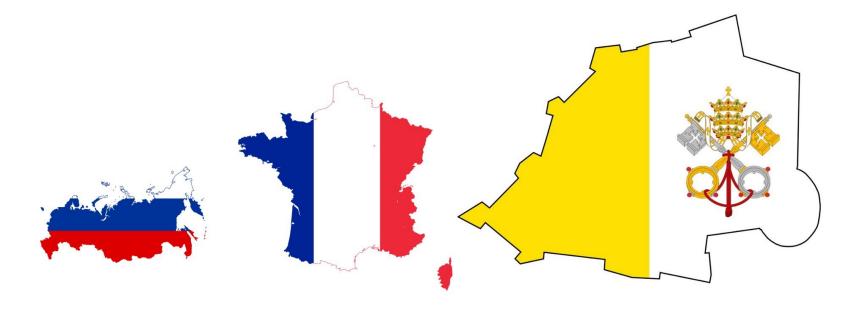
# However, grouping should be unbiased



# **Scale matters!**



# Areas of Russia, France and Vatican compared\*



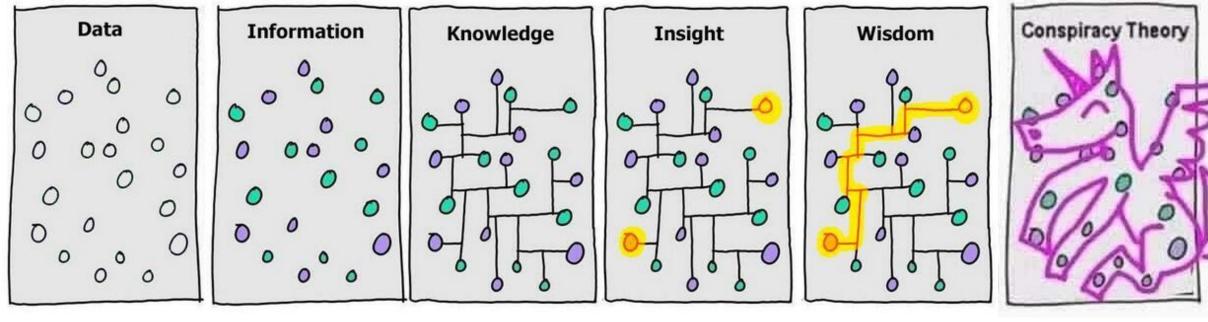
\*not to scale

Scale matters. It does!

# **Take-Home Tips**

- Always read the axes and pay attention to the scales on a plot.
- Bar charts should always start from zero on the linear scale.
- Data point density on a scatter plot is important.
- Group the data points if needed, but do it in an unbiased way.

# Having the data is not enough; presentation and interpretation matter



RNA-seq ChIP-seq DNase/ATAC-seq Hi-C Single-cell resolution...

. . .

Gene expression Protein factors Chromatin 3D genome Multi-omics

. . .

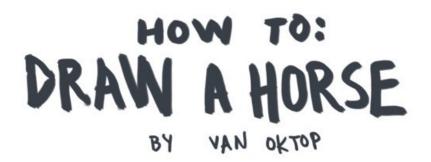
Transcriptional regulation, Chromatin organization,

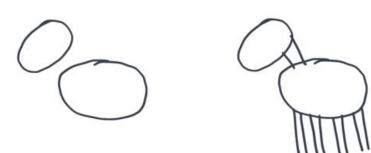
. . .

New insight?

New biology!

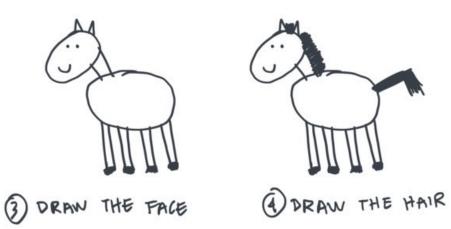
Overfitting, overinterpretation...





DRAW 2 CIRCLES

DRAW THE LEGS





# Thank you very much!

zang@virginia.edu

zanglab.org

