Epigenetics and ChIP-seq



CSAMA 2015, Brixen 16. 06. 2015. Aleksandra Pekowska aleksandra.pekowska@embl.de



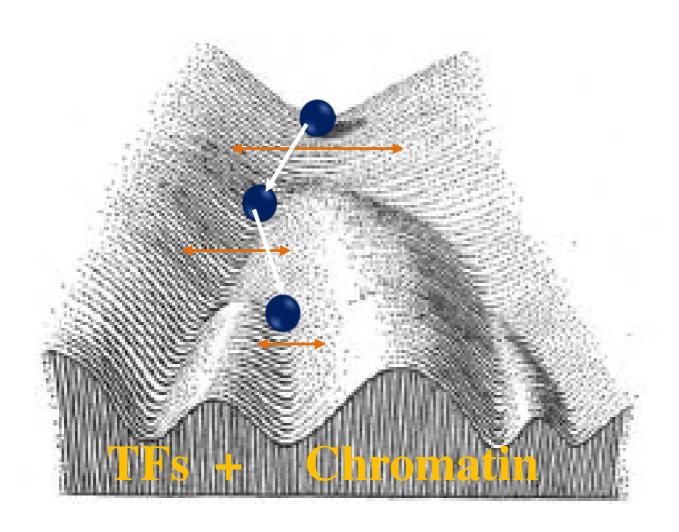
Outline of the lecture

Purpose: introduce basic steps and key considerations in ChIP-seq analysis

- 1. Epigenetics fundamental concepts
- 2. The ChIP-seq method
- 3. What kind of information can we obtain from ChIP-seq?
- 4. Study design
- 5. ChIP-seq analysis workflow:
 - a. Preprocessing
 - b. Quality controls
 - c. Isolation of enriched regions
 - d. Analysis of enriched regions
 - e. Visualization
 - f. Average profiles
 - g. Comparative analysis of enriched regions

Epigenetics - inheritance, but not as we know it

Non-genic memory of function transmitted from generation to generation (A. Bird)

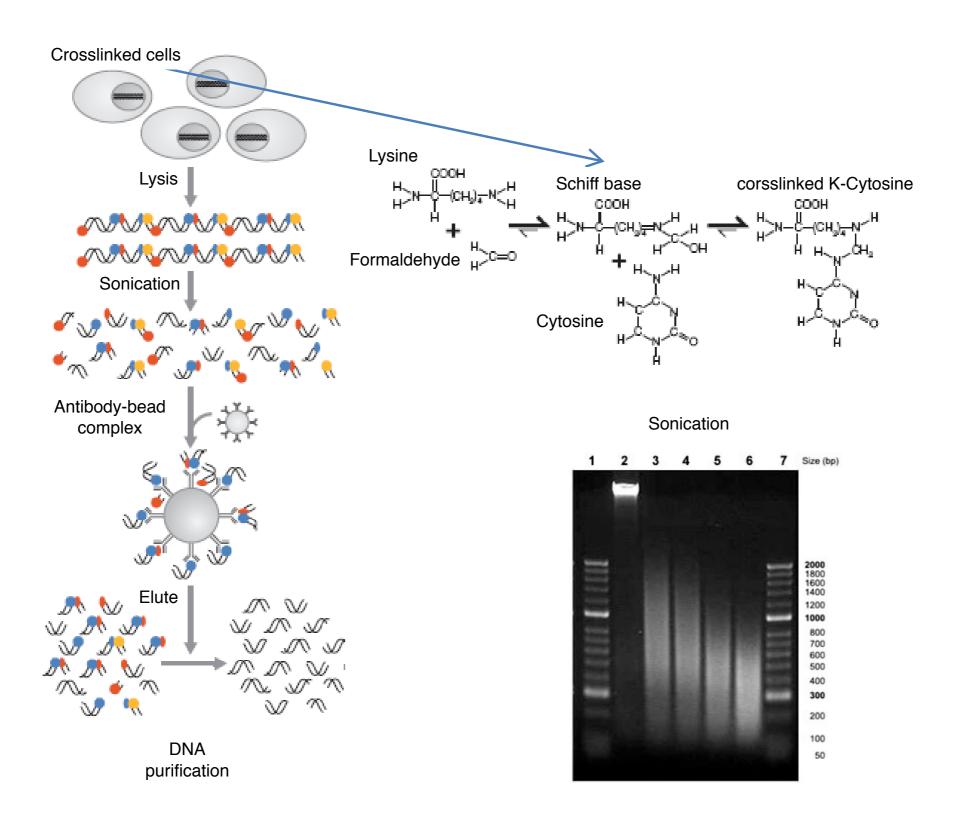


Adapted from Conrad Hal Waddington (1942)

Factors which are analysed:

- DNA methylation
- nucleosome occupancy
- histone modifications
- transcription factors
- RNA-polymerases
- chromatin modifying enzymes

Chromatin Immunoprecipitation



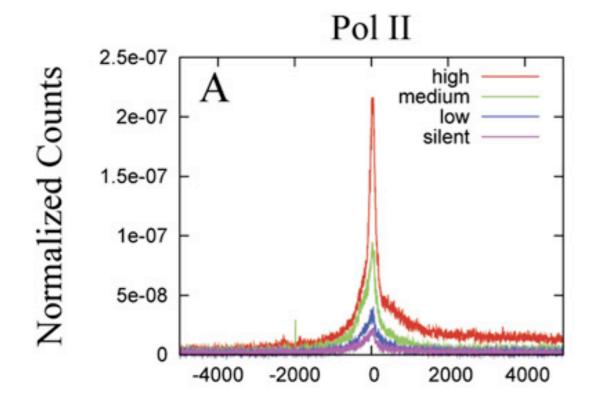
What kind of information can we obtain from the ChIP-seq experiments?

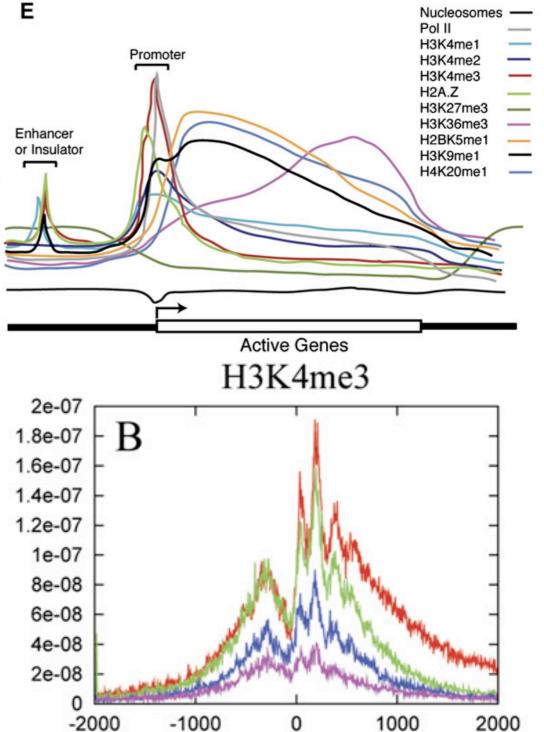
Resource

High-Resolution Profiling of Histone Methylations in the Human Genome

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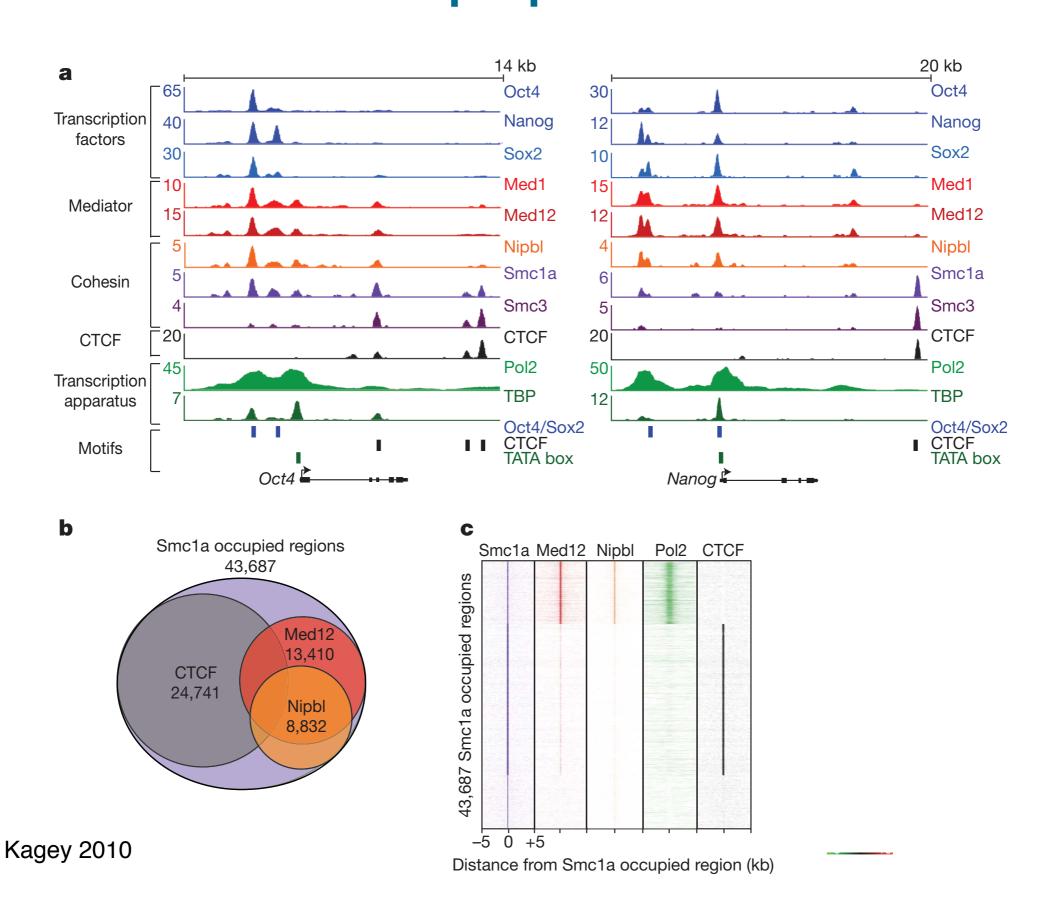
¹Laboratory of Molecular Immunology, National Heart, Lung, and Blood Institute, NIH, Bethesda, MD 20892, USA

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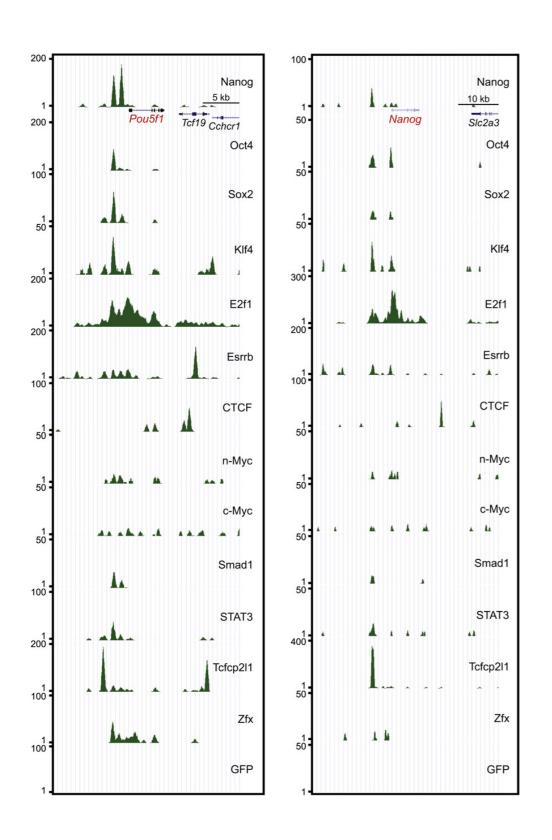
³These authors contributed equally to this work and are listed alphabetically.

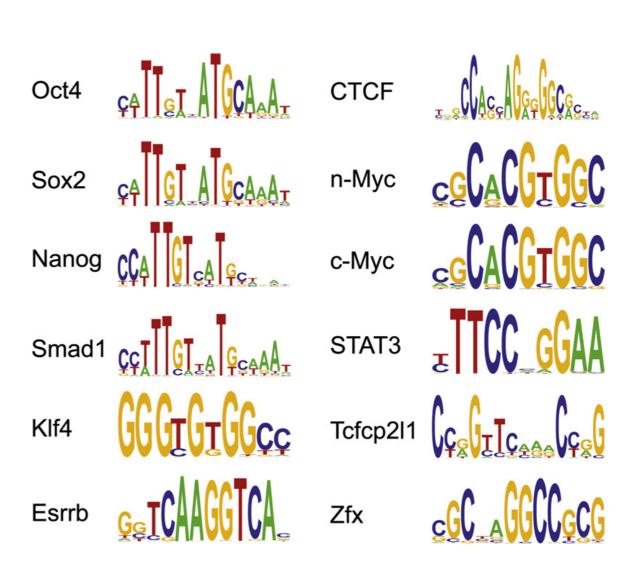
^{*}Correspondence: zhaok@nhlbi.nih.gov

What kind of information can we obtain from the ChIP-seq experiments?



What kind of information can we obtain from the ChIP-seq experiments?



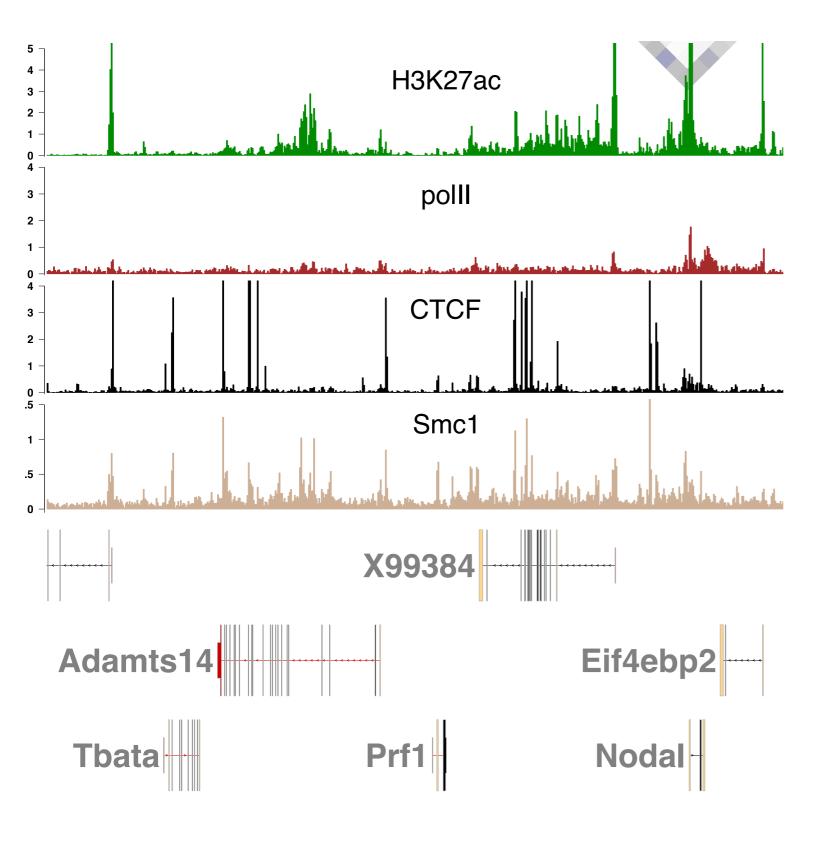


To summarize - the most frequent tasks are:

- 1. Visualization along the genome
- 2. Peak finding and analysis (localization, co-occurrences, motifs)
- 3. Heatmaps of signal and average profiles at various genomic *loci*

But before we start the analysis... ChIP-seq: considerations for study design

- Distribution of modification number of sequenced reads
- Paired vs. single end sequencing fragment length estimation
- IgG control (pros and cons)
- Input control
- Biological replication!



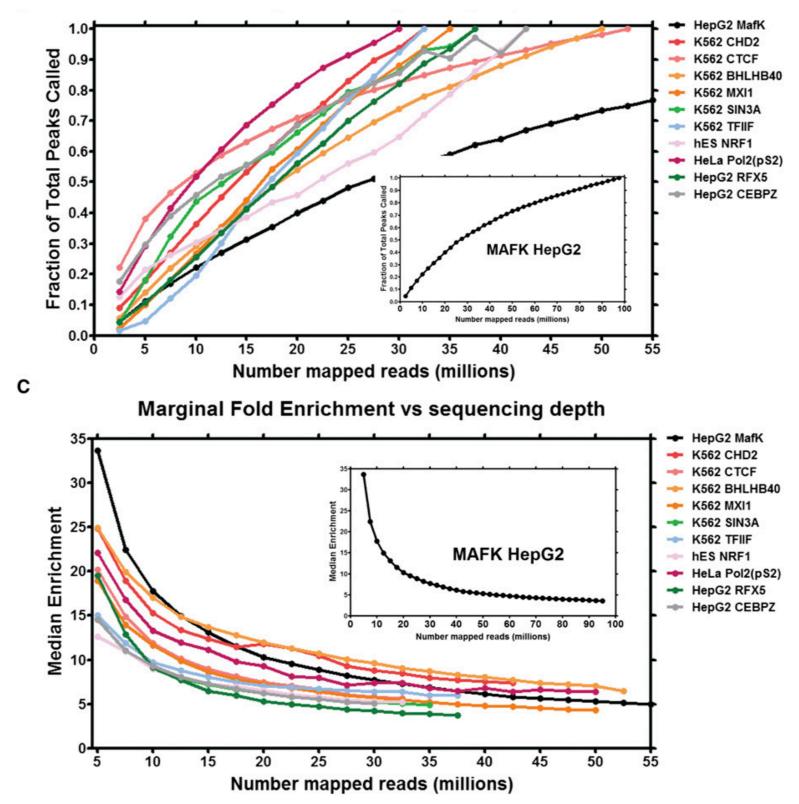
ChIP-seq profiles

- peaks vs. large domains
- signal to noise ratio



Data from: Creyghton 2010 Kagey 2010

ChIP-seq: sequencing depth matters



ENCODE consortium guidelines

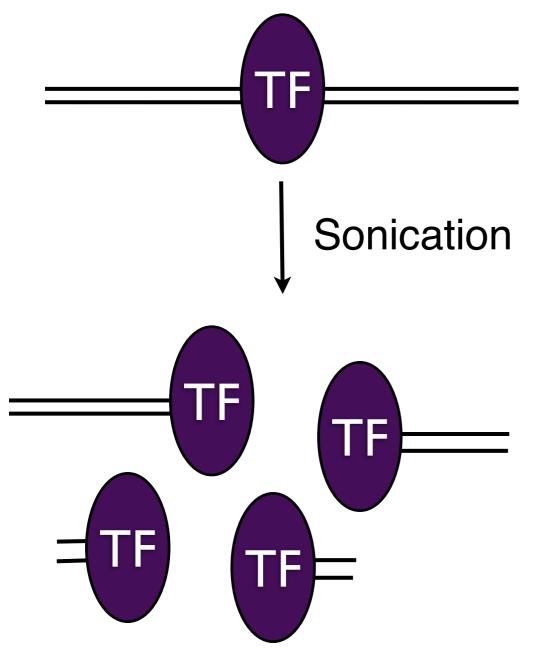
For mammalian genomes such as human and mouse:

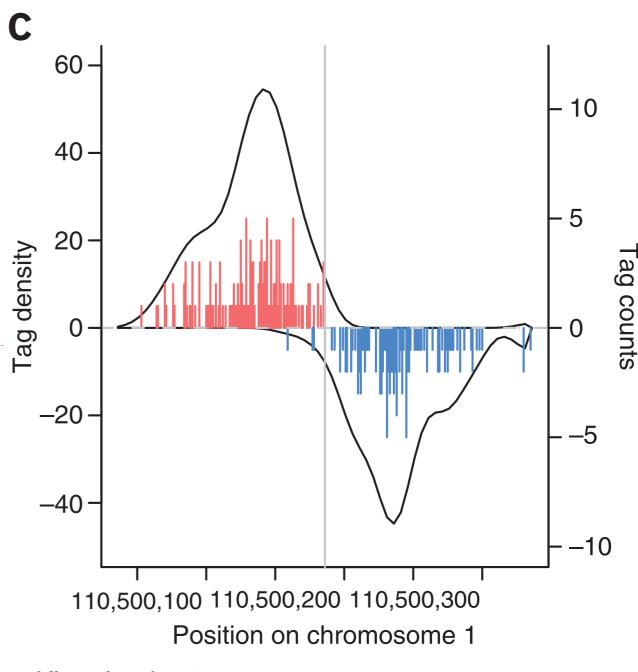
- 1. > 20M aligned reads for broad marks
- 2. > 10M aligned reads for TFs

Paired vs. single end sequencing

- paired end sequencing is always useful (nucleosome positioning)

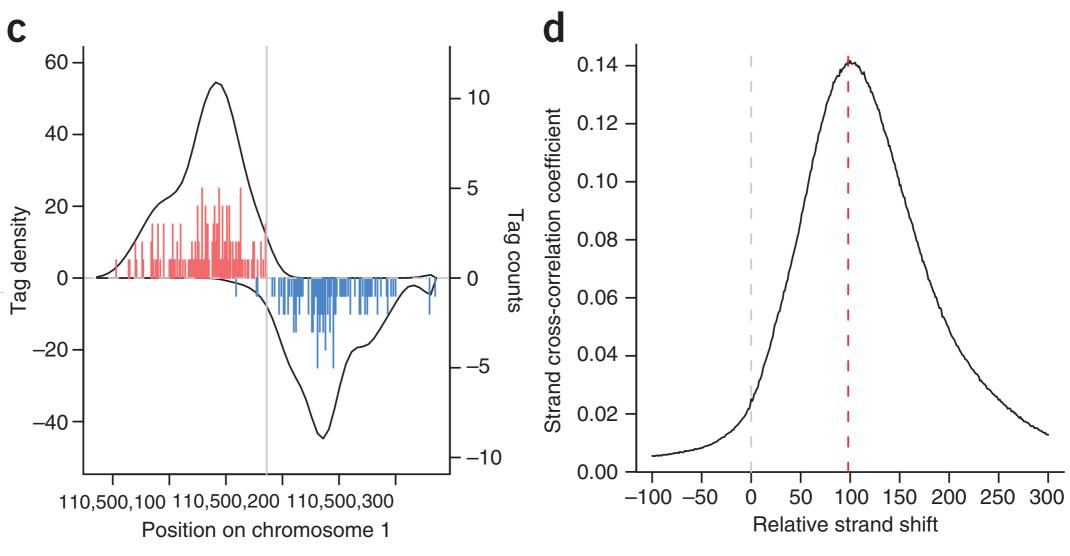
however not absolutely necessary





Kharchenko 2008

The estimation of the length of the ChIP fragments



- Kharchenko 2008
 - Binning visualization and signal distribution analysis
 - Quality control check
 - Peak finding

Fragment length estimation - quality controls

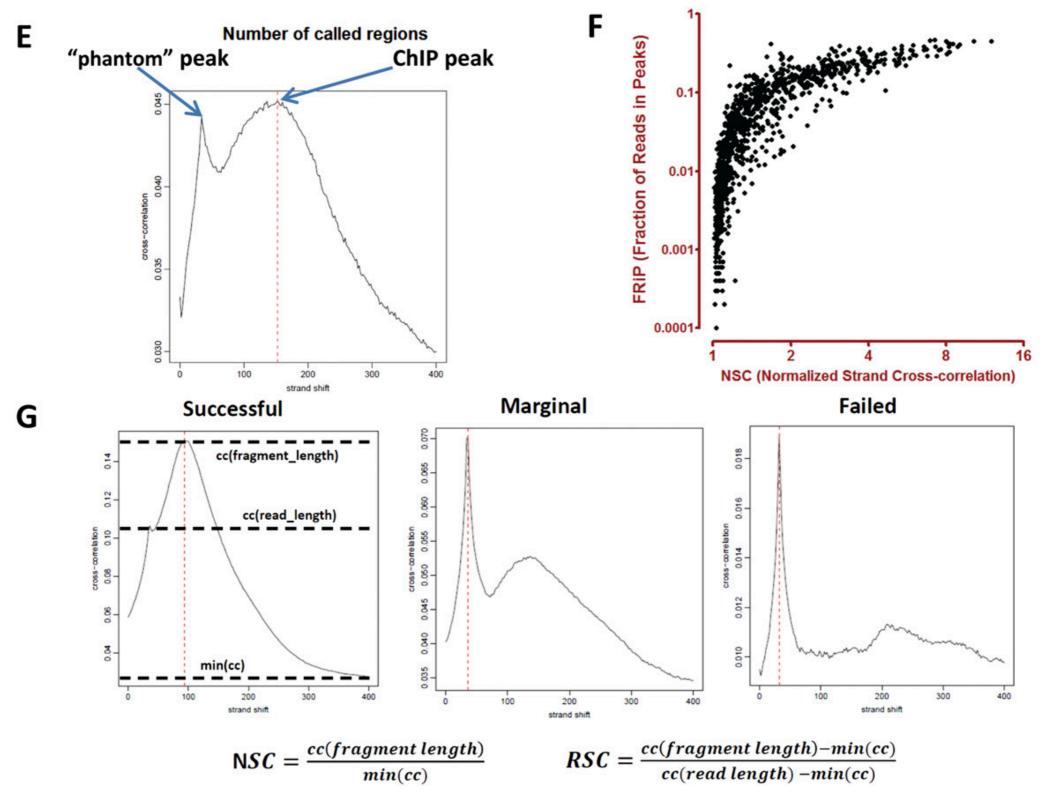
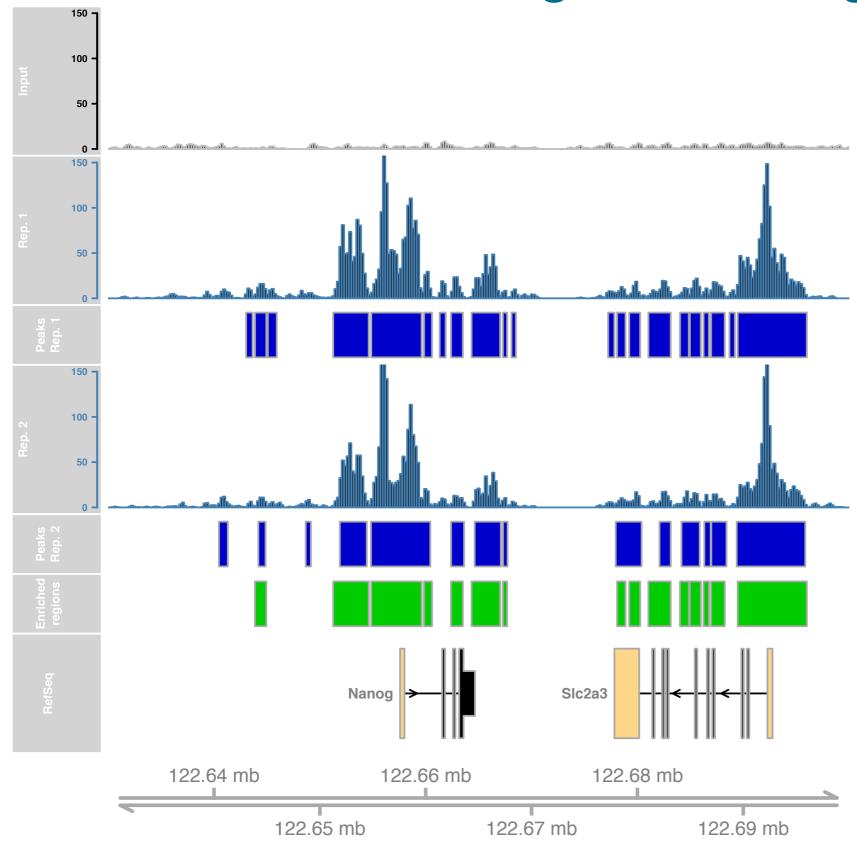


Figure 4. (Legend on next page)

ChIP-seq: considerations for study design

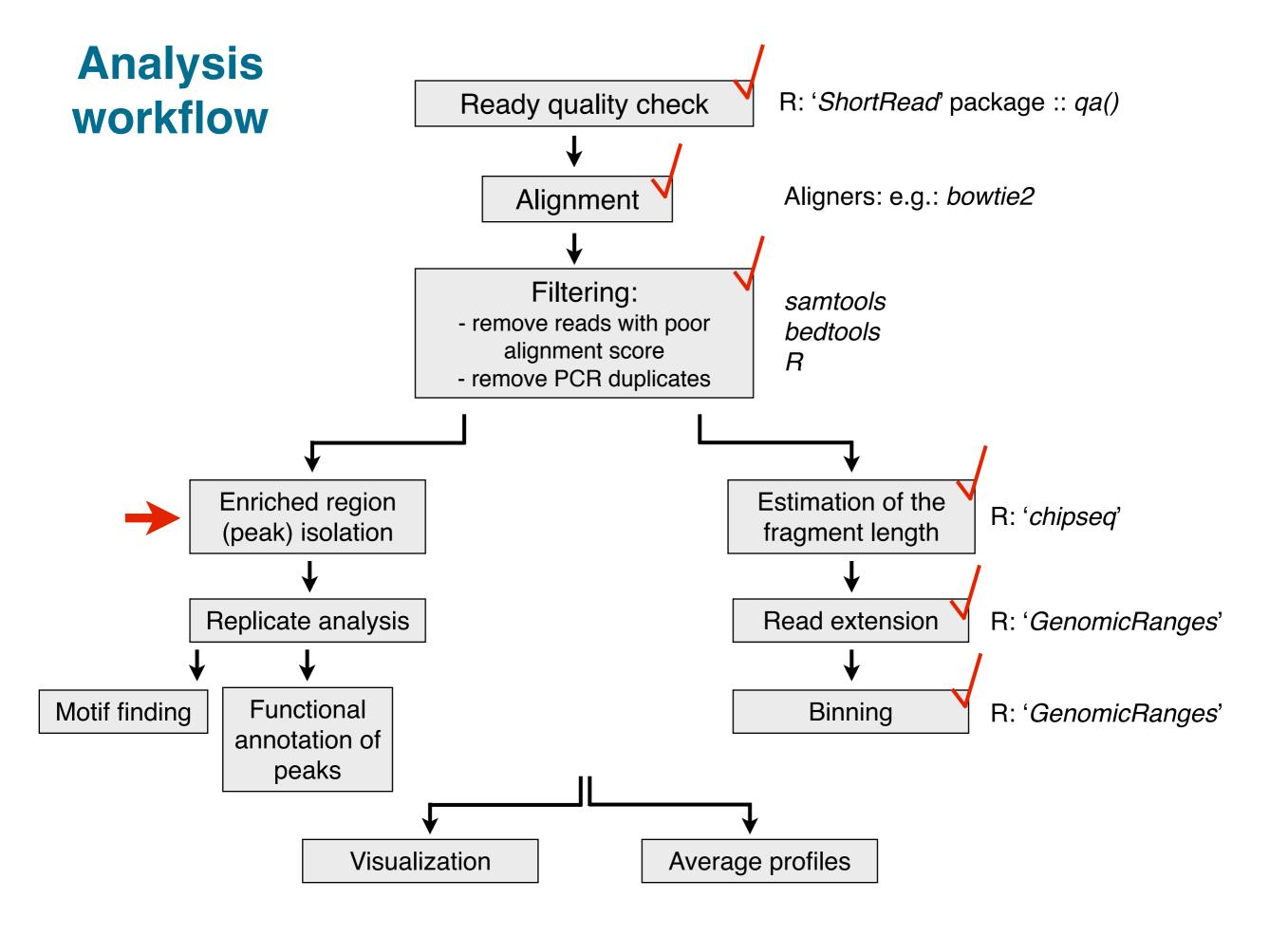
- IgG control (pros and cons)
- Input control
- Biological replication

Finding enriched regions



Enriched regions ('peaks') - regions with signal which is significantly higher than the background - input or IgG

Input reads - background reads' distribution exhibits a degree of clustering that is significantly greater than expected from a homogenous Poisson process (*P*-value< 10⁻⁶, Kharchenko et al., 2008)



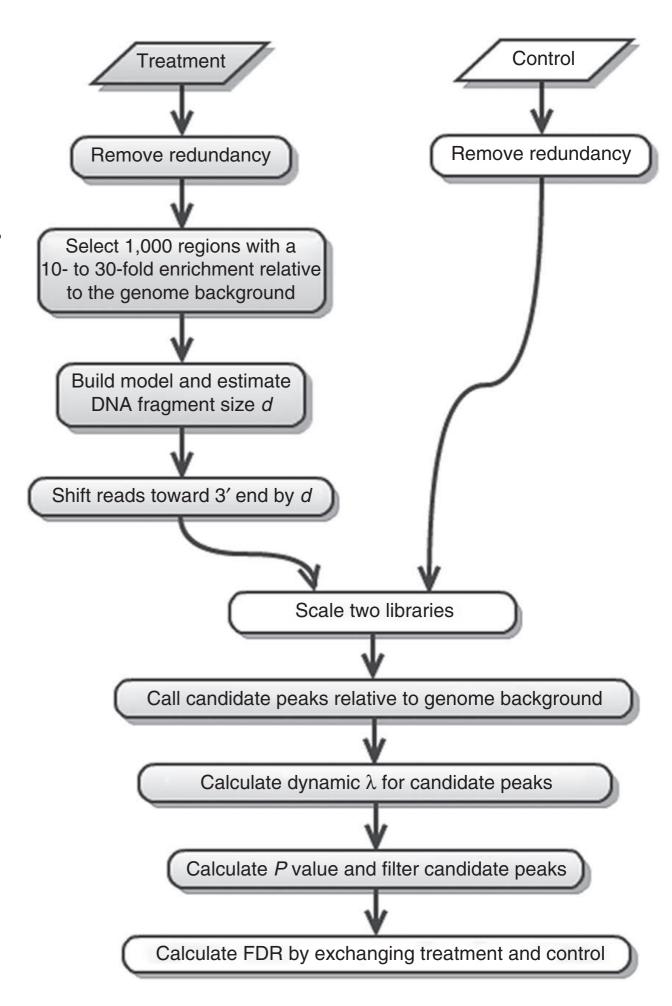
Model-based analysis of ChIP-seq (MACS)

Method

Model-based Analysis of ChIP-Seq (MACS)

Yong Zhang**, Tao Liu**, Clifford A Meyer*, Jérôme Eeckhoute†, David S Johnson*, Bradley E Bernstein§¶, Chad Nusbaum¶, Richard M Myers¥, Myles Brown†, Wei Li# and X Shirley Liu*

- removes PCR duplicates
- *d* is estimated by picking highly enriched regions and looking at the distance between modes of positive and negative strand read pileups. Reads are extended towards this midpoint (building peak model)
- Sliding window of 2d to find significantly enriched bins using λ_{local} . We obtain enrichment P-value
- eFDR by swapping control and treatment



Several examples of peak callers

SICER - designed to deal with histone type data

PeakSeq, chromHMM ...



MOSAiCS - suitable for TF and histone modification data

BayesPeak - suitable for TFs and histone modifications displaying peaklike signal

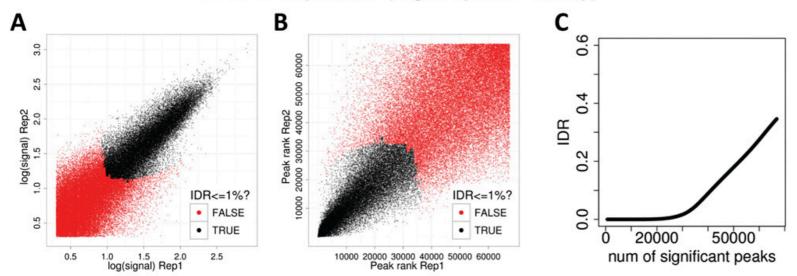
ChIPseqR - suitable for nucleosome positioning analysis

PICS CSAR NarrowPeaks CSSP

Peak processing - quality controls

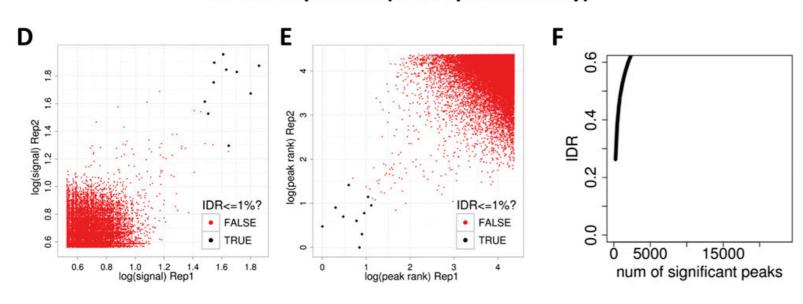
- how do we decide whether samples and peaks are OK?

RAD21 Replicates (high reproducibility)

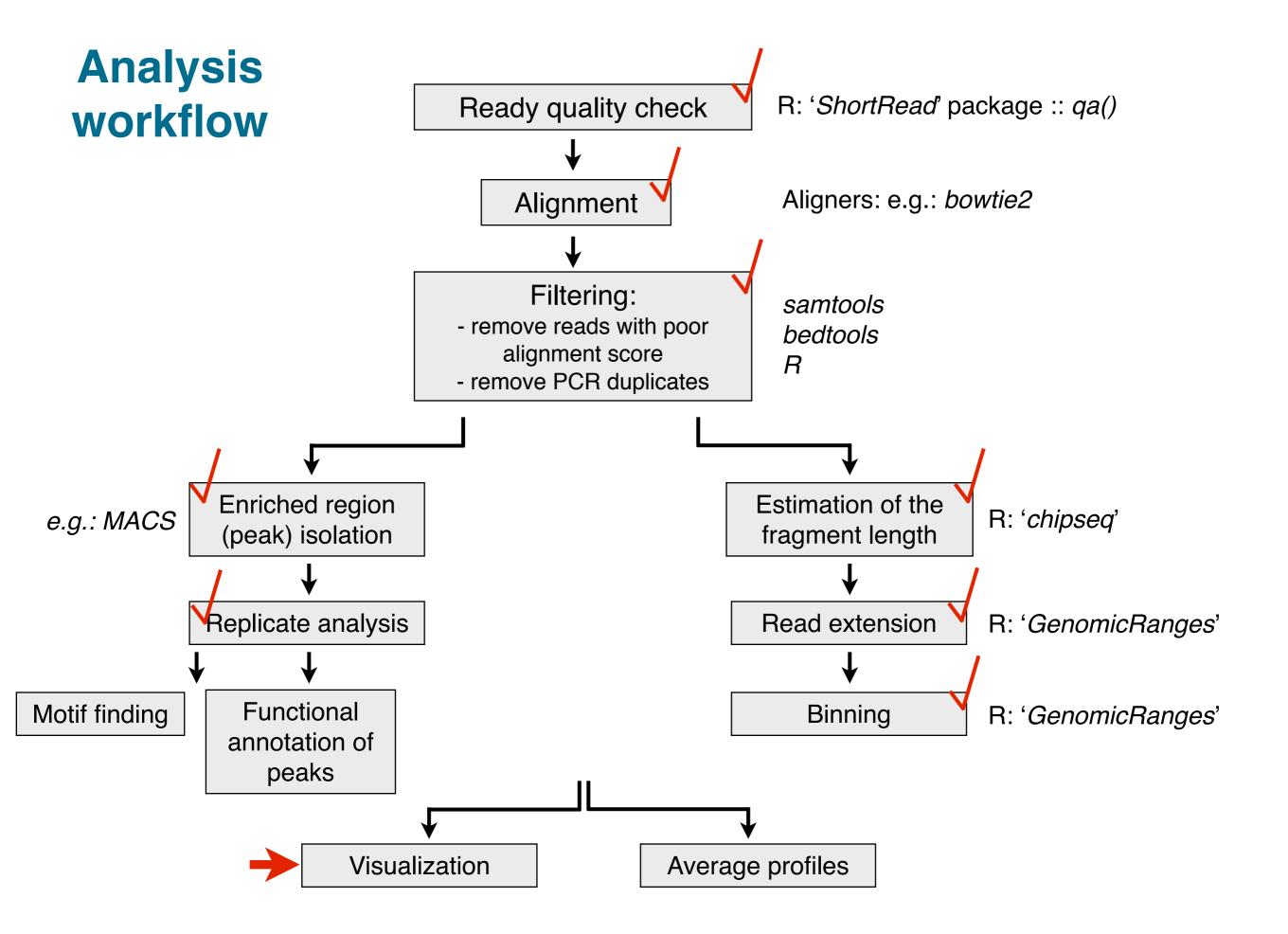


The irreproducible discovery rate (**IDR**, Li 2011) - rank peaks and assess for consistency

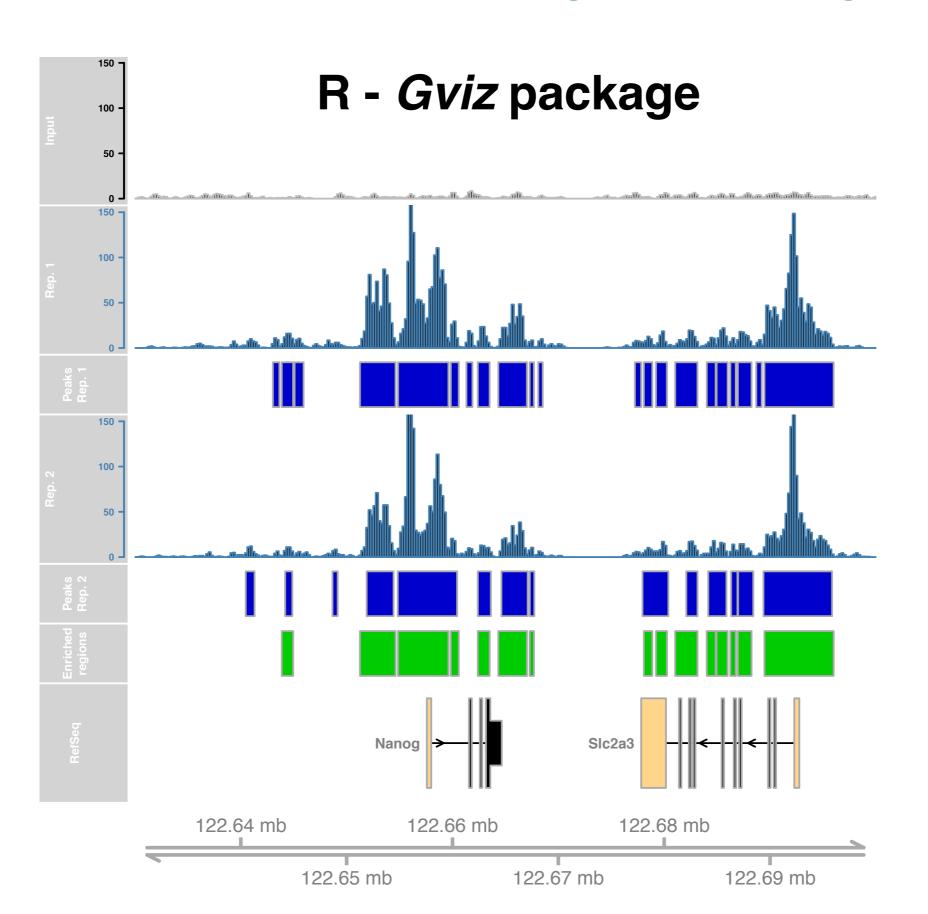
SPT20 Replicates (low reproducibility)



Distinct and strong peaks are often called by most of peak finding software
Low strength peaks are often noisy

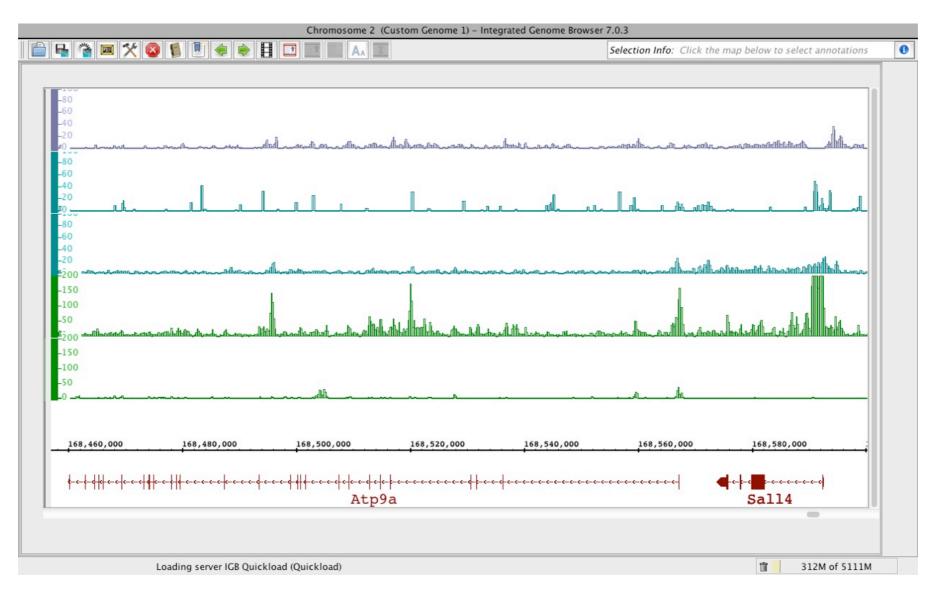


Visualization - seeing is believing



Visualization - other tools

IGB - Integrated Genome Browser - http://bioviz.org/igb/index.html



IGV - Integrative Genomics Viewer https://www.broadinstitute.org/igv/

Visualization - file formats

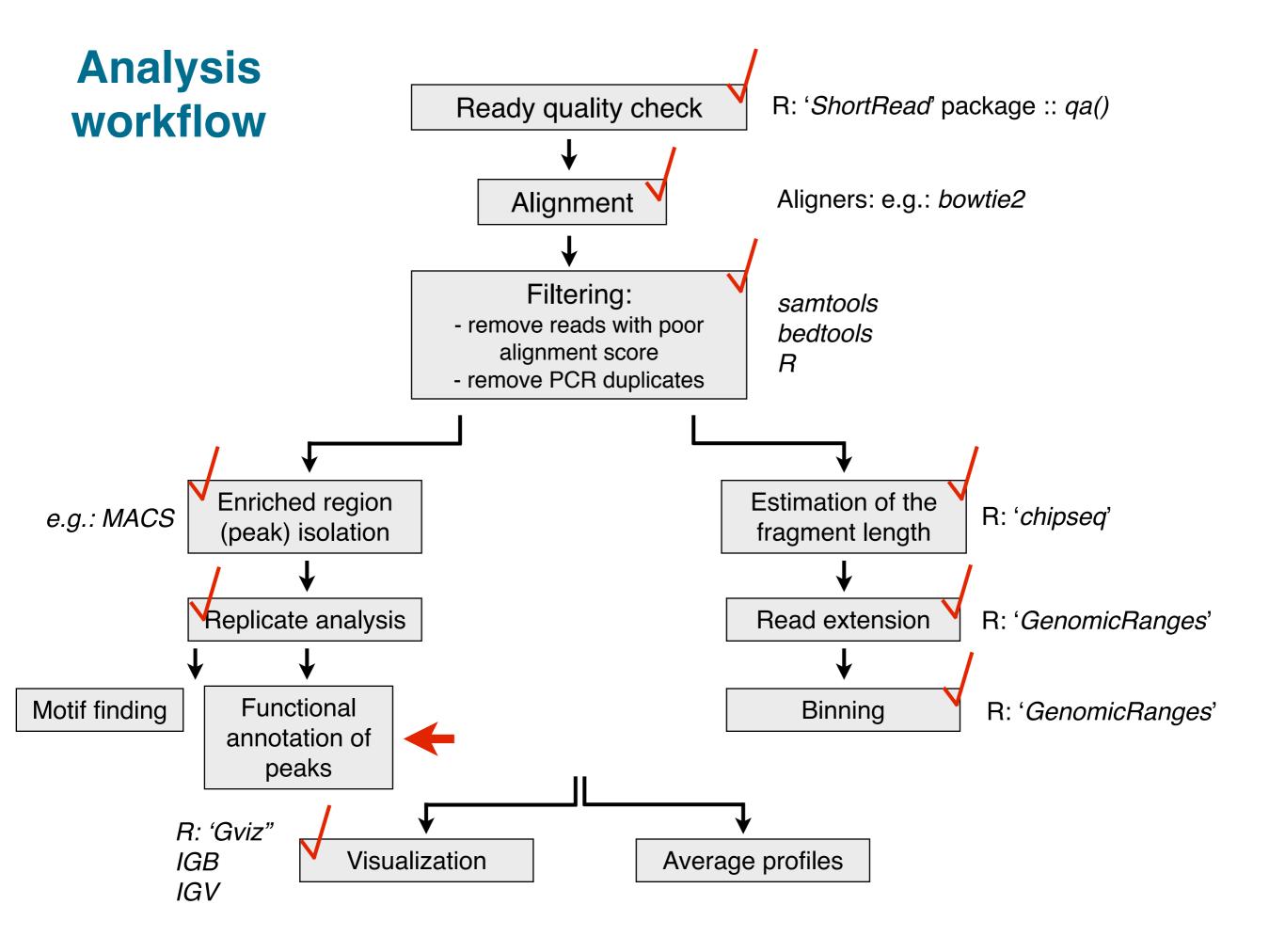
Binned or not data

 $egin{array}{c} R \\ \longrightarrow \end{array}$

.bed
.bedGraph

.wig

.bigWig



Peak analysis

Frequently asked questions include:

- Localization of peaks with respect to functional elements in the genome (promoters, gene body, introns, transcription termination sites, intergenic regions etc.)
- Co-ocurrence between enriched regions
- The distribution of signal at the peaks

ChIPpeakAnno - provides functions performing peak annotation to promoters etc.

biomaRt - easy access to data bases including gene annotation, sequence conservation, sequence retrieval etc.

GenomicRanges - fast comparison between genomic intervals:

findOverlaps()

countOverlaps()

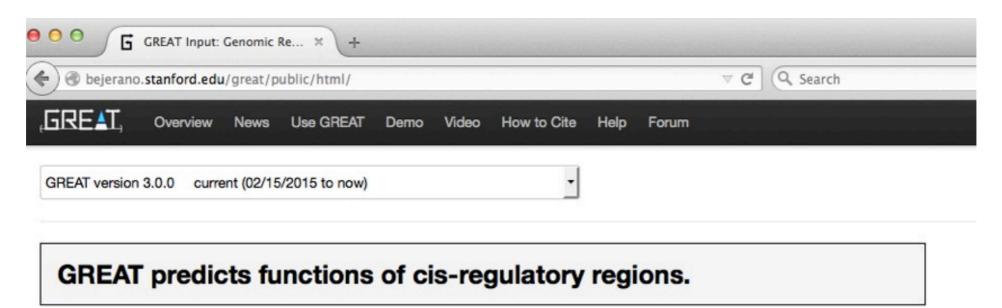
nearest()

Easy peak annotation to pre-established or new genomic features, cross-comparisons between peak locations and any kind of imaginable analysis

VennDiagram - visualization of two or multi-sample overlaps

Rcade - integrates ChIP-seq analysis with differential expression

Peak analysis - GREAT tool



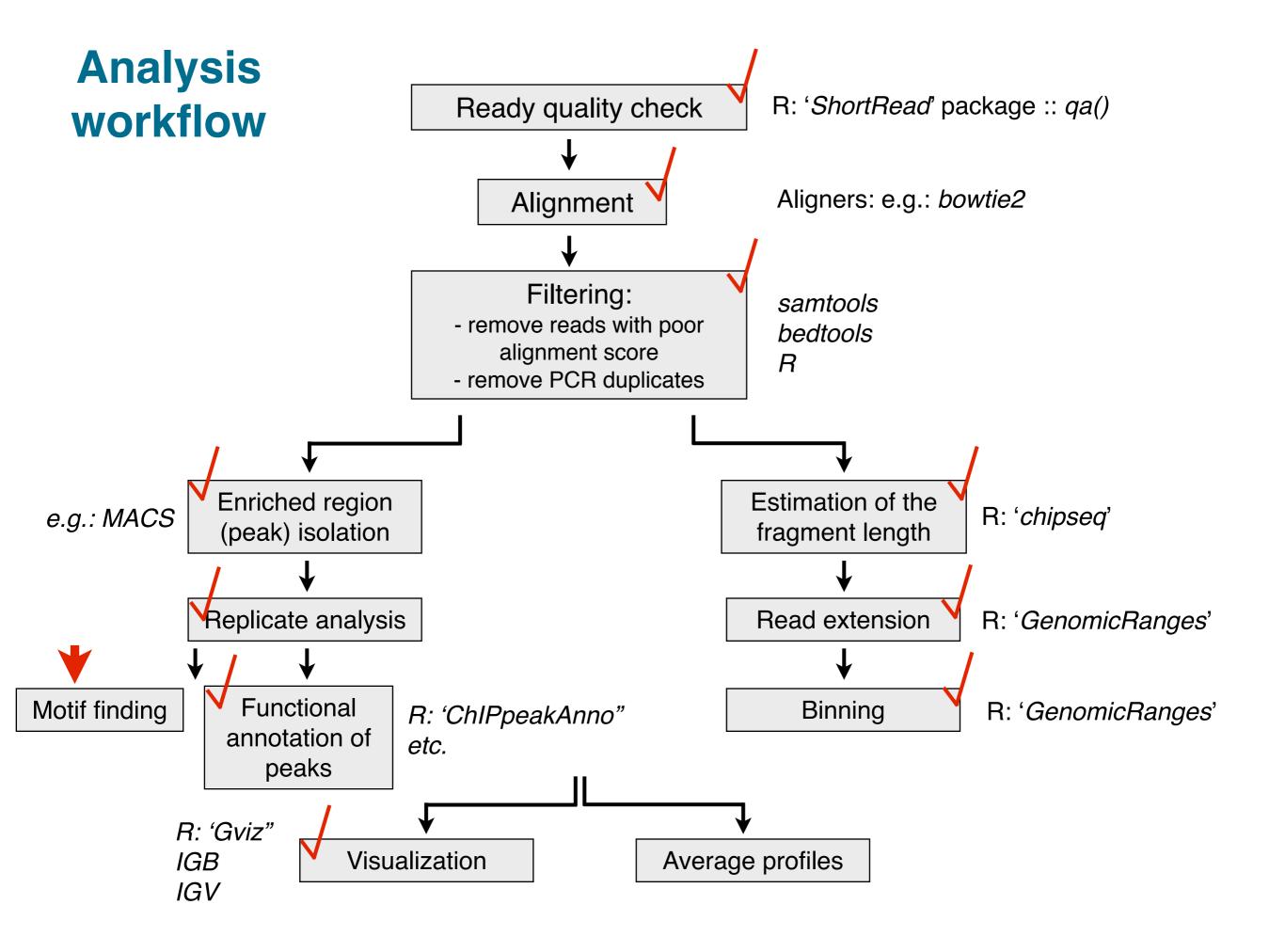
Many coding genes are well annotated with their biological functions. Non-coding regions typically lack such annotation. GREAT assigns biological meaning to a set of non-coding genomic regions by analyzing the annotations of the nearby genes. Thus, it is particularly useful in studying cis functions of sets of non-coding genomic regions. Cis-regulatory regions can be identified via both experimental methods (e.g. ChIP-seq) and by computational methods (e.g. comparative genomics). For more see our Nature Biotech Paper.

News

- Feb 15, 2015: GREAT version 3.0 switches to Ensembl genes, adds the mouse mm10 assembly, and adds new ontologies.
- Apr 3, 2012: GREAT version 2.0 adds new annotations to human and mouse ontologies and visualization tools for data exploration.
- Feb 18, 2012: The GREAT forums are released, allowing increased user-to-user interaction

More news items...





Peak analysis - motifs

MEME - provides functions performing motif discovery

RSAT - complete suite for motif finding

<u>Position Weight Matrix (PWM)</u> - describes the probability of each nucleotide at each position of a motif

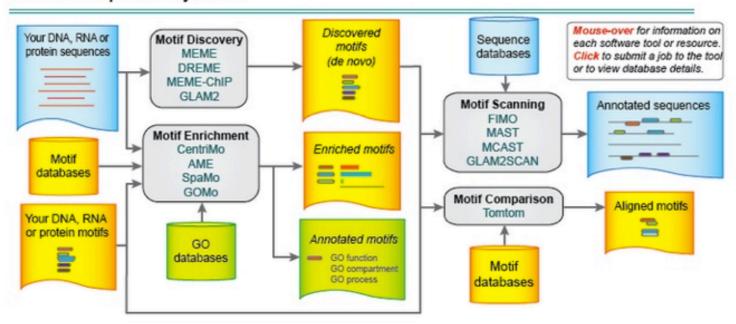
JASPAR/TRANSFAC - data bases of PWM

R: MotifDb, FIMO and others



The MEME Suite

Motif-based sequence analysis tools





















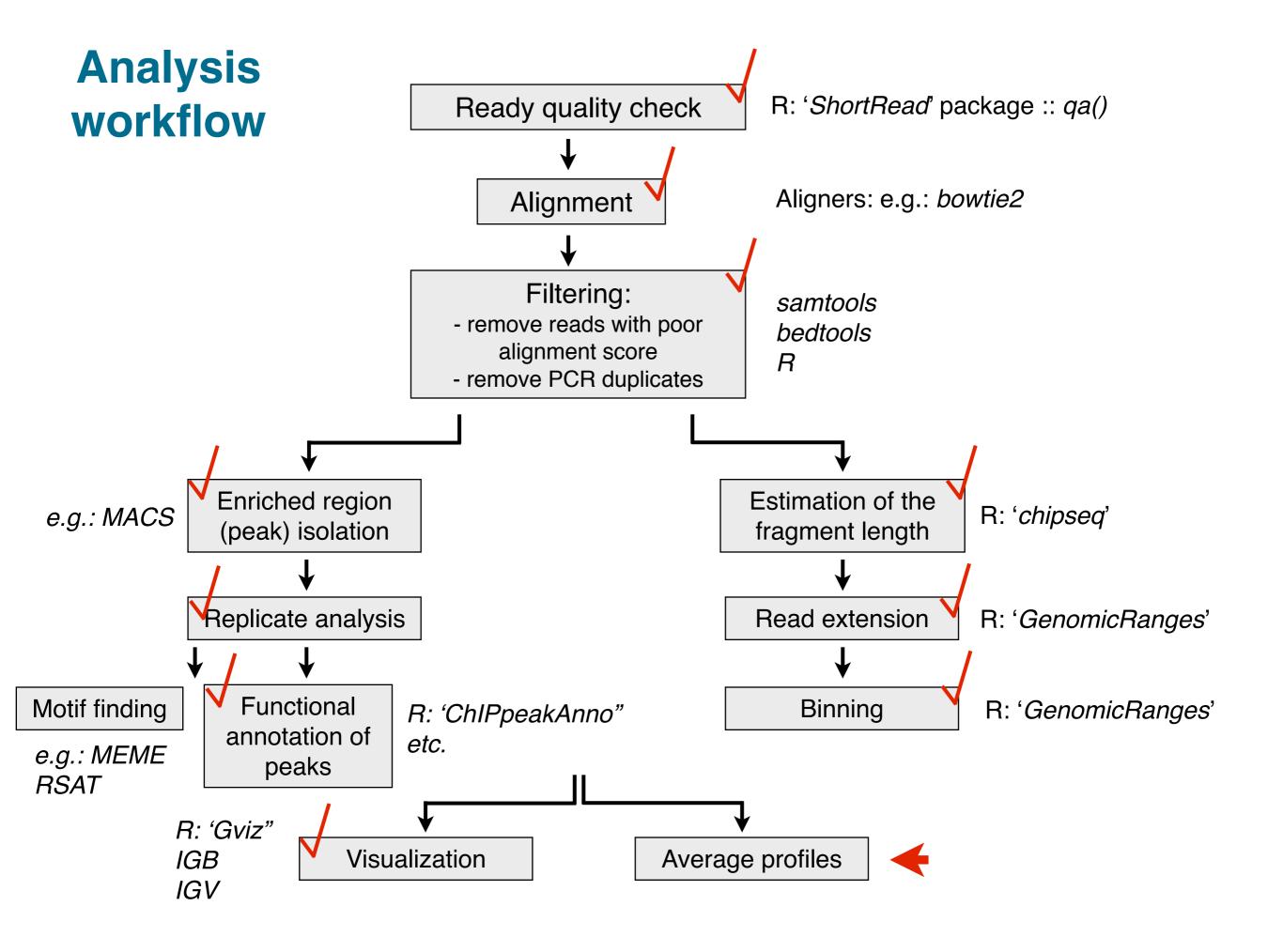




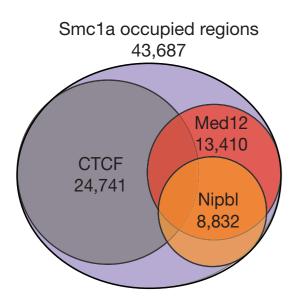




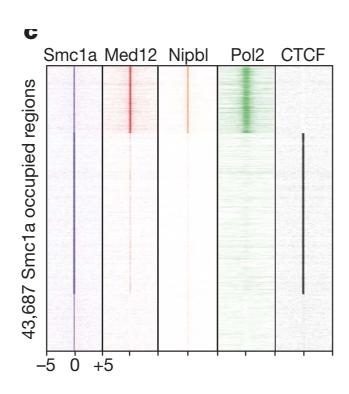


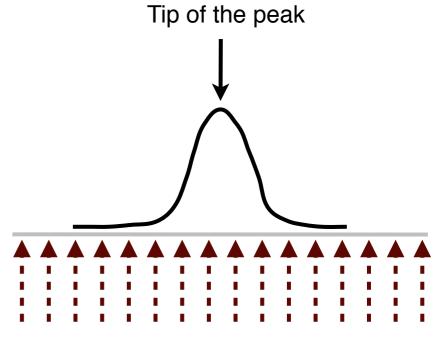


Co-enrichment and signal distribution analysis



Kagey 2010

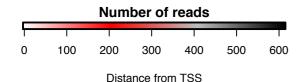


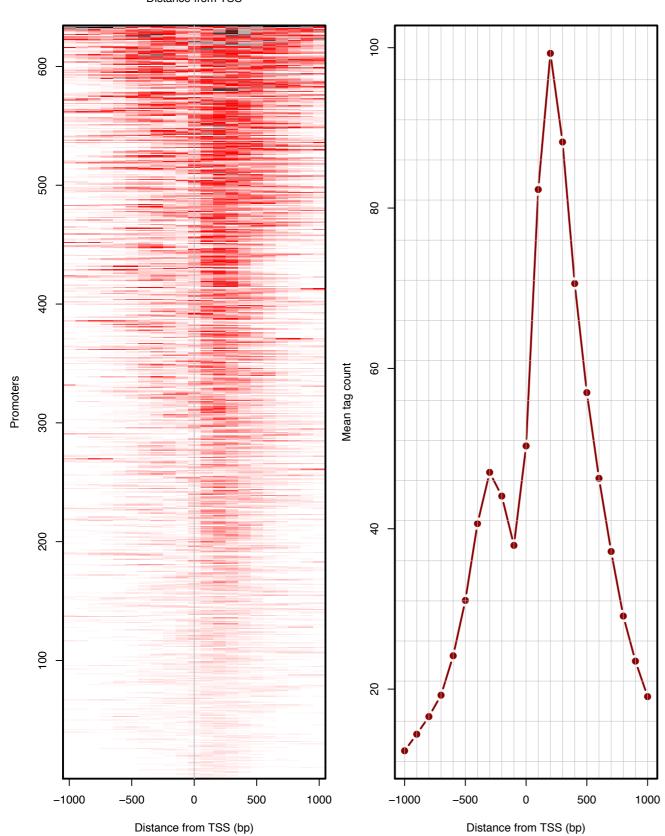


Region divided in to tiles

Count how many fragments fall into each tile

Visualization





Heatmaps of signal enrichment at - promoters

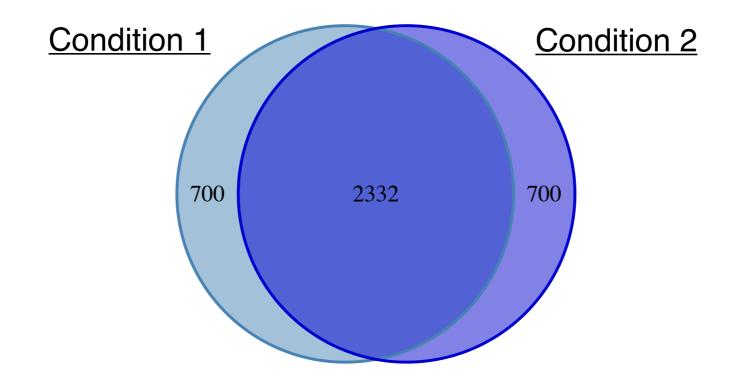
- loci enriched with factors of interest

We will see an example of such an analysis using R package

GenomicRanges

A nice alternative: *HT-Seq* (python)

Comparative peak analysis



Threshold issues affecting all qualitative analyses

Comparative peak analysis

DiffBind

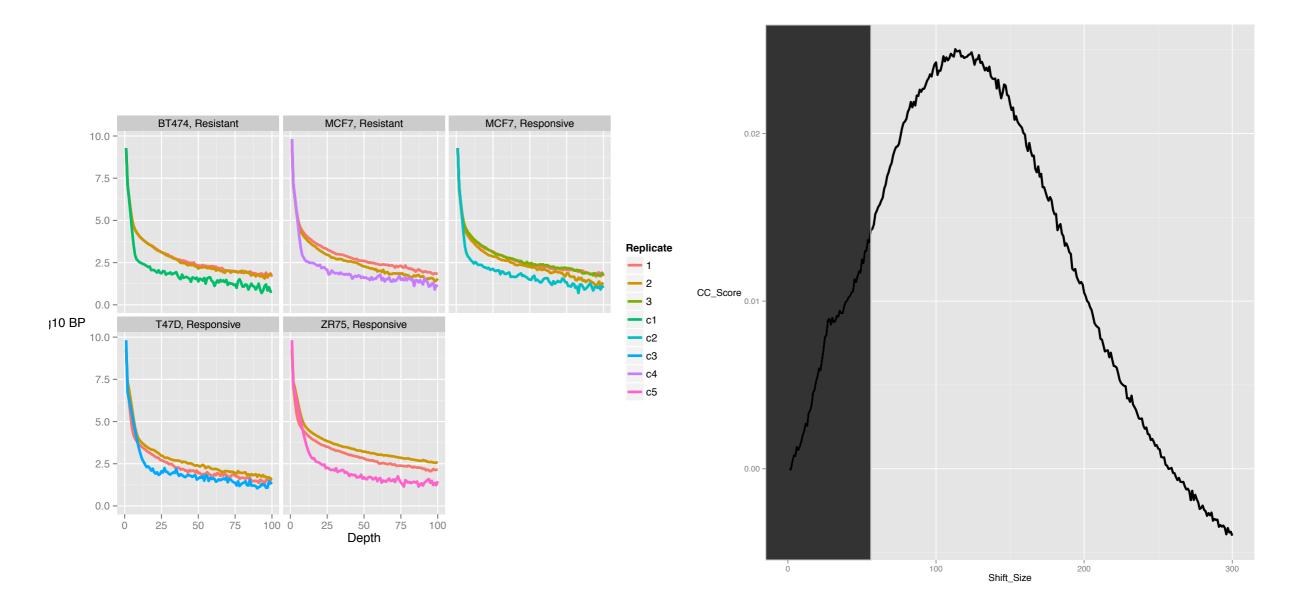
- 1. Count reads in peaks in all the replicates and conditions
- 2. Perform *edgeR* or *DESeq2* analysis *dba.analyze()*
- 3. Provides various plotting functions

<u>MMDiff</u>

- 1. Count reads in peaks in all the replicates and conditions
- 2. Performs *DESeq* normalisation
- 3. Compares peak shapes using kernel based statistical tests

ChIPQC package for quality control checks and quantitative analysis of peak strengths

- 1. Plotting coverage histograms for peaks
- 2. Cross-coverage analysis in the function of shift sizes
- 3. Plotting peak profiles
- 4. Sample clustering



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