

# Computational Pipeline for ChIP-Seq Data Analysis

Minghui Wang, Qi Sun  
Bioinformatics Facility  
Institute of Biotechnology

# Outline

ChIP-Seq experimental design

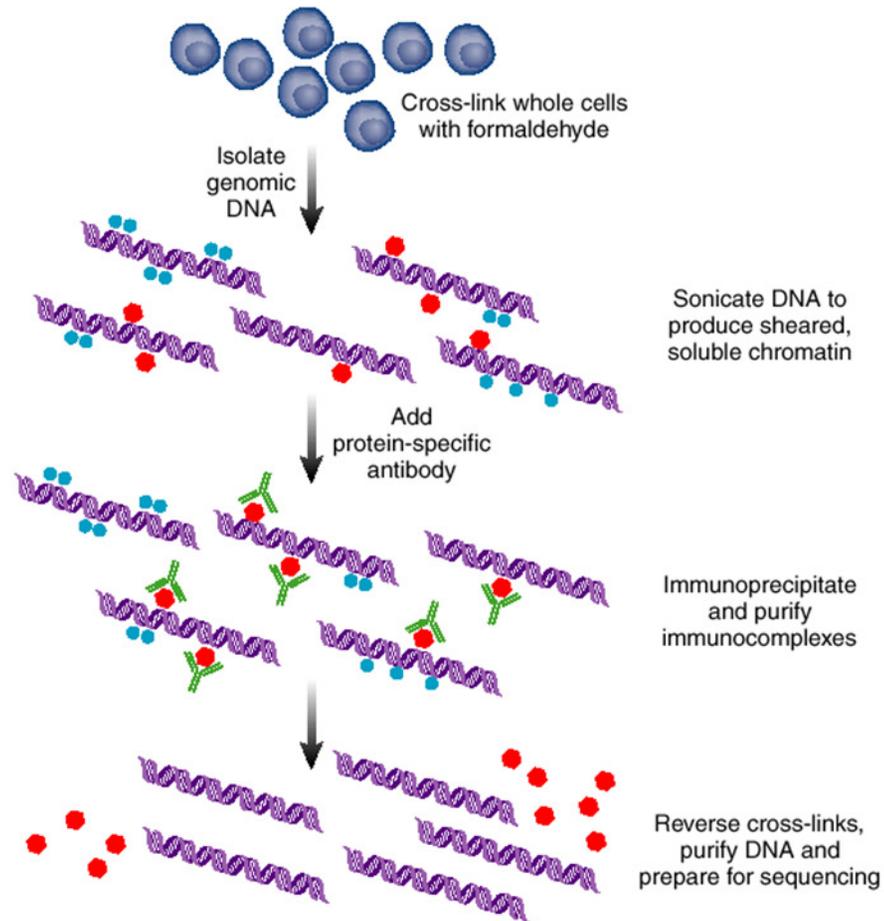
Data analysis

- Sequencing data evaluation
- Peak calling & evaluation
- GLM model for multiple replicates

Downstream analysis

- Peak annotation
- Function enrichment

# ChIP-seq workflow

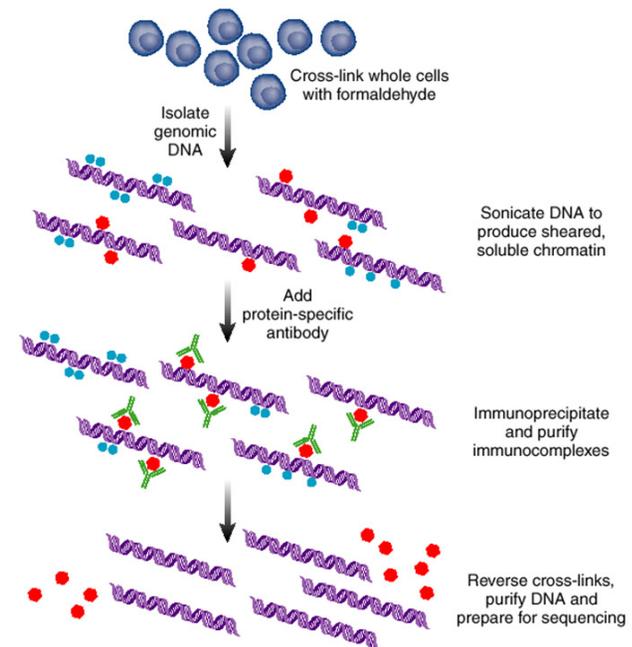


# Controls for ChIP-seq

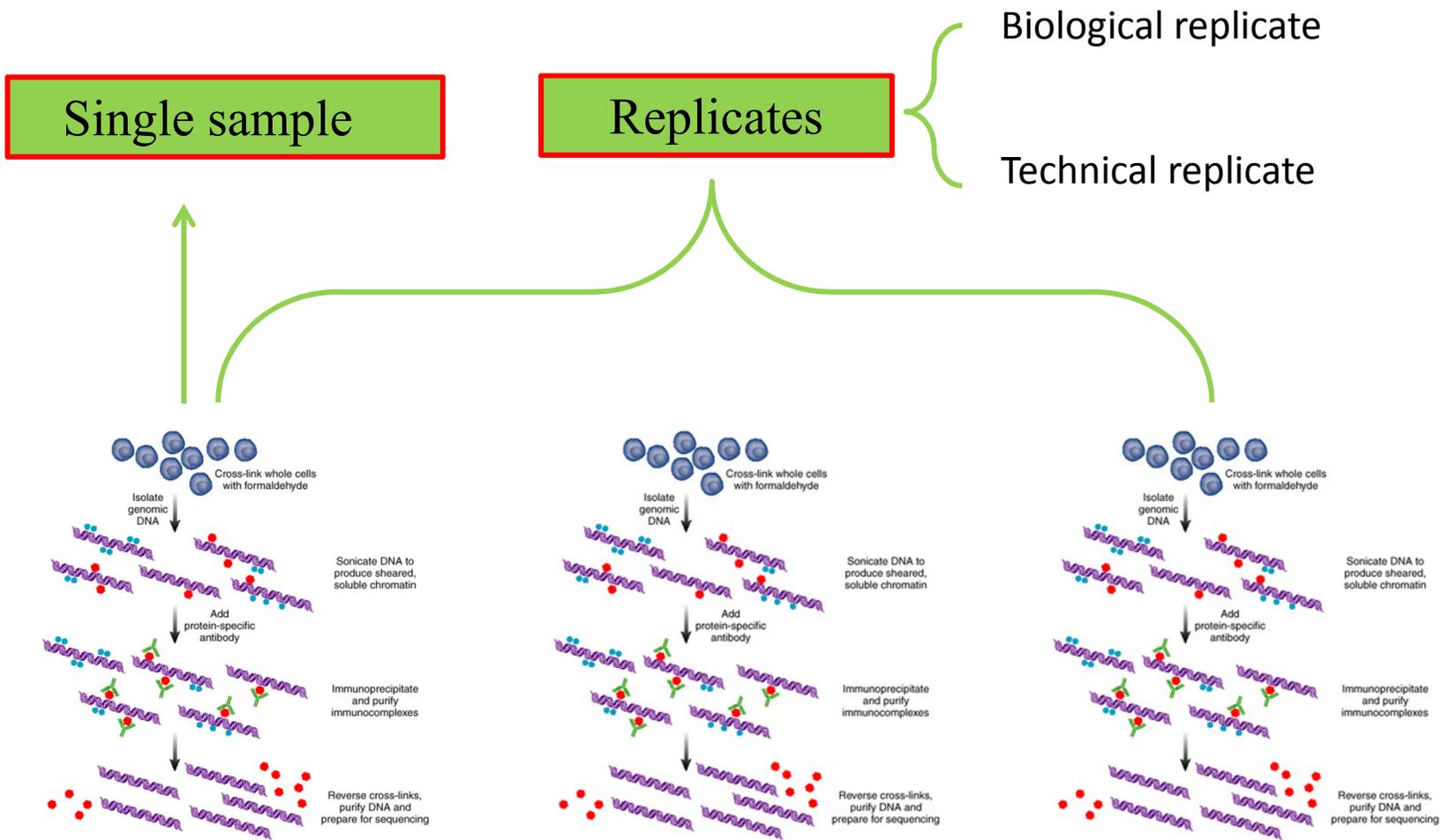
Most experimental protocols involve a control sample that is processed the same way as the test sample except that no immunoprecipitation step or no specific antibody

## Input DNA & IgG

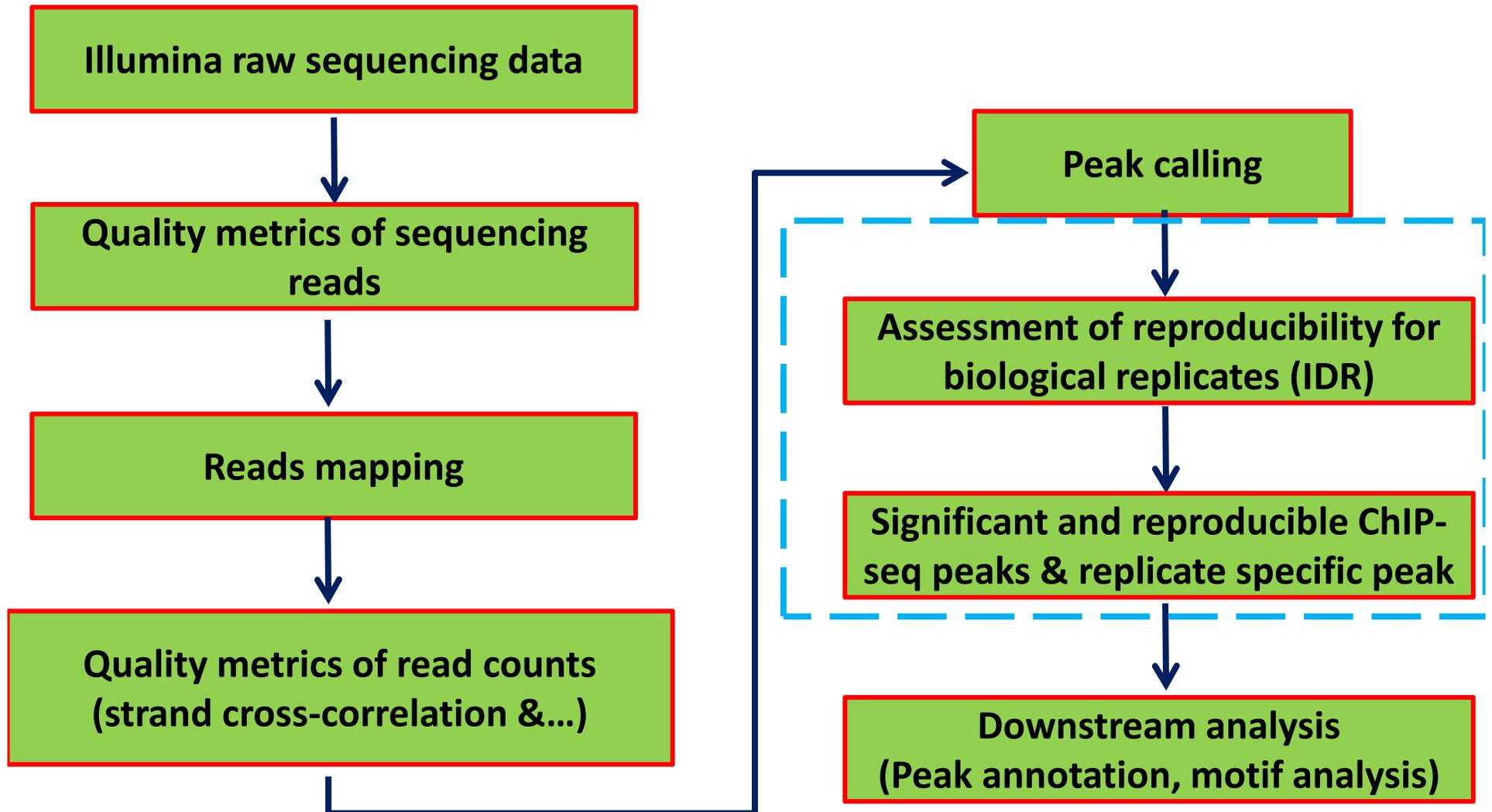
- Input DNA does not demonstrate “flat” or random (Poisson) distribution.
- Open chromatin regions tend to be fragmented more easily during shearing.
- Amplification bias.
- Mapping artifacts-increased coverage of more “mappable” regions (which also tend to be promoter regions) and repetitive regions due to inaccuracies in number of copies in assembled genome.



# ChIP-Seq experimental design

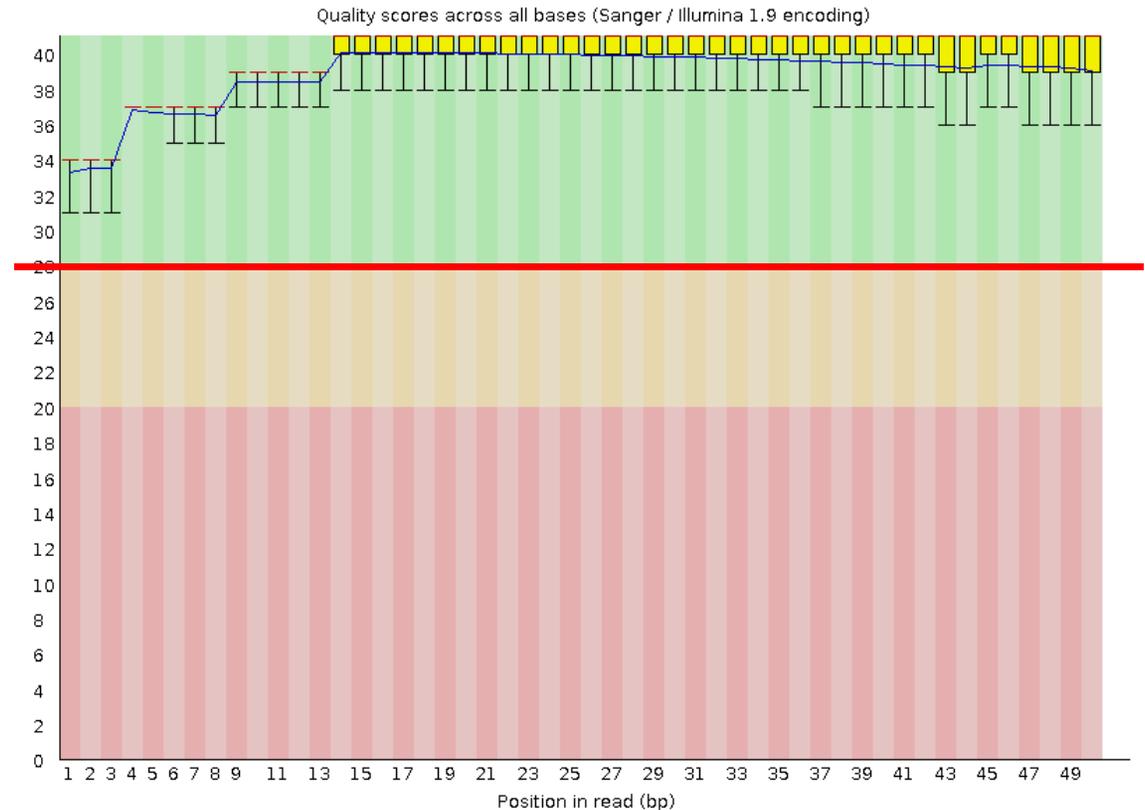


# Data analysis protocol



# Quality metrics of sequencing reads

- FastQC can be used for an overview of the data quality
- Phred quality scores used for trimming low quality bases
- $P = 10^{(-Q/10)}$ ;  $Q=30$  base is called incorrectly 1 in 1000





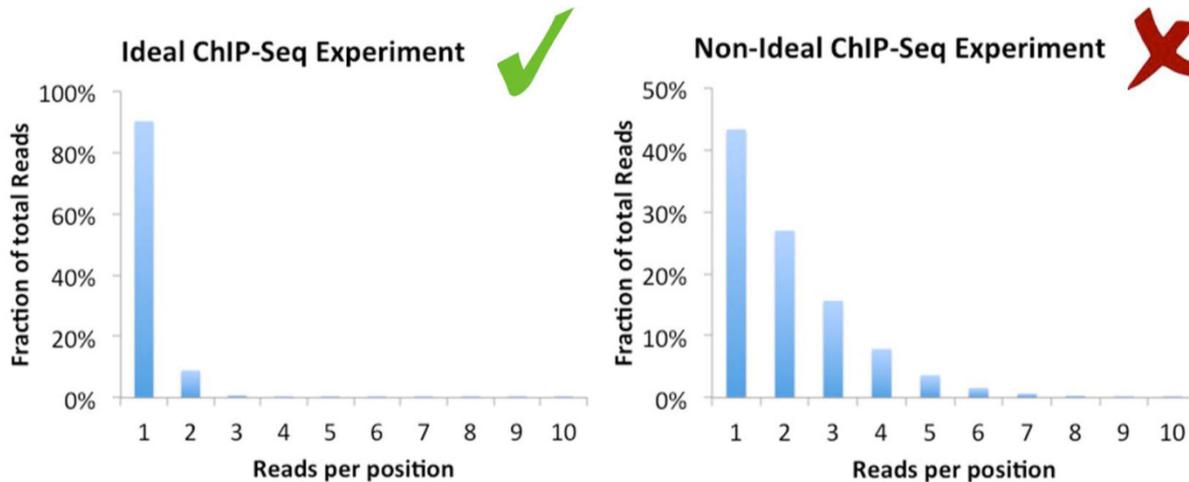


# Quality Control

## ➤ Nonredundant fraction (NRF)

$$\text{NRF} = \frac{\text{\#unique start positions of uniquely mappable reads}}{\text{\#uniquely mappable reads}}$$

ENCODE recommends target of NRF 0:8 for 10 million uniquely mapped reads



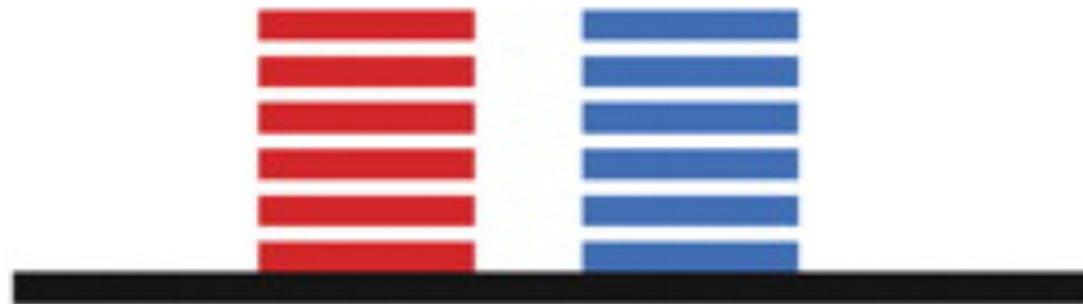
```
sh run_bam2bed.sh
perl CalbedNRF.pl rep5_D12K4.txt_trim_uniq_sorted_bamtobed.bed
```

<https://github.com/mel-astar/mel-ngs/tree/master/mel-chipseq/chipseq-metrics>

# Quality Control



**Typical ChIP-seq peak**

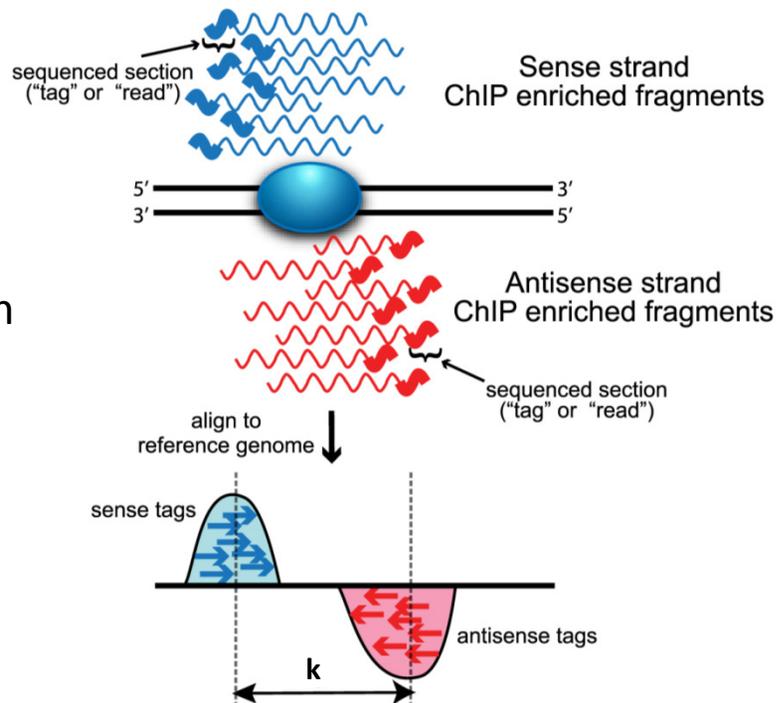


**Low-complexity ChIP-seq peak**

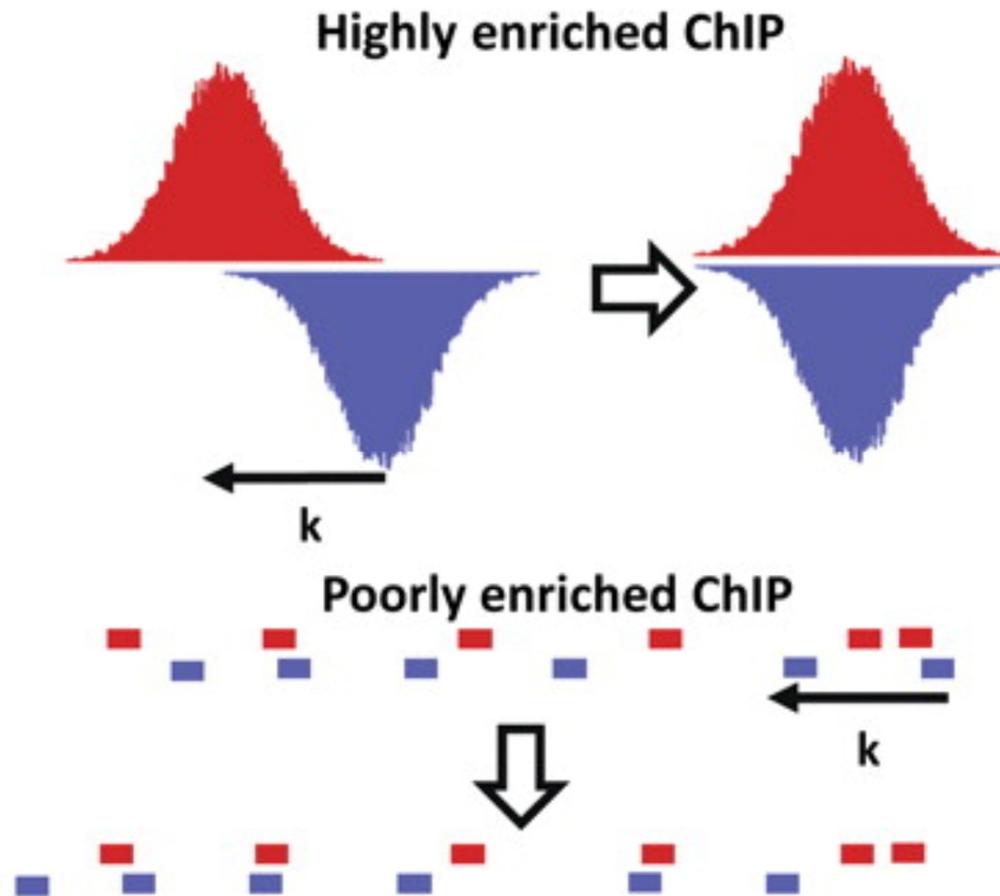
# Cross-correlation

DNA fragments from a chromatin immunoprecipitation experiment are sequenced from the 5' end.

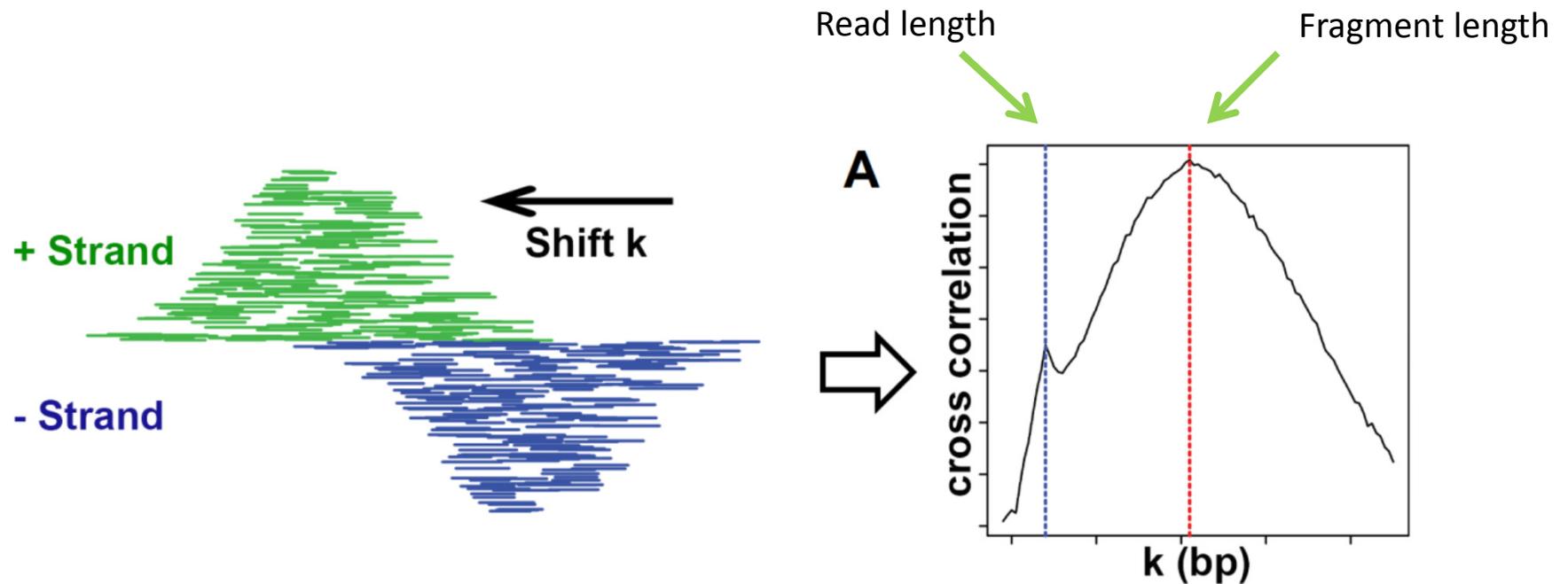
- With ChIP-seq, the alignment of the reads to the genome results in two peaks (one on each strand) that located on flanking sides of the protein or nucleosome of interest.
- The distance between strands specific peaks ( $k$ ) represents the average sequenced fragment.



# Cross-correlation



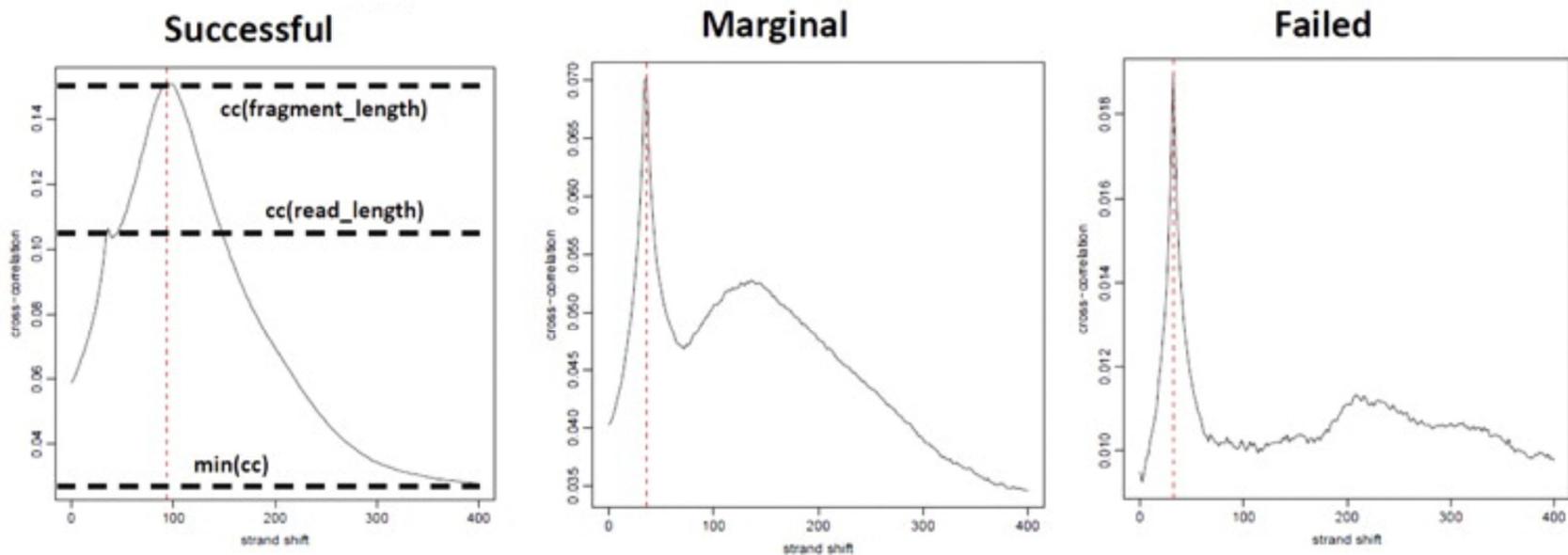
# Cross-correlation



Strand cross-correlation is computed as the Pearson correlation between the positive and the negative strand profiles at different strand shift distances,  $k$

<https://sites.google.com/a/brown.edu/bioinformatics-in-biomed/spp-r-from-chip-seq>

# Cross-correlation



$$NSC = \frac{cc(fragment\ length)}{min(cc)}$$

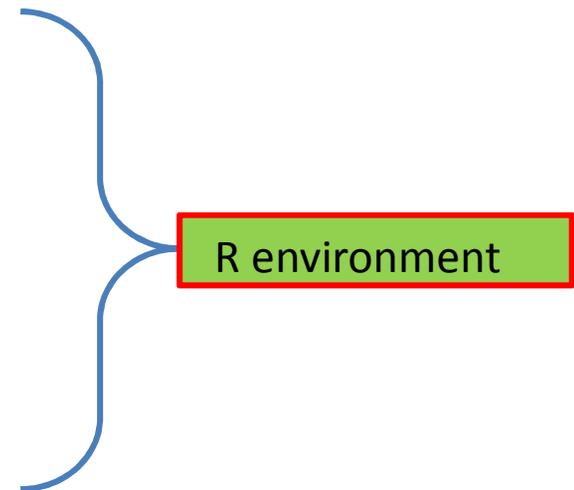
$$RSC = \frac{cc(fragment\ length) - min(cc)}{cc(read\ length) - min(cc)}$$

NSC values < 1.05 and RSC values < 0.8

<http://code.google.com/p/phantompeakqualtools/>

# Peak calling software

- MACS → Yong Zhang et al
- cisGenome → Hongkai Ji et al
- spp → Peter Park et al
- Rbrads → Julie Ahringer et al
- BayesPeak → Simon Tavaré et al
- ...



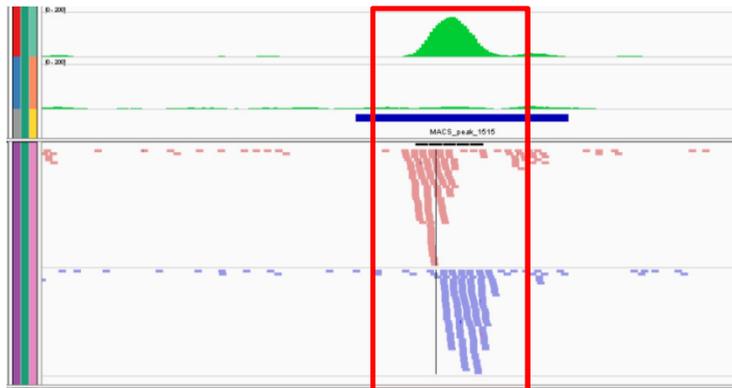
# Step1 of MACS2

## ➤ Estimating fragment length $d$

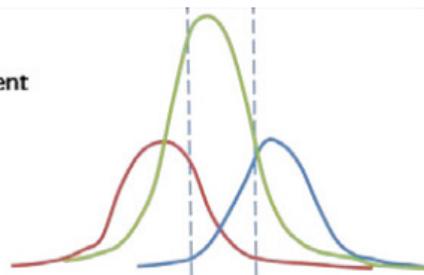
Slide a window of size  $2 \times \text{BANDWIDTH}$ , this value based on sonication size first

Keep top regions with **MFOLD** enrichment of treatment vs. control

Plot average +/- strand read densities  $\rightarrow$  estimate  $d$



Reads are shifted by  $d/2$   
toward the 3' ends, fragment  
are then added



---

### Algorithm 1 Estimate Fragment Size

---

- 1: Slide a window of  $2 \times \text{bandwidth}$  across genome
  - 2: Identify regions of moderate enrichment (mfold: 10-30 fold)
  - 3: **for each** peak  $i$  of 1000 randomly chosen enriched regions  
**do**
  - 4:   separate reads into + and - strand
  - 5:   Calculate mode of + and - summit
  - 6:    $d_i \leftarrow |\text{mode}_+ - \text{mode}_-|$
  - 7: **end for**
  - 8:  $d \leftarrow \text{average}_i(d_i)$
-

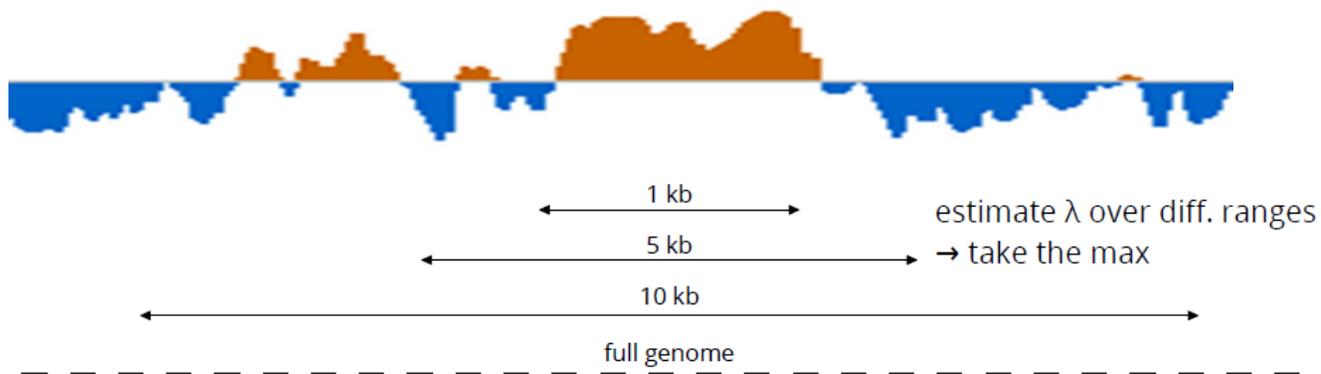
# Step2 of MACS2

## ➤ Identification of local noise parameter

shifting all reads by  $d/2$

slide a window of size  $2*d$  across treatment and input

estimate parameter  $\lambda_{\text{local}}$  of Poisson distribution



# Step3 of MACS2

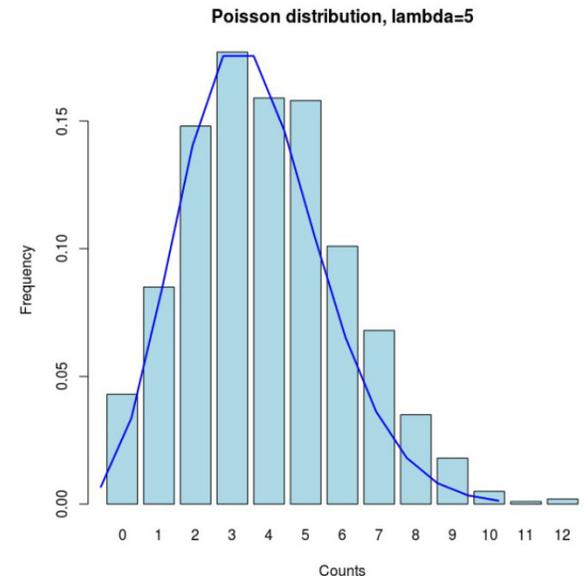
## ➤ Peaks identification

$$\lambda = \frac{\ell \times N}{G^*}$$

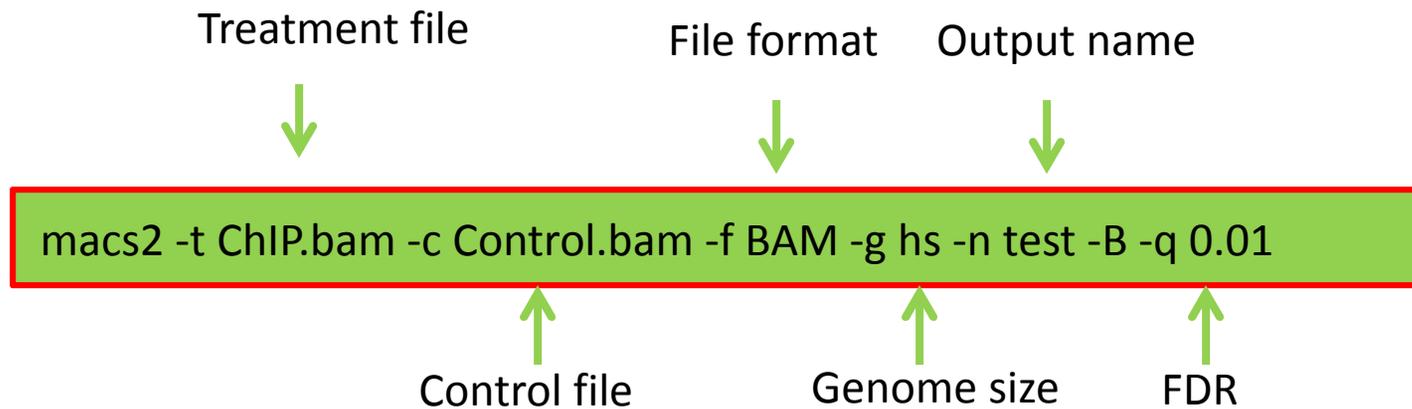
$$P(H \geq h) = \sum_{k=h}^{\infty} \frac{e^{-\lambda} \lambda^k}{k!} = 1 - \sum_{k=0}^{h-1} \frac{e^{-\lambda} \lambda^k}{k!}$$

$$p(j) \leq \delta \frac{j}{m}$$

<http://liulab.dfci.harvard.edu/MACS/index.html>



# Command of MACS2



## Option:

- s TSIZE, tsize=TSIZE
- m MFOLD, --mfold=MFOLD
- bw=BW
- nomodel
- shiftsize=SHIFTSIZE

# Output of MACS2

```
# This file is generated by MACS version 2.0.9 20111102 (tag:alpha)
# ARGUMENTS LIST:
# name = rep3_D2K4_H3
# format = AUTO
# ChIP-seq file = ../rep3_D2K4.txt_trim_uniq_sorted.bam
# control file = ../combinerep3_D2H3_sorted.bam
# effective genome size = 9.00e+07
# band width = 150
# model fold = 2,10
# qvalue cutoff = 1.00e-02
# Larger dataset will be scaled towards smaller dataset.
# Range for calculating regional lambda is: 1000 bps and 10000 bps
# Broad region calling is off
```

← Input files and parameters setting

```
# tag size is determined as 46 bps
# total tags in treatment: 8568994
# tags after filtering in treatment: 7814916
# maximum duplicate tags at the same position in treatment = 1
# Redundant rate in treatment: 0.09
# total tags in control: 28632645
# tags after filtering in control: 21760444
# maximum duplicate tags at the same position in control = 1
# Redundant rate in control: 0.24
# d = 150
```

chr	start	end	length	abs_summit	pileup	-log10(pvalue)	fold_enrichment	-log10(qvalue)
I	4060	4291	232	4148	51.00	15.47	3.92	13.65
I	16621	16867	247	16728	45.00	11.98	3.45	10.24
I	24154	24398	245	24267	50.00	14.87	3.84	13.06
I	24563	24868	306	24703	80.00	34.67	5.86	32.50
I	26425	27627	1203	26700	97.00	48.60	7.11	46.21
I	28284	28442	159	28355	44.00	7.72	2.55	6.13
I	30982	31131	150	31068	37.00	7.84	2.84	6.24
I	31802	32130	329	31899	46.00	10.08	2.98	8.41
I	33713	33899	187	33757	44.00	7.72	2.55	6.13
I	34606	35205	600	35057	52.00	9.08	2.59	7.44

← Peaks information

# Output of MACS2

```
[mingh@cbsumm11 H3K4]$ more rep3_D2K4_H3_peaks.encodePeak
track type=narrowPeak nextItemButton=on
```

I	4059	4291	MACS_peak_1	136	.	3.92	15.47	13.65	88	
I	16620	16867	MACS_peak_2	102	.	3.45	11.98	10.24	107	
I	24153	24398	MACS_peak_3	130	.	3.84	14.87	13.06	113	
I	24562	24868	MACS_peak_4	325	.	5.86	34.67	32.50	140	
I	26424	27627	MACS_peak_5	462	.	7.11	48.60	46.21	275	Enrichment score (fold-change)
I	28283	28442	MACS_peak_6	61	.	2.55	7.72	6.13	71	
I	30981	31131	MACS_peak_7	62	.	2.84	7.84	6.24	86	
I	31801	32130	MACS_peak_8	84	.	2.98	10.08	8.41	97	
I	33712	33899	MACS_peak_9	61	.	2.55	7.72	6.13	44	
I	34605	35205	MACS_peak_10	74	.	2.59	9.08	7.44	451	
I	35353	35741	MACS_peak_11	97	.	3.38	11.43	9.71	78	
I	36168	36391	MACS_peak_12	68	.	2.78	8.48	6.86	143	
I	39389	39878	MACS_peak_13	148	.	4.07	16.71	14.86	345	-log10pvalue
I	40039	40344	MACS_peak_14	71	.	2.99	8.81	7.18	214	
I	40930	41090	MACS_peak_15	53	.	2.69	6.91	5.35	69	
I	46949	47213	MACS_peak_16	180	.	4.45	19.92	18.00	135	
I	47288	47607	MACS_peak_17	124	.	3.76	14.27	12.47	203	
I	70140	70613	MACS_peak_18	354	.	6.30	37.65	35.43	135	
I	93000	93232	MACS_peak_19	62	.	2.84	7.84	6.24	100	
I	97597	98073	MACS_peak_20	305	.	5.15	32.72	30.59	292	
I	98224	98465	MACS_peak_21	180	.	4.45	19.92	18.00	123	-log10qvalue
I	107919	108073	MACS_peak_22	60	.	2.78	7.62	6.04	65	
I	108184	109005	MACS_peak_23	202	.	4.34	22.16	20.20	604	
I	109091	111927	MACS_peak_24	820	.	9.26	85.23	82.07	2142	
I	171398	171571	MACS_peak_25	45	.	2.53	6.03	4.51	79	
I	182407	182922	MACS_peak_26	307	.	5.83	32.90	30.76	378	
I	237822	237986	MACS_peak_27	81	.	3.15	9.83	8.16	65	
I	288519	289415	MACS_peak_28	496	.	7.60	52.09	49.64	299	
I	310449	310912	MACS_peak_29	148	.	4.07	16.71	14.86	145	Summit position to peak start
I	310963	311271	MACS_peak_30	90	.	3.12	10.75	9.06	145	
I	314136	315758	MACS_peak_31	659	.	8.98	68.73	65.96	782	
I	315988	316213	MACS_peak_32	81	.	3.15	9.83	8.16	90	

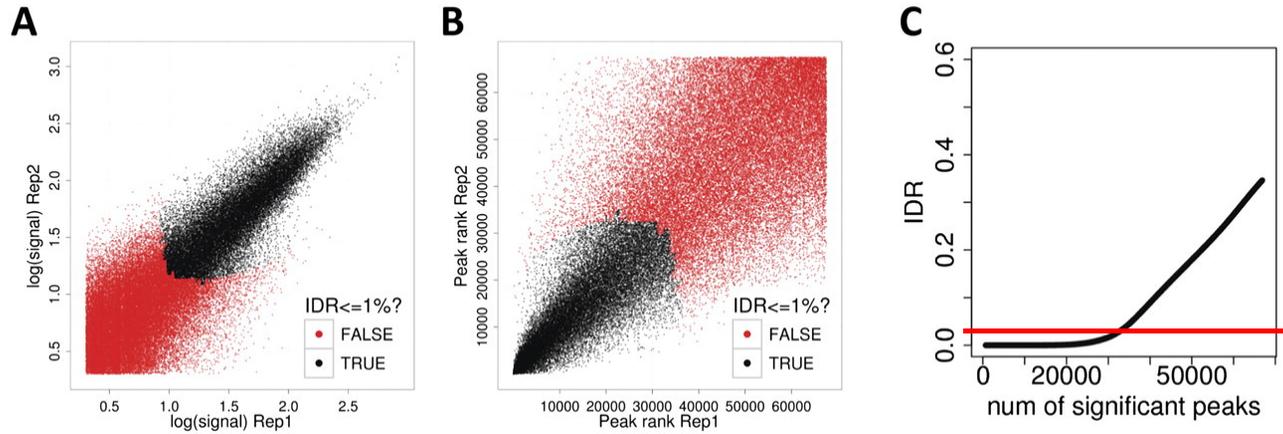
# Consistency of replicates: IDR

- IDR the irreproducible discovery rate
- Each list of peaks is ranked according to p-value or signal score
- The IDR method adopted the bivariate rank distributions over the replicates in order to separate signal from noise based on consistency and reproducibility of identifications

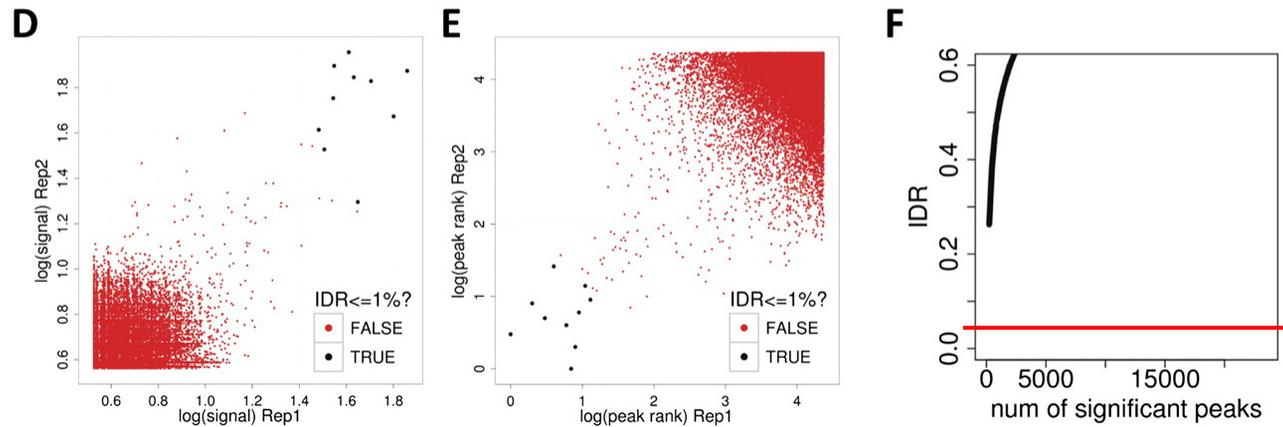
```
Rscript batch-consistency-analysis.r [peakfile1] [peakfile2] -1 [outfile.prefix] 0 F p.value  
Rscript batch-consistency-plot.r [npairs] [output.prefix] [input.file.prefix1] [input.file.prefix2] [input.file.prefix3]
```

# IDR

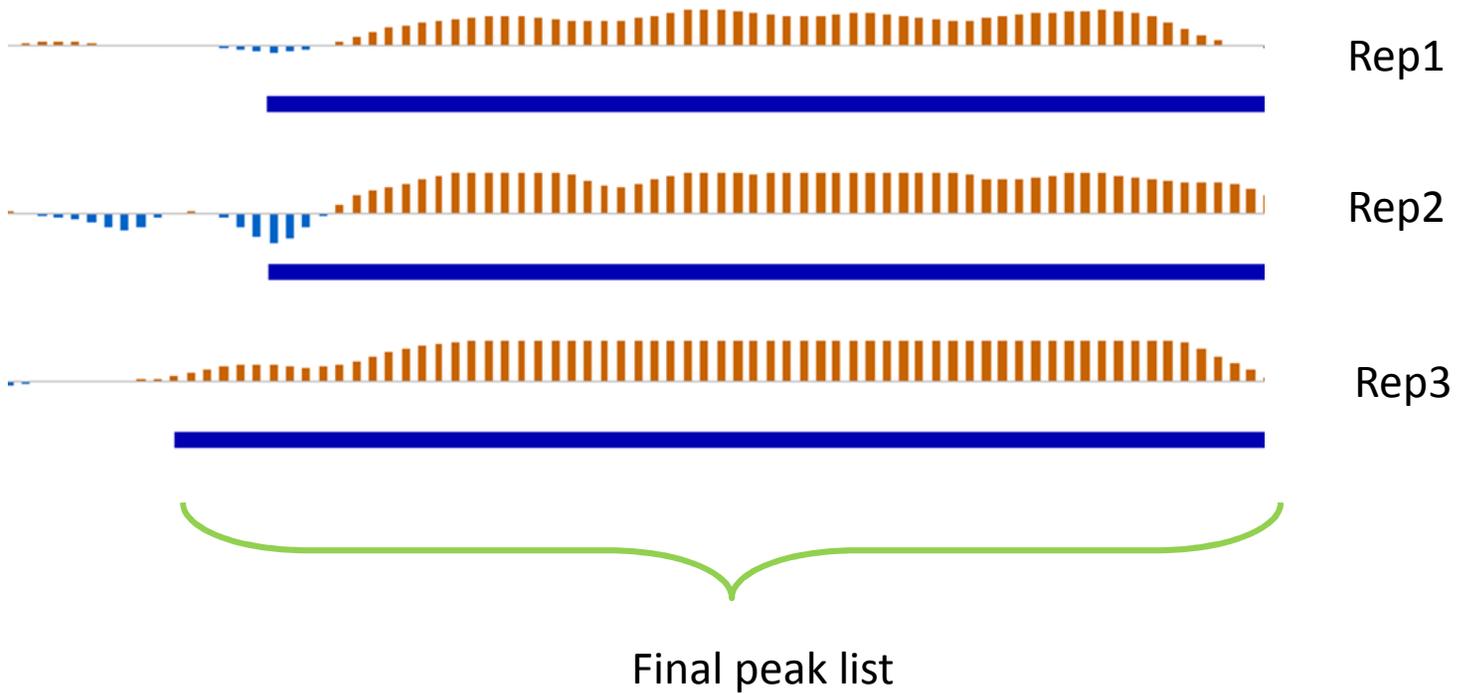
## RAD21 Replicates (high reproducibility)



## SPT20 Replicates (low reproducibility)



# Peak region merging



# Multiple replicates

$$g(N_{ij}) = \mu + x_i \beta_i + z_j u_j + \varepsilon_{ij}$$

$N_{ij}$  : observed reads count for  $i^{\text{th}}$  sample and  $j^{\text{th}}$  biological replicate

$\beta_i$  :  $i^{\text{th}}$  sample effect (fixed)

$u_j$  : random effect due to  $j^{\text{th}}$  biological replicate

$\varepsilon_{ij}$  : error

Link function: log - link for Poisson family

# R scripts for replicates

```
> time=factor(c(rep(12,2),rep(2,2),rep(12,2),rep(2,2)))
> trt=factor(rep(c("K4","H3"),4,each=1))
> design<- model.matrix(~time*trt)
> design
```

	(Intercept)	time12	trtK4	time12:trtK4
1	1	1	1	1
2	1	1	0	0
3	1	0	1	0
4	1	0	0	0
5	1	1	1	1
6	1	1	0	0
7	1	0	1	0
8	1	0	0	0

```
attr(,"assign")
```

```
[1] 0 1 2 3
```

```
attr(,"contrasts")
```

```
attr(,"contrasts")$time
```

```
[1] "contr.treatment"
```

```
attr(,"contrasts")$trt
```

```
[1] "contr.treatment"
```

# Comparing pairs

Parameter	Contrast 1	Contrast 2	Contrast 3
$\beta_{Young\_ChIP}$	1	0	0.5
$\beta_{Young\_control}$	-1	0	-0.5
$\beta_{Old\_ChIP}$	0	1	-0.5
$\beta_{Old\_control}$	0	-1	0.5

Yong IP Vs control; Old IP Vs control and Yong Vs Old