Protein – RNA interactions: Analysis of iCLIP-seq data



FU Berlin Seminar RNA Bioinformatics, WS 14/15 Sabrina Krakau

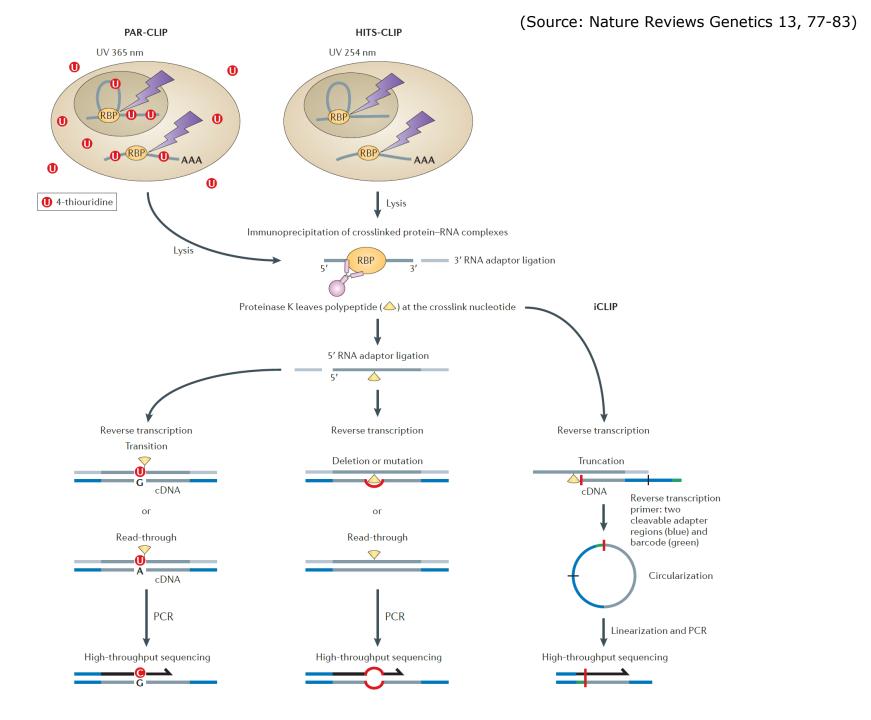
Protein – RNA interactions

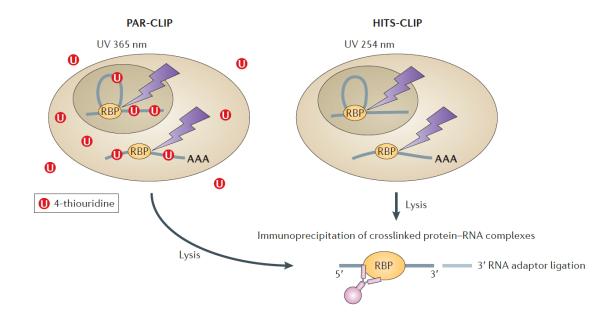


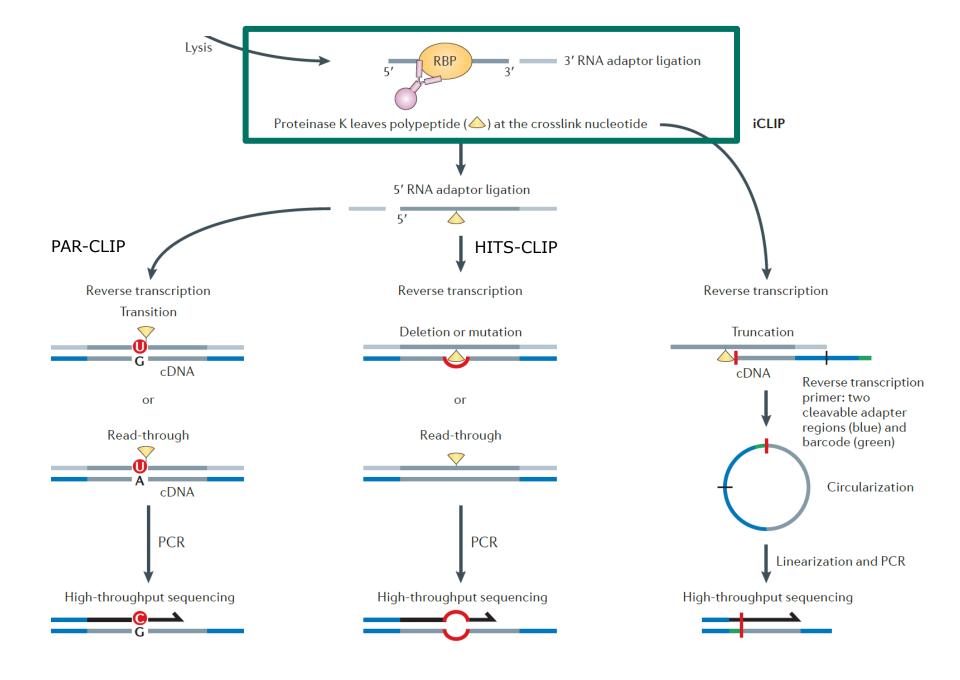
- Core of post-transcriptional regulation
- RNA binding proteins (RBPs) often bind several sites on most RNAs
 Iandcape of interactions

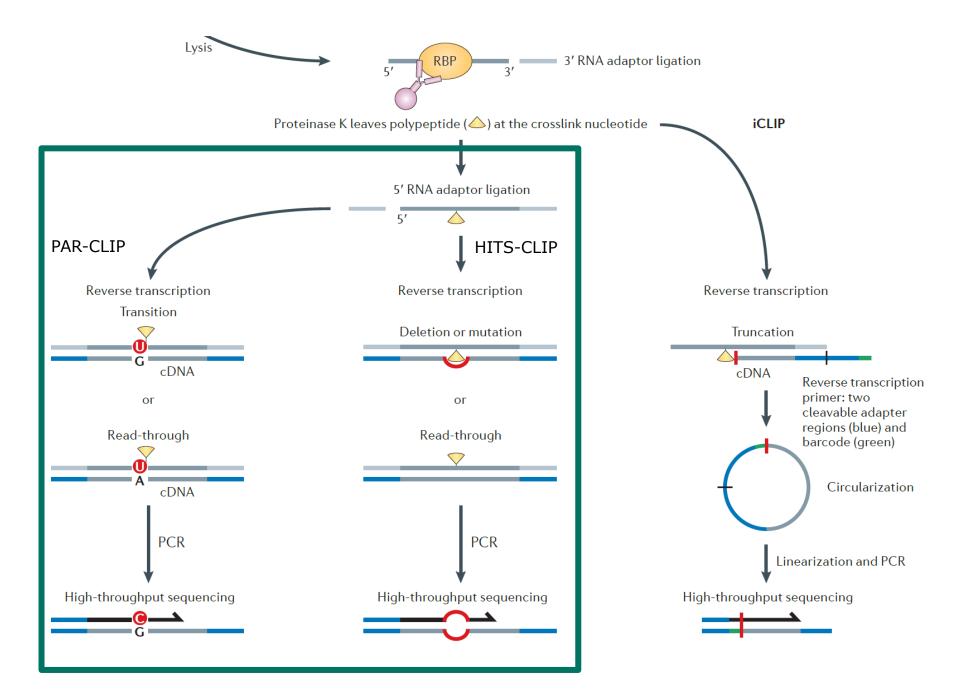
CLIP-seq (cross-linking immunoprecipitation combined with HTS)

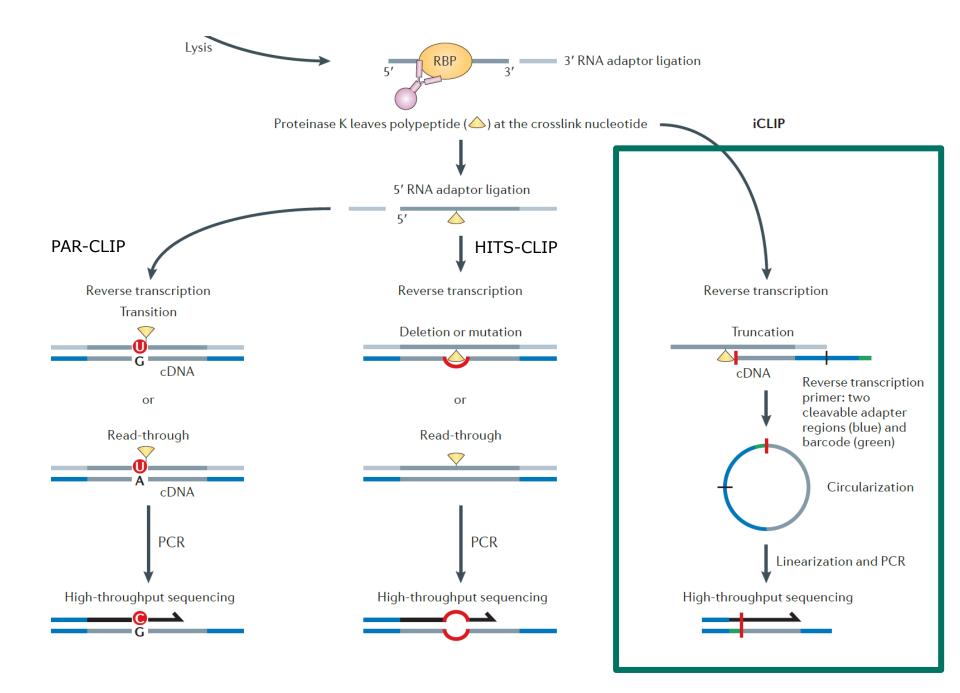
- Binding site detection with high-resolution for a given RBP
- Transcriptome-wide analysis



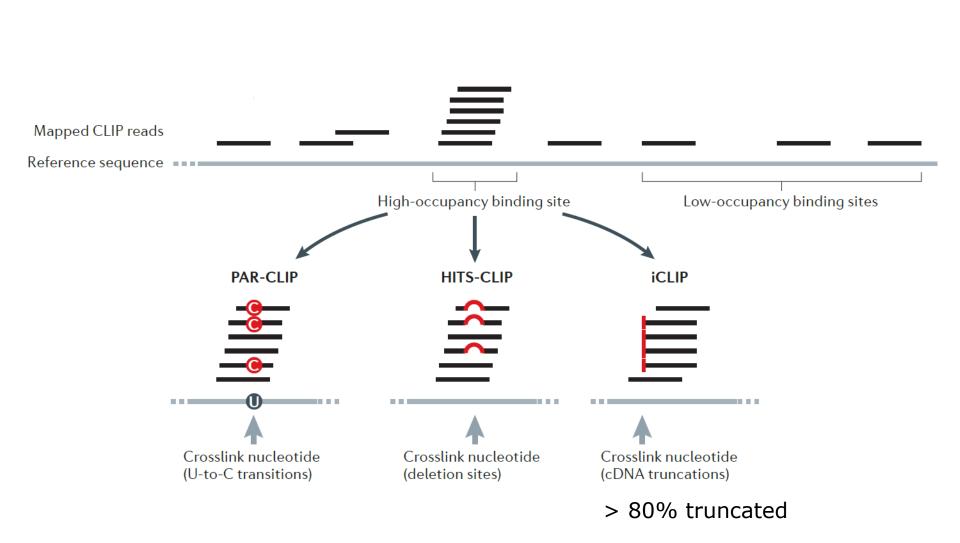








Identification of binding sites



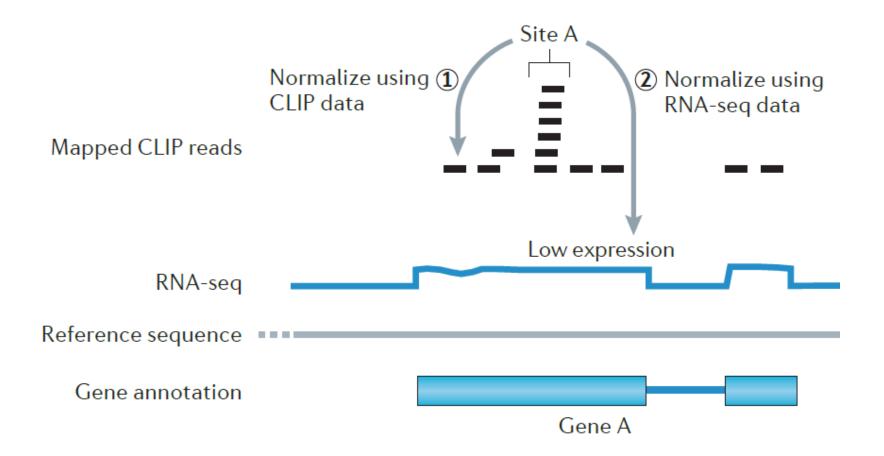
 \rightarrow diagnostic events (DEs)

Read counts \Leftrightarrow RBP binding affinity?

Normalization



Read count depends on expression level:



Which peaks are significant?



 Model underlying read count distribution to distinguish background from binding site

Take DEs into account



Piranha (2012)

- Models read count distribution of bins using the ZTNB (zero truncated negative binomial) distribution
 - Given (untruncated) mean read count μ
 - \rightarrow find dispersion parameter maximizing the ZTNB log-likelihood function
- External data as covariates X (e.g. transcript abundances, DEs)
 - \rightarrow ZTNB regression model: $\mu_i = \exp \left[\vec{\beta}^T \vec{X}_i \right]$
 - \rightarrow Find dispersion and regression parameters β that maximize the log-likelihood function





Calling peaks/enriched clusters:

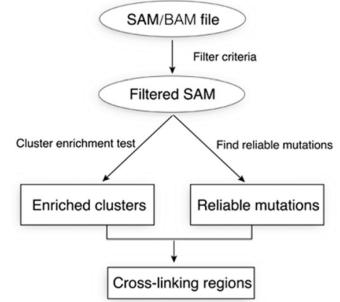
- ZTNB regression model for read counts of cluster
 - \rightarrow p-value \rightarrow FDR

Detecting cross-linking sites:

 Number of DEs is modeled with binomial distribution (no. of mapped reads, DEs and global success rate)

 \rightarrow p-value \rightarrow FDR

- → Combine p-values for final calling (using Fisher's method)
- No normalization for transcript abundances!





dCLIP (2014)

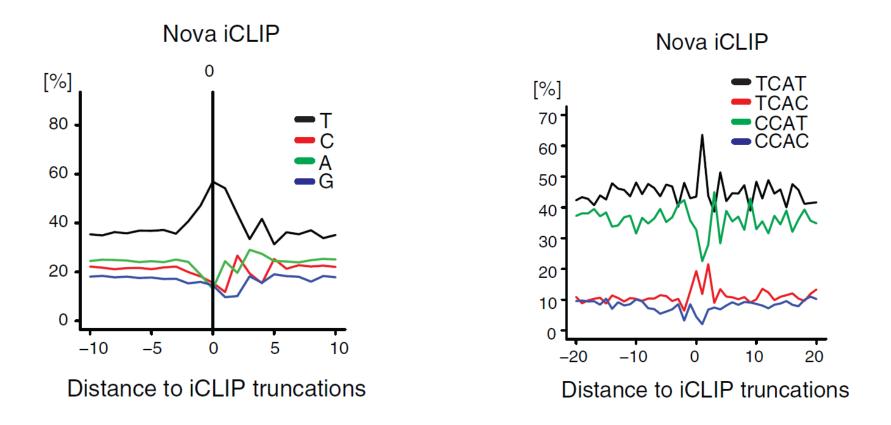
- Comparative CLIP-seq analysis
- Normalization: MA-plot (assuming a large number of common binding sites with similar binding strengths)
- Detection of RBP sites using HHM:
 - Differential binding vs. non-differential binding site

Is it that simple?

Sequence bias



UV-C induced cross-linking preferentially occurs at Us (Sugimoto et al., 2012):



 \rightarrow Bias can be avoided by analysis of motifs enriched in the vicinity

(Source: Sugimoto et al. Genome Biology 2012)

Background binding



- 1) Binding to proteins != RBP of interest
- 2) False cross-linking events

Friedersdorf et al., 2014:

- 8 45% of reads from published PAR-CLIP datasets overlap with background sites from FLAG-GFP PAR-CLIP
- Background reads are mostly derived from direct protein-RNA interactions → DEs
- → Use control CLIP with unspecific protein (or publicly available results in GEO for PAR-CLIP) for correction

GC bias



Read counts depend on GC content:

 GC rich and poor sequences are underrepresented (due to different melting temperatures in PCR)

 \rightarrow GC normalization

Motifs:

Refining binding sites and characterization

Motif discovery



RBP binding sites:

- Shorter than TF binding sites
- Characteristic secondary structures (not trivially determined by sequence)!
- Low sequence specificity in some RBPs

MEMEris: uses RNA secondary structure to guide motif search towards single-stranded regions

RNAcontext: learning RBP-specific sequence and structural preferences

RNAmotifs: identifies multivalent regulatory motifs (clusters of short and degenerate sequences)

GraphProt: learning sequence and structural preferences

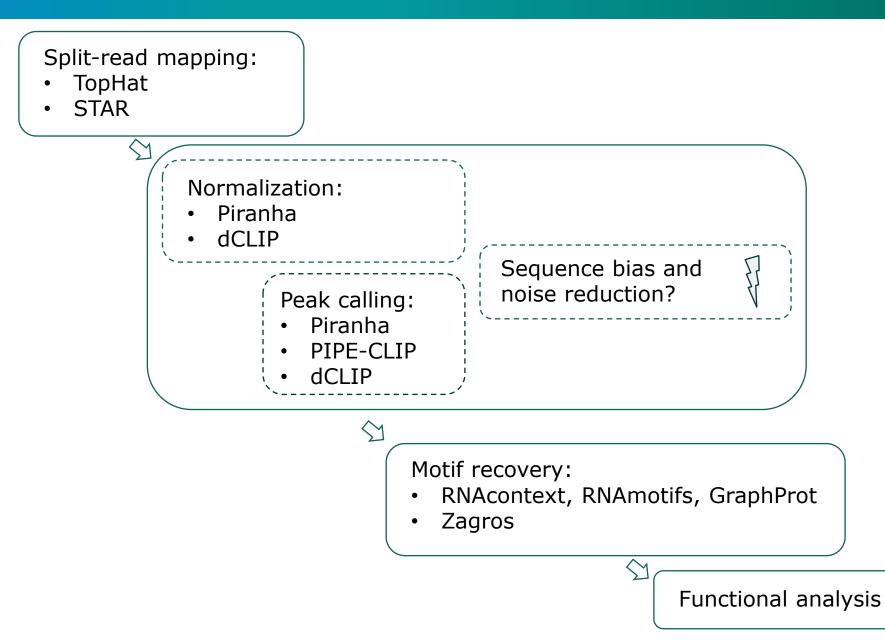
Simultaneous binding site location and motif discovery



- Zagros (Bahrami-Samani et al., 2014)
- Simultaneous motif characterization and binding site localization
- EM algorithm:
 - estimate parameters motif model M and background model f
 - Taking sequence, structure and DEs into account
 - Recompute motif occurrence indicators at each iteration
 → binding sites
- Improved motif discovery compared to methods taking only sequence into account

Conclusion





Conclusion



Open problems

- Accurate quantitative analysis remains challenging
- Need for computational methods taking sequence bias, background noise into account

Future

- Combinatorial interactions of proteins on RNAs?
- Interactions with DNA?
- How does RNA editing or epigenetic modifications influence these interactions or vice versa?

References



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