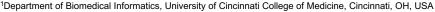
# Context-specific splicing regulatory network inference from large-scale alternative splicing data

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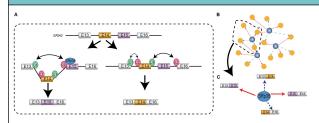


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# Cincinnati Children's\* changing the outcome together

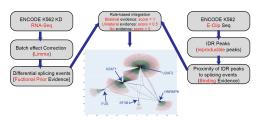


### **Splicing Regulatory Network (SRN)**



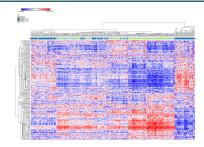
- Alternative Splicing (AS) alters the composition of encoded protein and is tightly regulated by Splicing Factor (SF) via splicing site selection.
- 2. Figure (A) depicts binding of SRp48 lead to different resultant isoforms
- Splicing Regulatory Network (SRN), shown in <u>Figure (B.C.)</u> characterize the regulatory strength between splicing Factor and regulated alternative splicing events.

### Building prior network using K562 cell



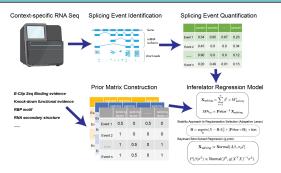
- Splicing events were quantified from ENCODE K562 knock-down RNA-Seq data, followed by rigorous batch effect correction. Splicing that are present in the differentially spliced events in certain SF knockdown experiments were deemed functional evidence.
- E-clip Seq reproducible peaks that fall into the proximal region of splicing events suggest binding evidence for potential regulation
  - We integrate the evidence from both sides via a rule-based schema, part of the prior network (U2AF1, U2AF2, FUS, SF3B1, HNRNPK) are shown above

### Inferred SF activity recapitulate the splicing architectures in AML



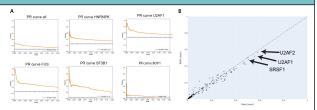
- . U2AF1-CV and SRSF1-CV are two clinically relevant cohorts that we previously identified, corresponding to poor and good prognosis
- Splice-Inferelator predicted AML specific SRN, and then inferred the splicing factor activity in each sample.

## Splice-Inferelator: in-silico construction of SRN leveraging RNA-Seq and prior knowledge



- Splice-Inferelator is a multivariate regression model to define partial dependency between splicing event and splicing factor.
- Splice-Inferelator utilizes either adaptive Lasso or G-prior to incorporate prior information into regression objective function.
- Splice\_infrelator derive splicing factor activity through context-specific transcriptomic data and prior known interactions

## Predicting and validating Splicing Regulatory Network in HepG2 cell line



- . We apply splice-Inferelator algorithm on HepG2 cell line using the high-confident prior network derived from K562 cell line. We chose HepG2 for validation because the existence of gold standard deriving from E-clip and large-scale knock-down data.
- Figure (A) shows the Area Under the Precision-Recall (AUPR) curve for all splicing factors (all) and five representative splicing factors (HNRNPK, U2AF1, FUS, SF3B1, BOP1).
- Benchmark Two regression algorithms (G-prior and adaptive lasso) shows similar results while StARS-lasso performs marginally better for bestperforming splicing factors (U2AF1, U2AF2, SRSF1)

### Conclusion

- Splice-Inferelator is capable of building contextspecific splicing regulatory network from transcriptomic data alone.
- Splice-Inferelator allows flexible incorporation of various sources of prior interactions
- Splice-Inferelator yields high-confident prediction measured by AUPR
- Splice-Inferelator can infer sample level splicing factor activity to nominate disease associated SF

### **Download**

- Code: <a href="https://github.com/frankligy/splice-inferelator">https://github.com/frankligy/splice-inferelator</a>
- Manuscript in preparation

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