

Regulus infers signed, context-dependent and process-based regulatory circuits between few cell types

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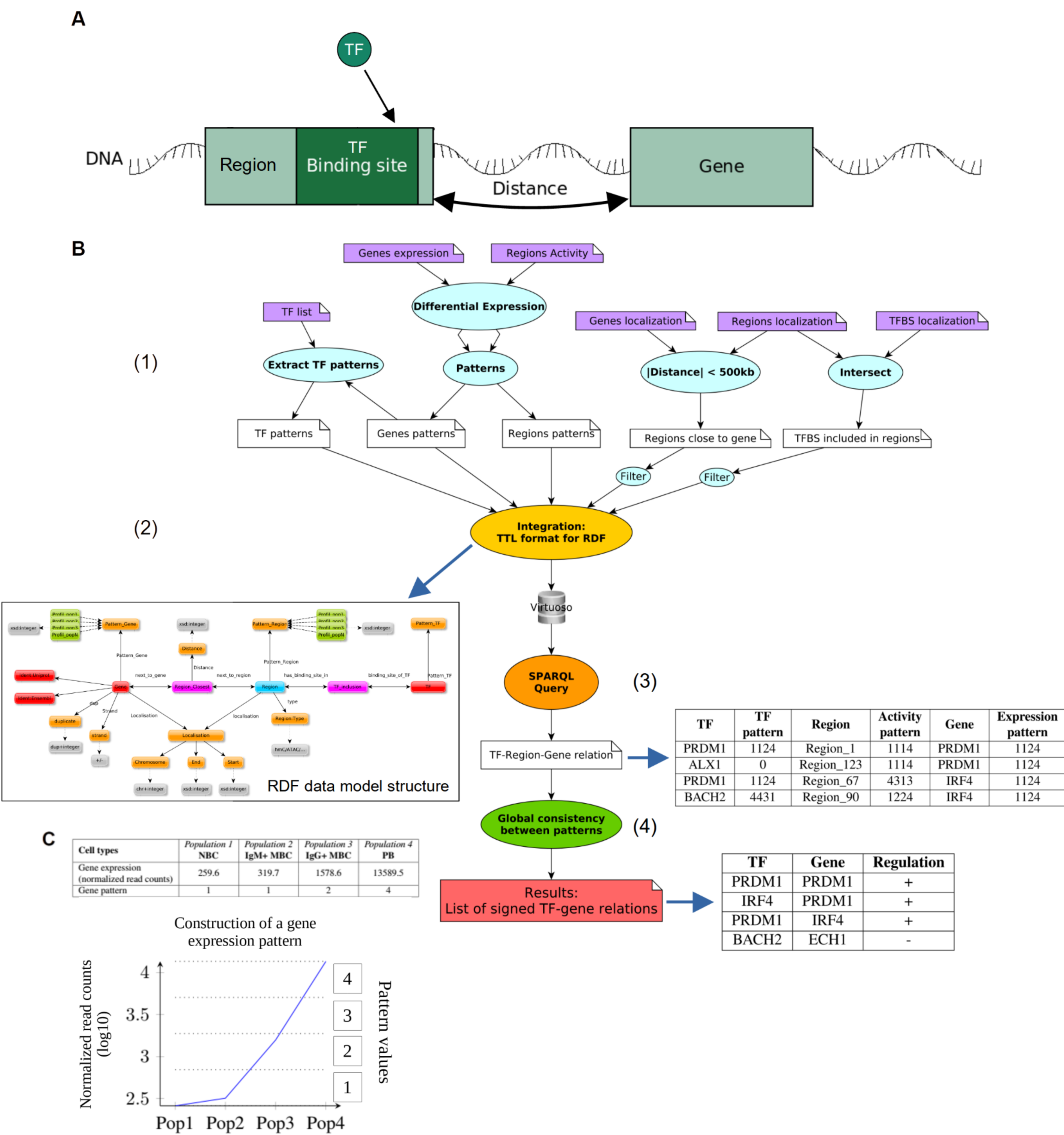
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Motivation:

- Understanding the dynamic system & the keys regulators of cell differentiation between closely related cell types, which are finely tuned by regulatory mechanisms.
- Most existing methods for inferring regulation are not applicable to common experimental or clinical settings, where the number of samples is limited.
- We created a context specific regulatory circuits inference tool, based on our previous work with Semantic Web technologies (Louarn et al., 2019, 2022).

Regulus pipeline:

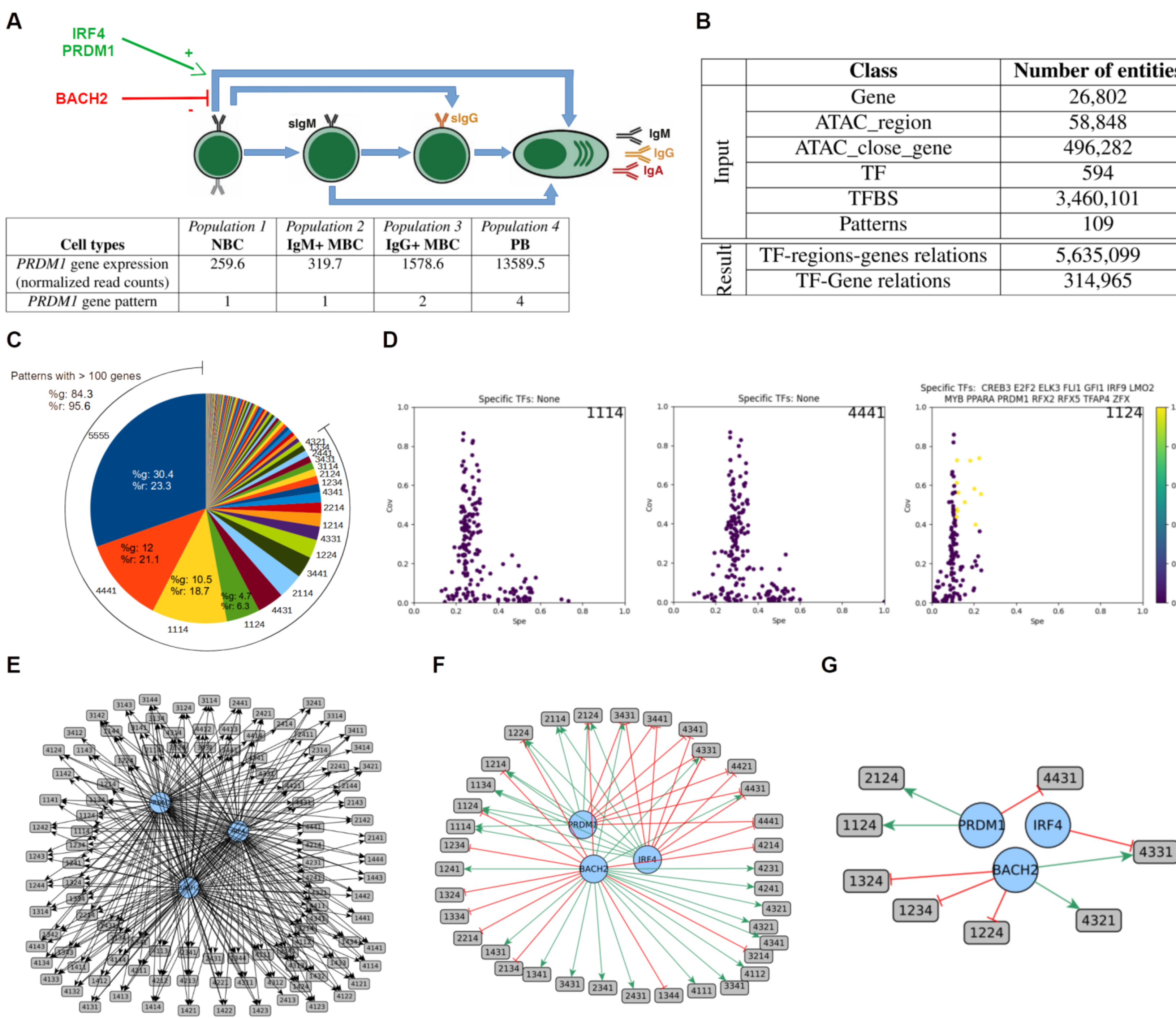


- A** - Biological relationships between the entities used to compute the regulatory circuit.
- B** - Different steps of the pipeline.
- (1) Pre-process to explicit relations between entities & activity patterns creation
- (2) RDF graph to formalize entities relations & generate a data model structure
- (3) SPARQL query to extract all TF-Region-Gene relations & respective patterns
- (4) Logical consistency constraints to get a signed & filtered network of unique TF-gene relations
- C** - Gene expression pattern construction.

Results:

- Our context-specific regulatory inference tool, *Regulus* computes TF-gene relations from gene expressions, regulatory region activities & TF binding sites data.
- Expression & activity patterns describe data dynamics.
- Data are integrated & queried to retrieve all potential TF-region relations.
- Relations filtered & signed using global consistency constraints on patterns.
- *Regulus* is well suited for scarce-sample & closely-related cell populations settings.

Application to B cells & Further filtering



- A** - Known biology behind B cell differentiation: with PRDM1, IRF4 & BACH2 known regulators.
- B** - Main in/outputs of Regulus for B-cell differentiation regulatory network.
- C** - Number of genes per expression patterns: 18 patterns contain >100 genes.
- D** - Coverage & specificity distribution for all TFs that target specific patterns. Yellow dots indicate TFs over threshold.
- E to G** - Interaction graphs of 3 known regulators & their targeted patterns
- (E) before filtering
- (F) after filtering with consistency step
- (G) after filtering with coverage & specificity. Note that relations are consistent with the known roles of BACH2, IRF4 and PRDM1 during the PB differentiation.

Applied to B cell differentiation data, Regulus identifies :

- known regulators: IRF4, PRDM1, BACH2 and PAX5
- 6 candidate new regulators (FOXJ3, KLF16, TFAP4, TGIF1, ZNF219, ZNF75A)

Conclusion

- *Regulus* provides signed, globally consistent TF-gene relations over few samples of closely-related cell types
- Applied to B cells, *Regulus* identifies both known & potential new regulators

Acknowledgment

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