

Comparative evaluation of genomic footprinting algorithms for predicting transcription factor binding sites in single-cell data

Amanda Everitt, Sean Whalen, Katherine S. Pollard

Chromatin Biology Club

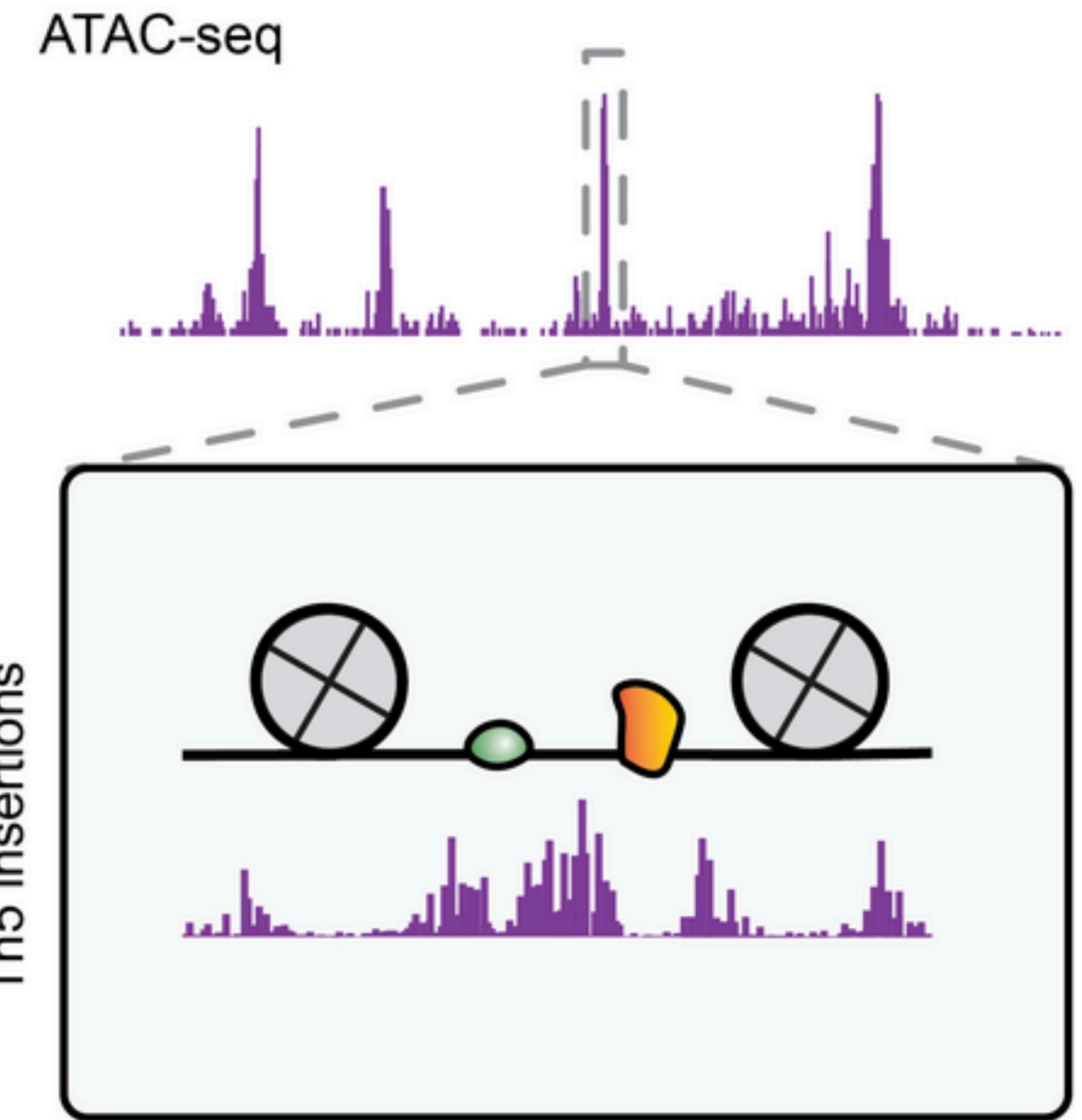
Agata Smialowska 9 Jan 2026

TF footprinting concept

- genome-wide prediction of active TF binding sites (TFBS)
- PWM-centric or signal-centric
- sequence bias correction
- supervised training (ChIP-seq data)

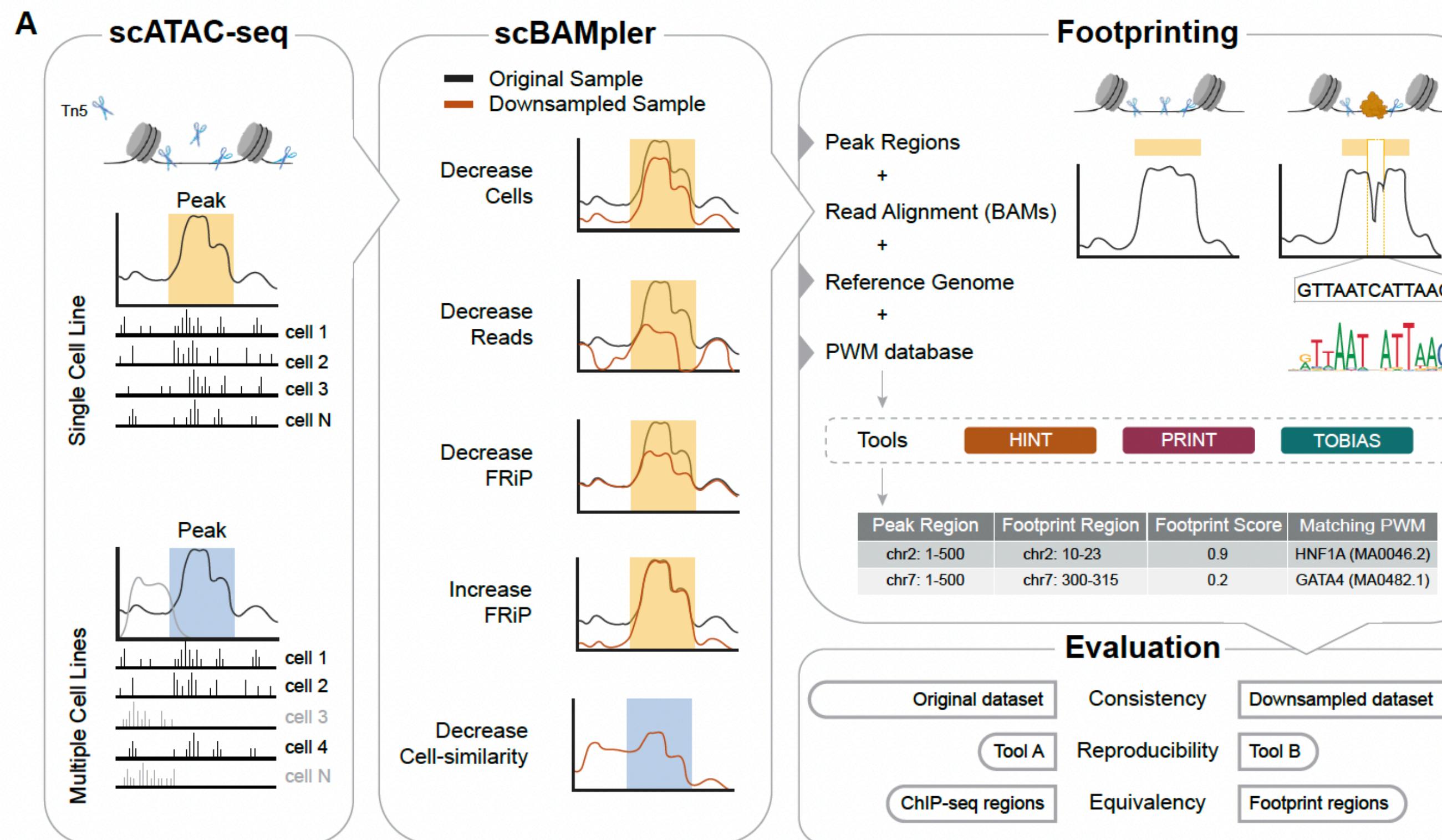
sc-ATAC-seq

- pseudobulking
- (-) data sparsity
- (+) context-dependent relationships between TFs and gene expression can be addressed
- (!) homogeneous vs. heterogeneous clusters (signal-to-noise)



scATACseq TF footprinting benchmarking pipeline: scBAMpler

- alignment level downsampling (read count, cell count, FRiP, homogeneity)
- tool evaluation
 - TOBIAS (<https://github.com/loosolab/TOBIAS/wiki/>) - motif centric; insertion sites
 - PRINT (<https://github.com/buenrostrolab/PRINT>) - both motif-centric and *de novo* modes (motif mode evaluated); insertion sites
 - HINT (<https://reg-gen.readthedocs.io/en/latest/hint/introduction.html>) - footprint region w/o motif match; NFR reads

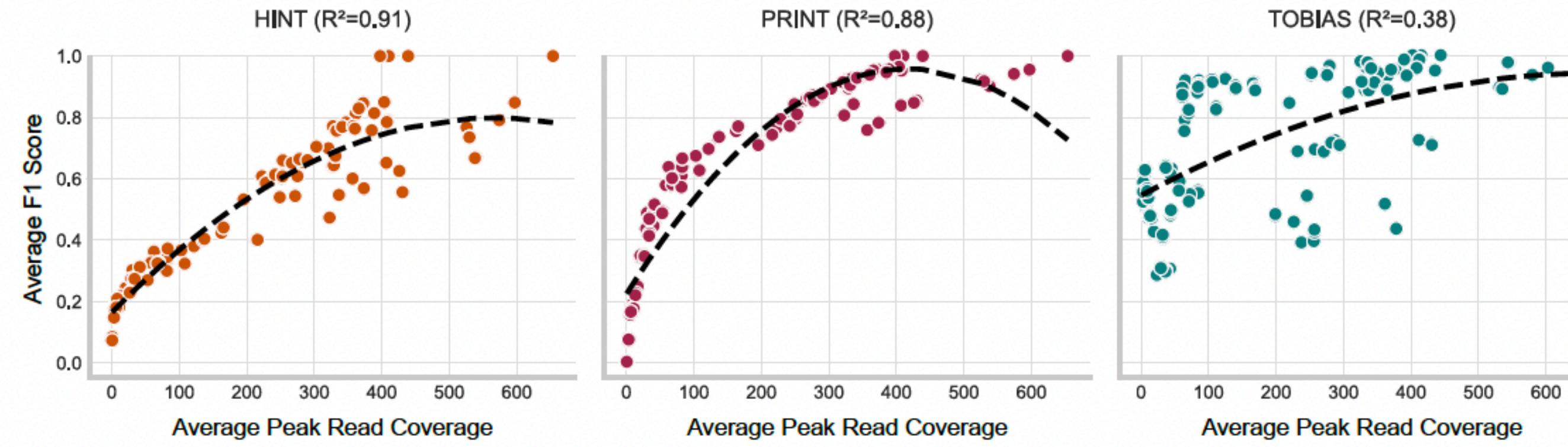


n=3 for each downsampled BAM

total ~400 BAM files

Peak read counts were the strongest factor in footprinting consistency

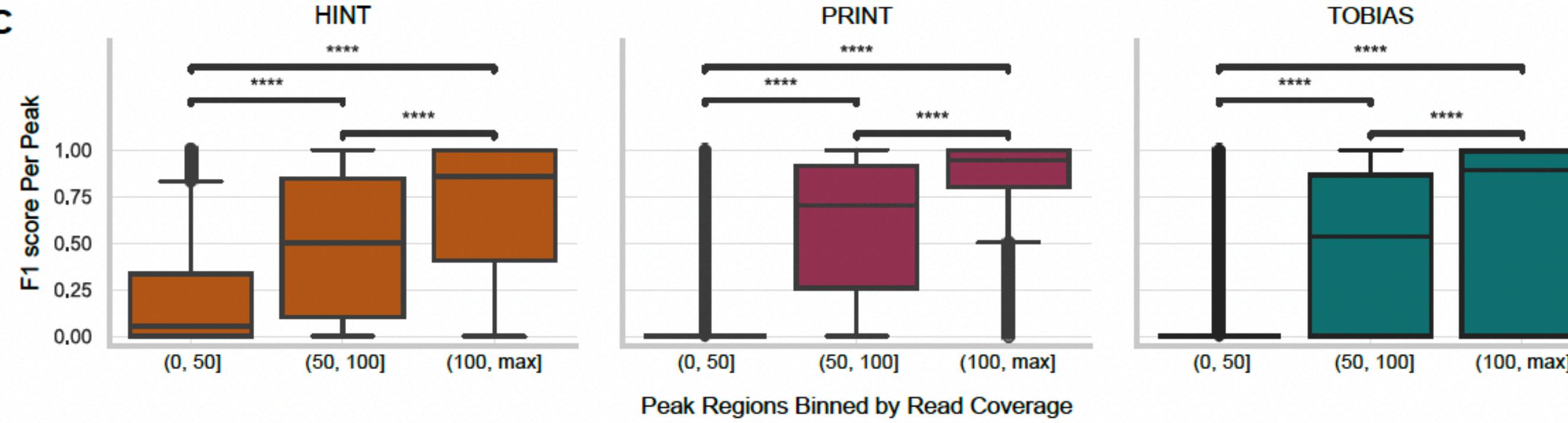
B



F1 score is the harmonic mean of precision and recall

0 (no prediction) - 1 (perfect)

C

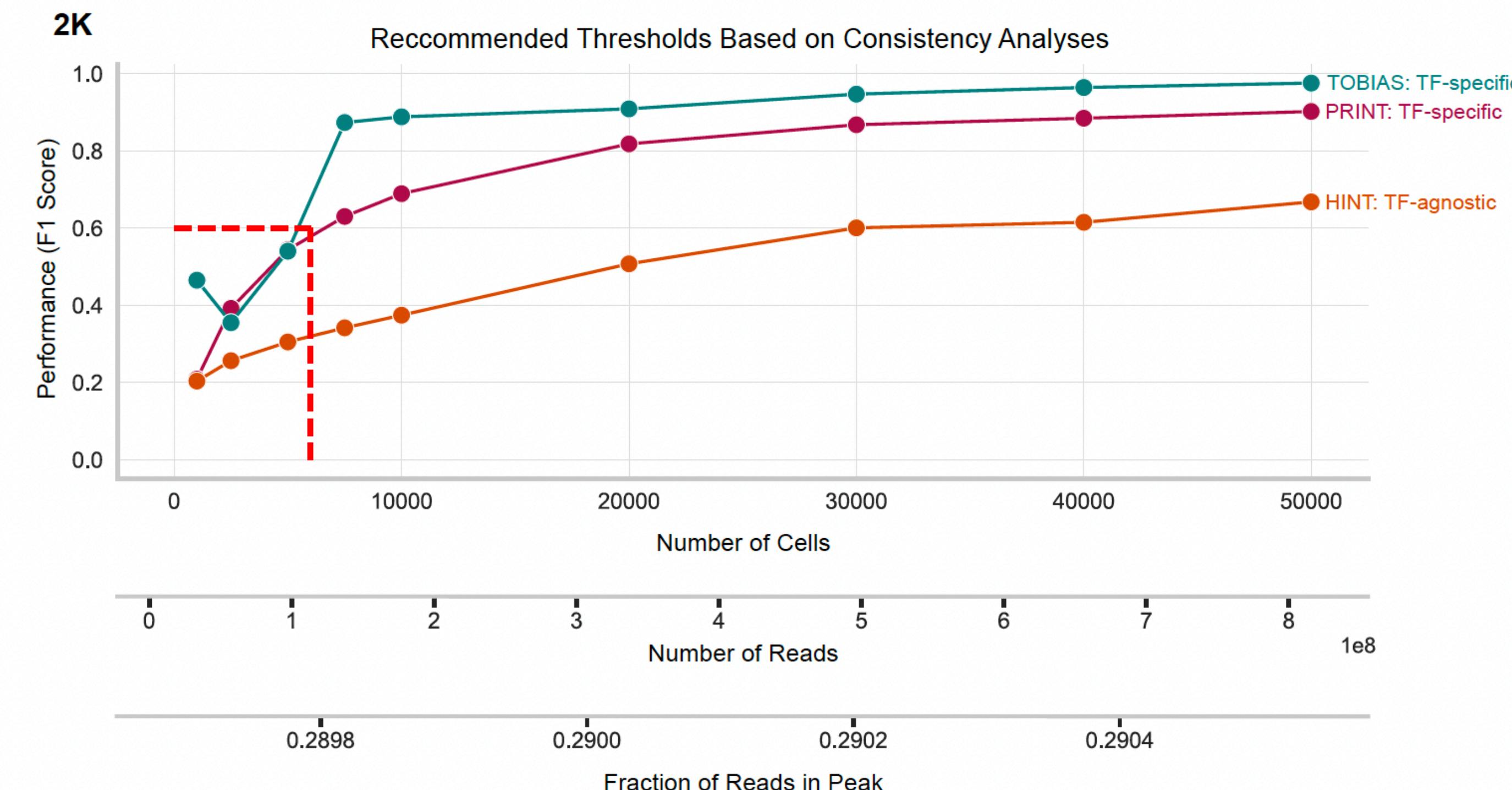


peaks with <100 reads had lowest F1 scores

retain peaks with >100 reads

Different tools, different bottlenecks

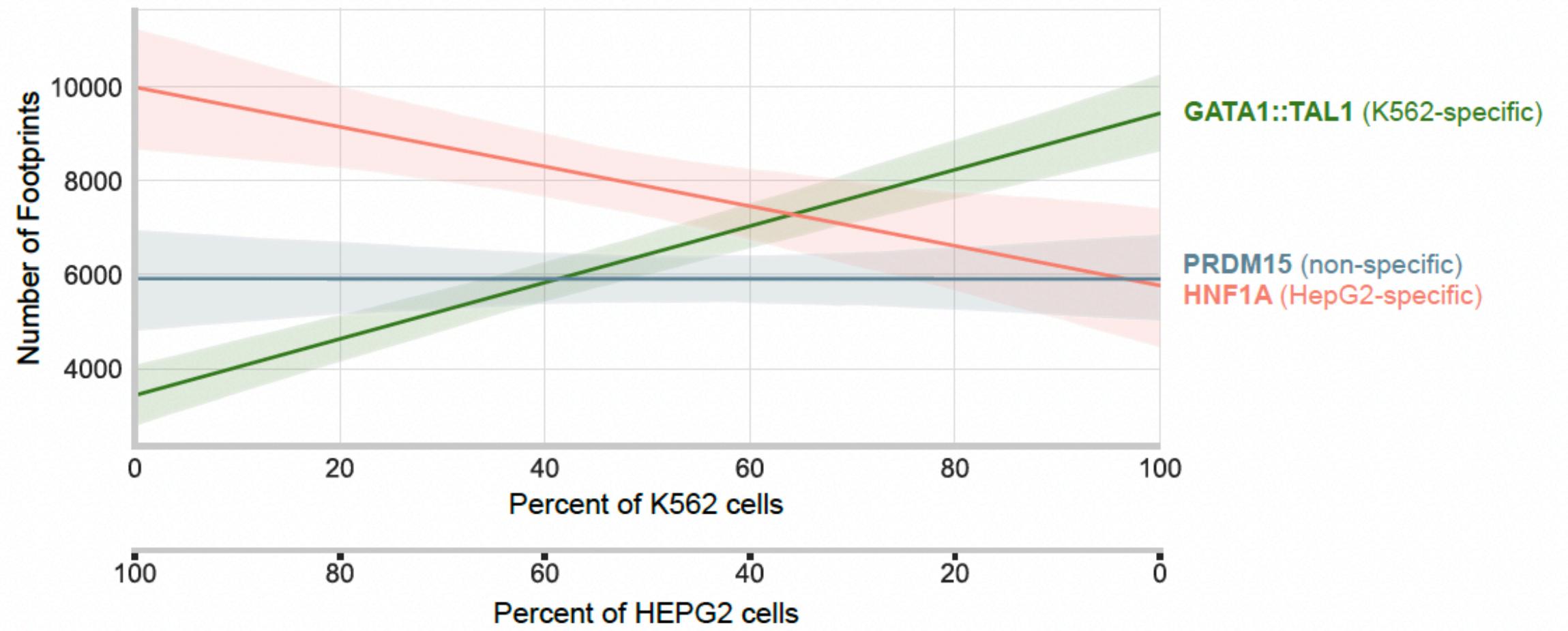
- *de novo* methods (HINT): stable footprint regions, small positional differences and hence resulting PWMs;
- TOBIAS has issues with estimating the *bound threshold* metric from low quality or sparse data; also sensitive to FRiP
- **100M PE reads / cell population** (20k fragments / cell, 0.29 FRiP pseudobulk), ca 6k cells



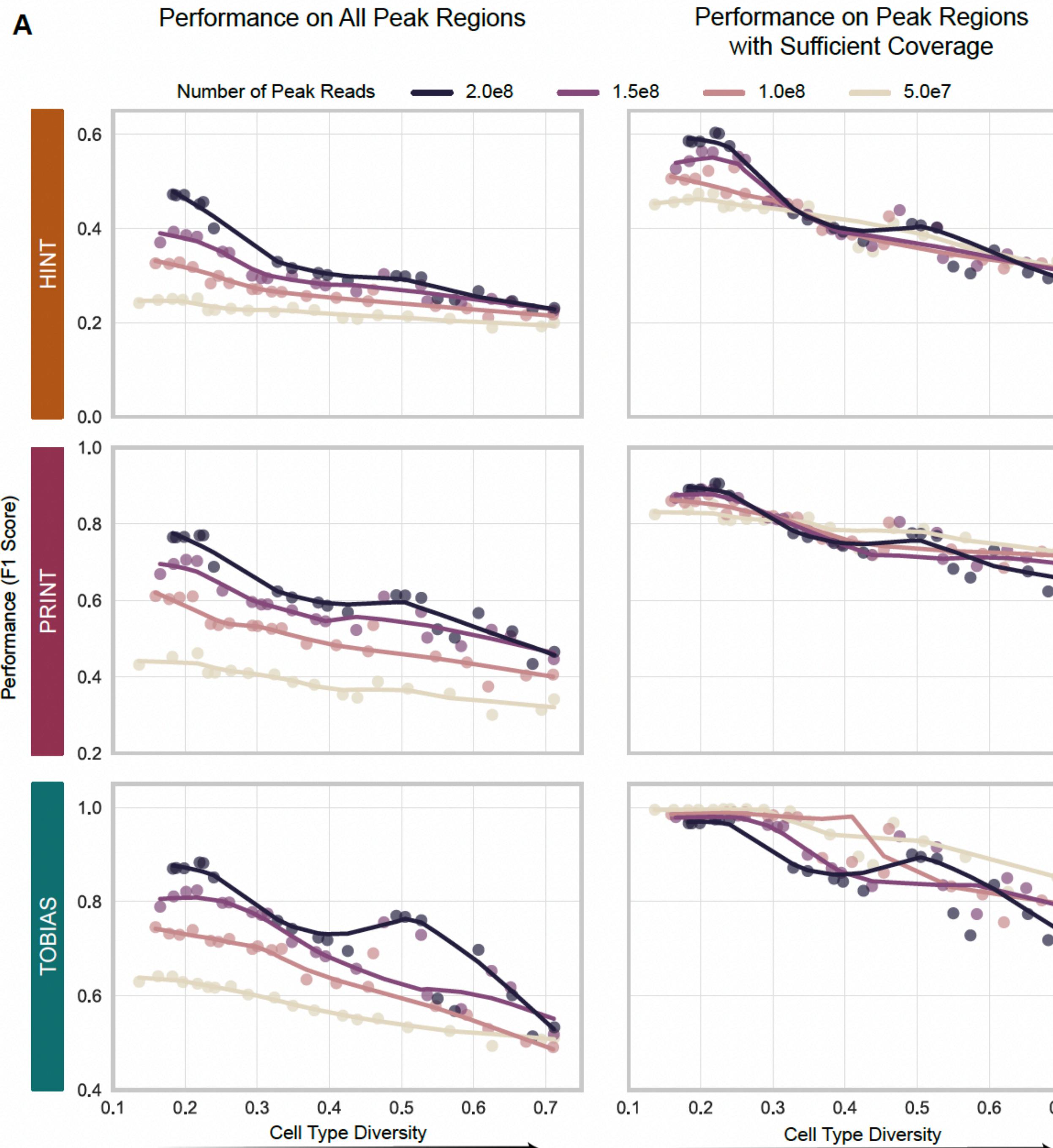
Cell diversity impacts footprint consistency

- Pseudobulk heterogeneity obscures footprinting results
- Clusters with different proportion of two cell types show different footprints
- Pseudobulking based on peak count similarity

B



A

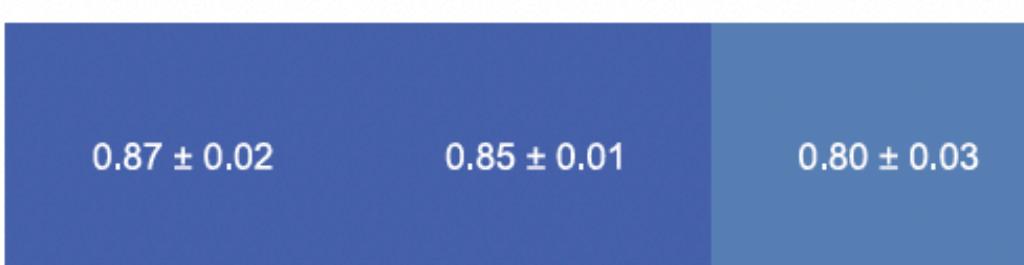


Method consistency

- low % of *footprint regions* shared between three tools; pairwise best between PRINT & TOBIAS
- motif-centric: TFBS candidate selection
- *de novo*: small differences in coordinates result in selection of a (related but different) PWM
- *PWM rankings* (“most frequent TF”) were concordant across tools and downsampling schemes
- PWM rankings across tools were more consistent than rankings based on motif scans (number of potential TFBS in peaks: monaLisa) or motif enrichment scores (HOMER)

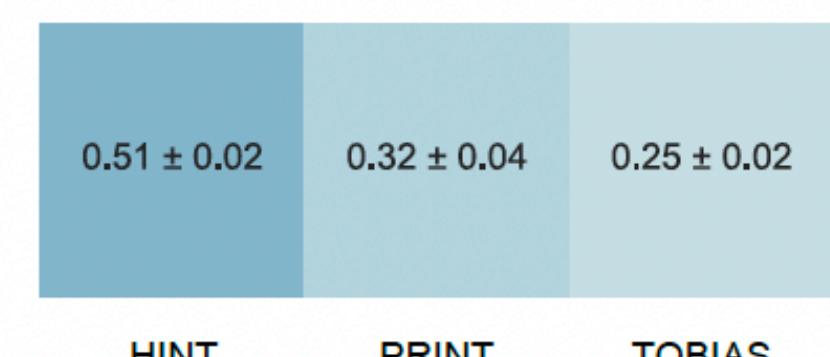
4E Similarity across Methods

(PWMs ranked by total number of footprints compared to each other)



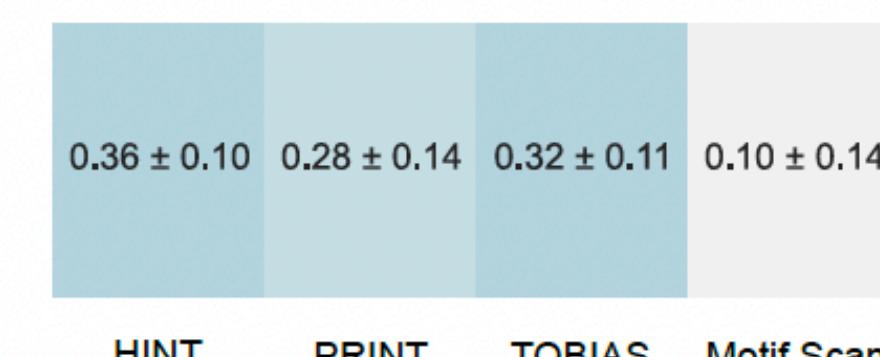
Similarity to Motif Scan

(PWMs ranked by total number of footprints compared to PWMs ranked by the number of TFBS in peak regions)

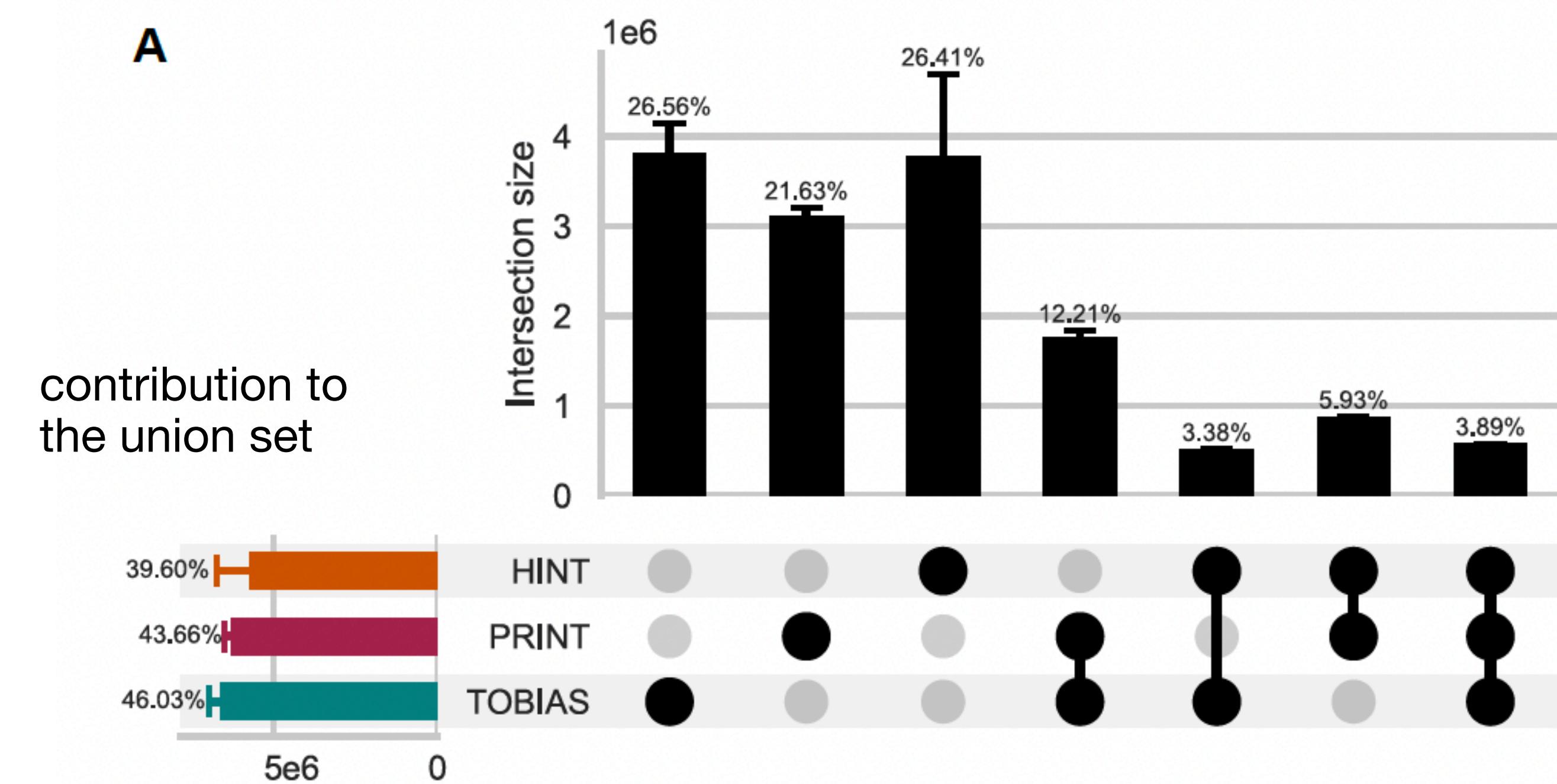


Similarity to Motif Enrichment

(PWMs ranked by total number of observations compared to PWMs ranked by significance of enrichment in peak regions)

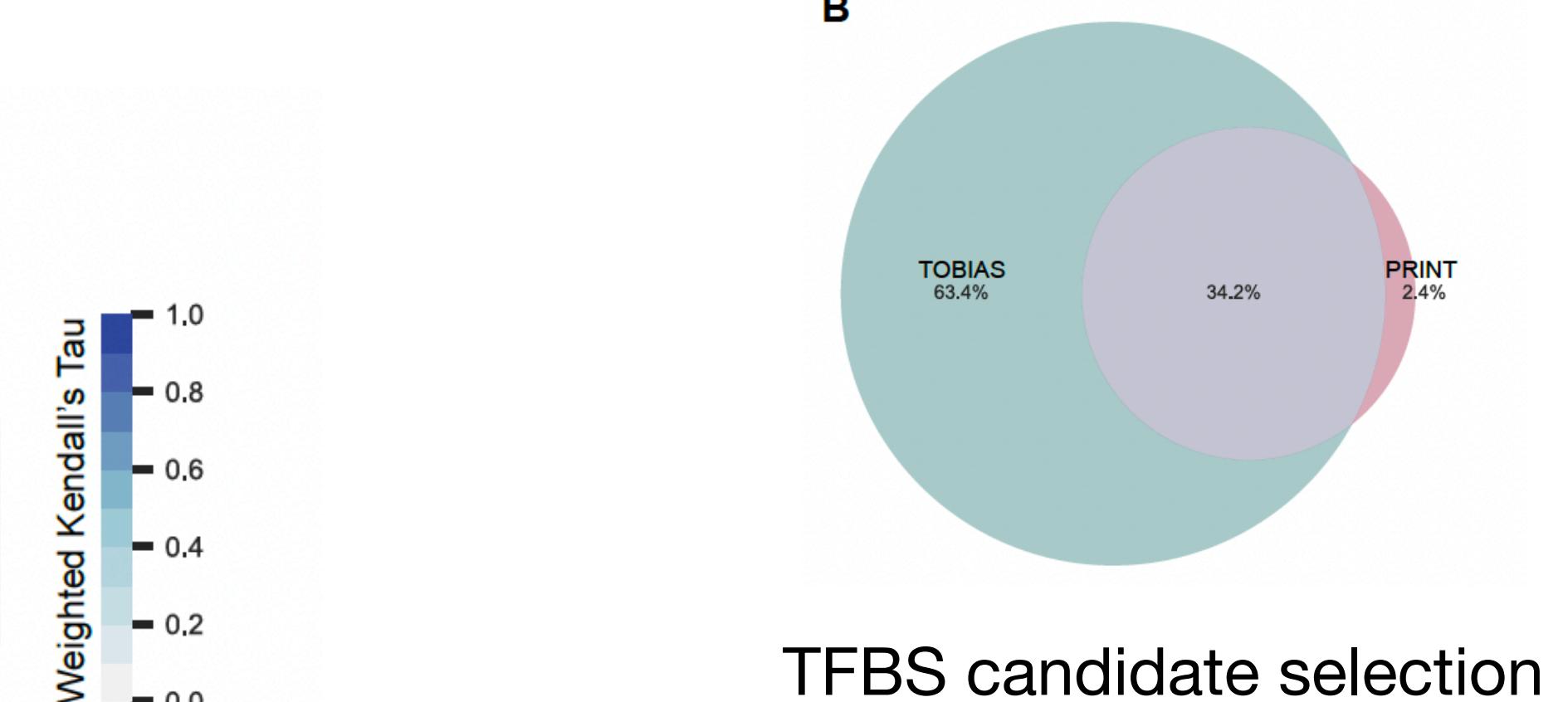


A



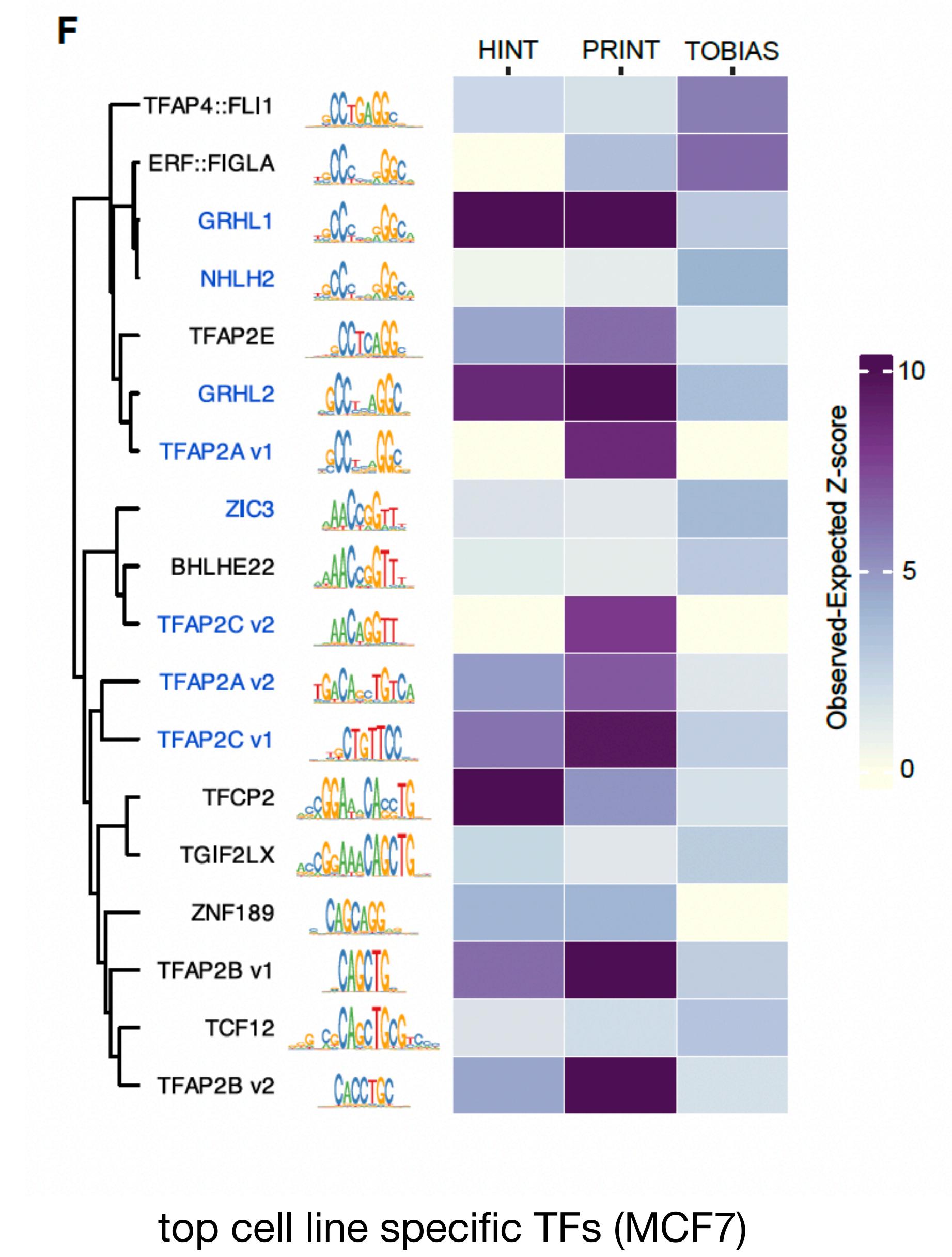
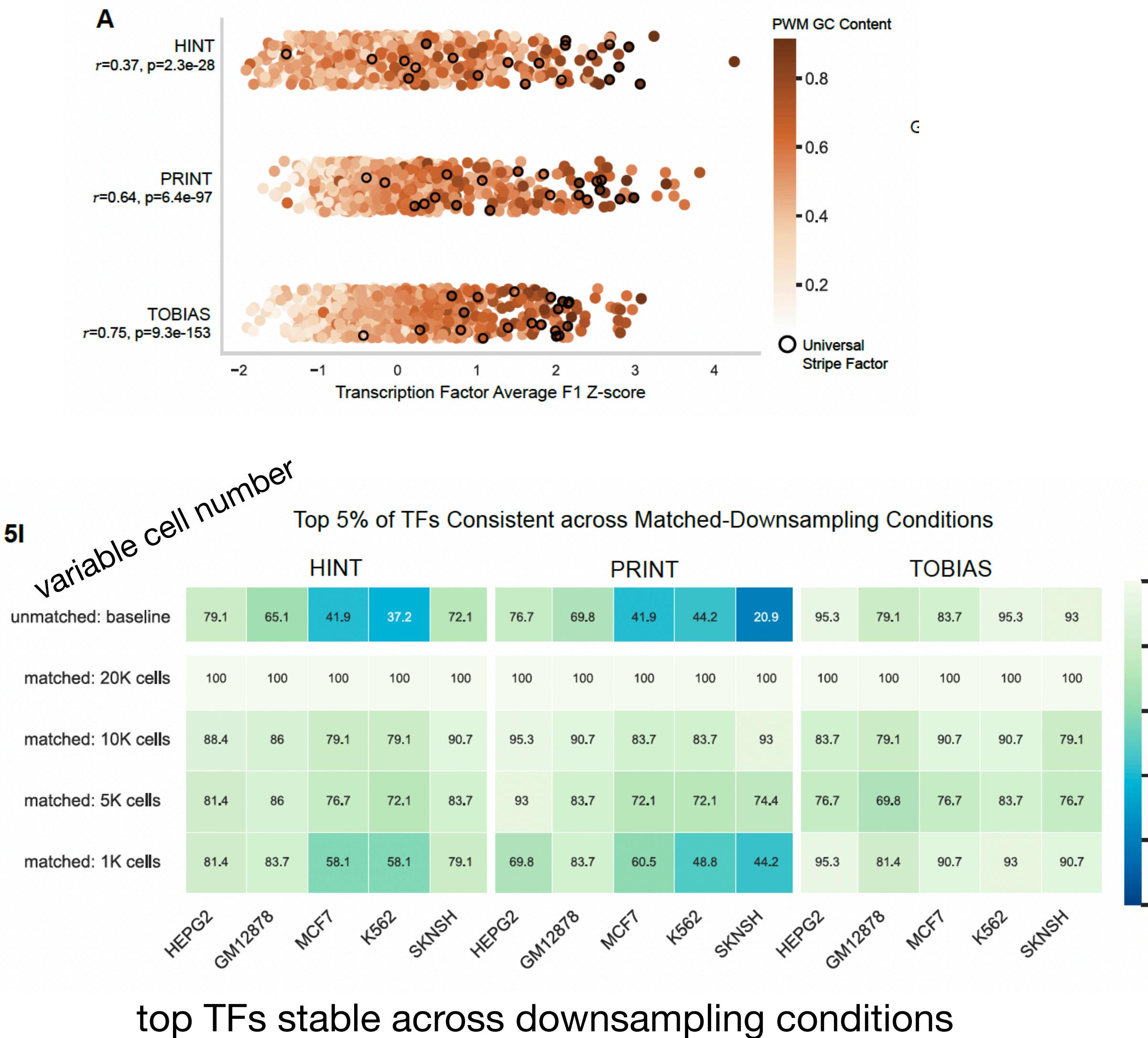
PWM presence in the same OCR

B



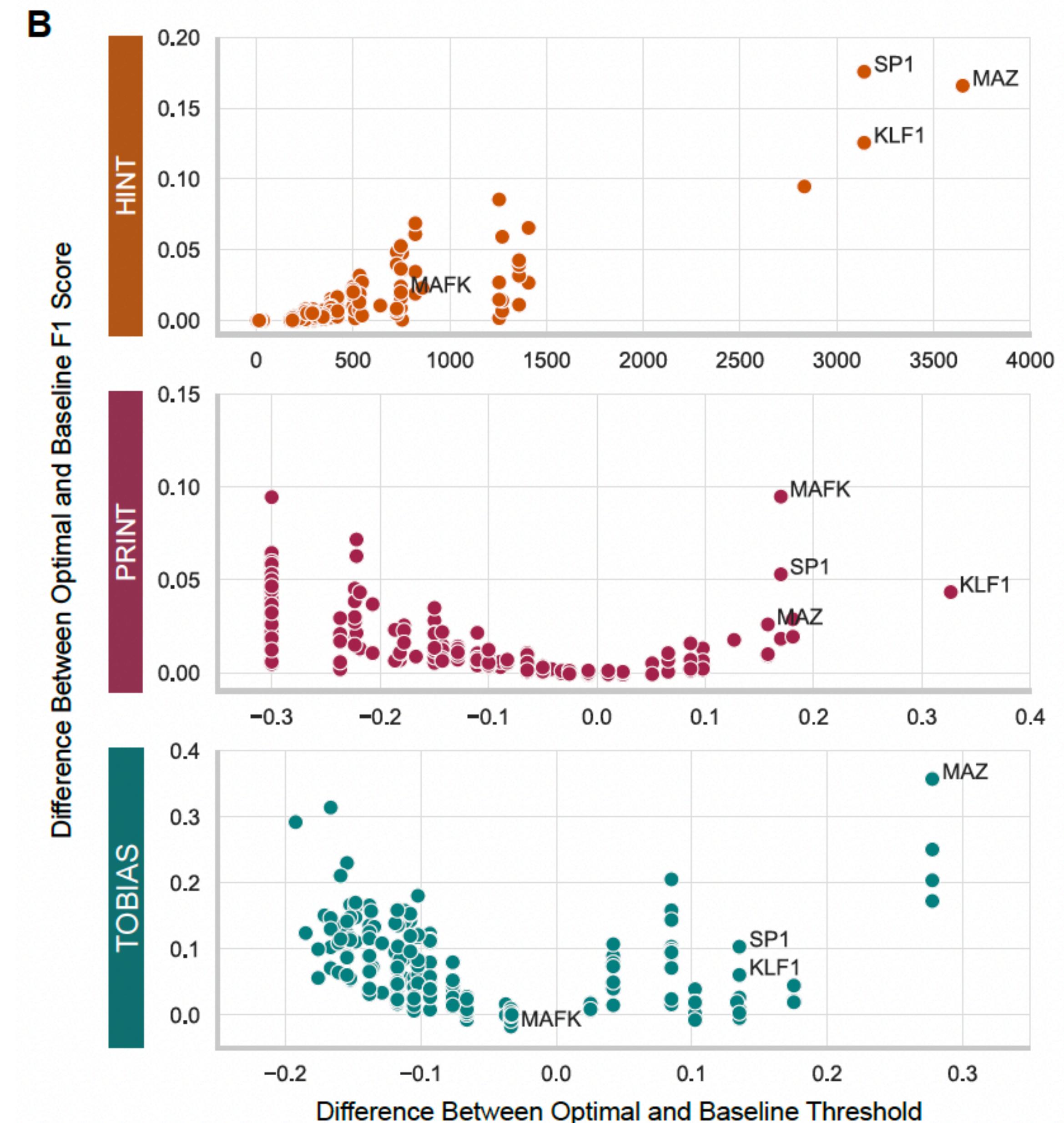
TFBS sequence vs footprinting results

- GC, information content, length, ...



Also discussed

- differential binding (TOBIAS, HINT)
- ChIP-seq overlaps
 - per TF *bound threshold* tuning
- Cell line effects



Take home message

- ATAC-seq signal quality had greater impact on footprinting performance than PWM quality or tool choice
- read depth (read per peak) dominant factor for footprinting performance (false negatives)
- performance decline was mainly due to increase in false negatives in <100 read peaks
- motif-centric tools seem to be more congruent with ChIP-seq signal (100M read pairs)
- recommendations
 - 100M read pairs / cell population (6k cells)
 - global different TF occupancy across cell populations: matching sample quality; downsampling (by cell count) all cell populations to match the smallest group (or split larger ones to pseudoreplicates)
 - merging similar clusters can be used to boost depth
 - focus on peaks >100 reads in both populations (skip cell type specific peaks): data quality matching less important (low cell number datasets - evaluate well covered regions, reduce high false negative low signal peaks)
 - motif scan more appropriate if: <1k cells, FRIP < 0.1, low peak-read counts
 - peak calling: conservative filtering (MACS3 peak summit filter 20)

Thanks for listening

