

Genomic Computing Politecnico di Milano







Discovering similar (epi)genomics feature patterns in multiple genome browser tracks

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Discovering similar patterns in genomics tracks Background



- **Next Generation Sequencing** (NGS) is opening many interesting practical and theoretical computational problems
- **Genome browsers** (UCSC Genome Browser, Integrated Genome Browser (IGB)) allow:
 - Visual inspection and identification of interesting patterns on multiple genome browser tracks
- **Pattern**: a set of (epi)genomic regions / peaks at given distances from each other in different tracks



 e.g.: gene regulatory DNA areas, which include heterogeneous (epi)genomic features (different histone modifications, TFBSs, , ...)

Discovering similar patterns in genomics tracks Pattern-based queries from genome browser



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Discovering similar patterns in genomics tracks *Motivation*



Once such patterns are visually identified in a genome section:

- Search for pattern occurrences in whole genome:
 - Complex computational task
 - Currently not supported
- Their discovery in whole genome very important for:
 - Biological interpretation of NGS experimental results
 - Comprehension of biomolecular phenomena



Discovering similar patterns in genomics tracks Reached goal



- We defined an optimized pattern-search algorithm to find genomic region sets that are similar to a given pattern
 - Efficiently
 - In large (epi)genomic / transcriptomic data sets
- We implemented the algorithm within an IGB plugin, named SimSearch, which allows intuitive user interaction in both:
 - Visual selection of an interesting pattern on loaded IGB tracks
 - Visualization of occurrences of similar patterns identified in the whole genome

Discovering similar patterns in genomics tracks Method



- Best-Matching Problem (BMP) is suspected to be NP-hard
- For the considered application:
 - Aligned genomic data -> strictly increasing region sequences
 - M << N (M and N: number of elements in a pattern and in the target tracks to be compared, respectively)
- Our proposal: a Root-element approach (R-BMP) and Windowed
 Dynamic Programming algorithm (WDP-BMP) to lower complexity:
 - Best matching for each element of the pattern in the target track through a **binary search**
 - Complexity only of O(N*log(N))
 - Applicable also to (very) large data problems



- **Pattern** = single track
- **Regions** = points, identified by their linear genomic coordinate
 - Patterns and tracks = sequences of integers (region coordinates)
- Only <u>relative</u> distances between regions/points are relevant



- A **matching** is a strictly increasing function *f* that assigns to each pattern (query) element a (target) track element
 - Preserving element order



- Pattern matching typically solved by a cost based approach, where lower cost implies higher similarity
- **Root-element approach** (R-BMP): The cost C of a **matching** *f* is the sum of squared distances relative to the first (root) matching pair

$$C_f(Q,T) = \sum_i (T_{f(i)} - Q_i - (T_{f(1)} - Q_1))^2$$

- Q and T: sequences of query (pattern) and target elements
- $T_{f(1)}$ Q_1 : root-distance (relative distance between root elements)
- **Goal**: find the matching with minimum cost



Example

- Pattern (query) Q = <1,7,10>
- Target track T = <3,5,9,11,13,14,18,21>
- A possible matching: {(1->3),(7->9),(10->21)}
 - Root-distance = 3-1 = 2
 - Cost = $(3-1-2)^2 + (9-7-2)^2 + (21-10-2)^2 = 81$
- Best matching {(1->5),(7->11),(10->14)}
 - Root-distance = 5-1 = 4
 - $\text{ Cost} = (5-1-4)^2 + (11-7-4)^2 + (14-10-4)^2 = 0$



Windowed Dynamic Programming algorithm (WDP-BMP)

- Main result: A matching is optimal if and only if all its partial matchings have minimum cost
- For all possible root positions in T:
 - They are |T|-|Q|+1
- For all regions Q_i in Q:
 - Find (with binary search) the best match for Q_i in T
 - Possibly, widen the "window" to avoid conflicts (regions in T that are best match for multiple regions in Q)
- For all possible matches:
 - Compute the cost using the partial cost obtained till then
 - Abandon the current solution if partial cost \geq best cost till then
- Overall complexity: O(|T||Q|(log|T|+|Q|)
 - But **O(|T|log|T|)** if |Q| << |T| (usually $|Q| = 1 \div 10$, $|T| \sim 10^5 \div 10^6$)

Discovering similar patterns in genomics tracks Method – Extended model



- Multi-track patterns
 - Same approach is repeated for each pattern track
- Negative matching tracks
 - Pattern tracks with regions that should not appear in the solution
 - Removed from the solution search space (before search start)

Partial matching tracks

- Pattern tracks with regions that might be missing in the results
- A cost is considered for those regions that remain unmatched

Region features

- The cost function includes the (dis-)similarity between features (attributes) of query and target track regions
- Regions as intervals
 - Region size modeled as a region feature
- Top-K distinct matchings
 - To consider diverse results, we require that the best K results have no matched region in common

Discovering similar patterns in genomics tracks *Validation*



Finding enhancer regions

- Complex pattern
 - 2 single region positive tracks (H3K4me1 and H3K27ac)
 - 2 single region negative tracks (TSS and H3K4me3)
 - 4 single region partial tracks (CTCF, DHS, P300 and Pol2)
- Target data tracks about K562 cell line (AML) from ENCODE
- Top-100 results visually inspected by an expert
 - **100% precision** (of the resulting matchings to the pattern)
- All (1,651) results compared with "*enhancer*" chromatin state regions from ChromHMM tool (Broad Institute), data available in ENCODE
 - 85.46% precision (but ChromHMM uses more histone marks)

Discovering similar patterns in genomics tracks SimSearch: a plugin for IGB



Available at: https://arnaudceol.github.io/simsearch/



Discovering similar patterns in genomics tracks SimSearch: define the pattern to search for





(e.g. CHromHMM chromatin states from Roadmap Epigenomics Consortium, Nature 2015)



Genome% (average)

Discovering similar patterns in genomics tracks SimSearch: searching for new enhancers



Pradeepa et al. (*Nature Genetics* 2016) identified a new class of enhancer (with H3K122ac and absence of H3K27ac) in mouse



Several examples validated in mouse

- Homozygous deletion of group 2 putative enhancer 42 kb **downstream** of *Leukemia inhibitory factor* (*Lif*) gene led to **reduced expression** of *Lif*, but not of flanking *Hormad2* gene

- Deletion of one allele of putative enhancer 30 kb upstream Tbx3 led to downregulation of Tbx3

Discovering similar patterns in genomics tracks SimSearch: searching for new enhancers

Look for this new pattern in human samples:



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Discovering similar patterns in genomics tracks SimSearch: editing the pattern

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In a new track (red), see all active regions (matches) found ~



Discovering similar patterns in genomics tracks Conclusions



- Definition of a method for **finding patterns in genomic sequences** and of a dynamic programming algorithm for efficient solution
 - Experimental evaluation found it accurate and efficient
- Algorithm implemented as a plugin for Integrated Genome Browser
- Applicable to any (epi)genomic / transcriptomic region data, e.g.:
 - Histone Modifications (HM), Transcription Factor Binding Sites (TFBS), Single Nucleotide Polymorphisms (SNP), Differentially Expressed Genes (DEG), DNase I Hypersensitive Sites (DHS), Transcription Start Sites (TSS), or any other annotations
- Support for understanding biomolecular mechanisms
 - Response to treatments, pathology onset/development, ...
- Thanks to the GenData 2020 PRIN project and GeCo ERC project!



Thank you for your attention!

Any question?





