Discovering cis-regulatory motifs using genome-wide sequence, expression and protein binding microarray data

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Gene expression regulation

- Transcription is regulated mainly by transcription factors (TFs) - proteins that bind to DNA subsequences, called binding sites (BSs)
- TFBSs are located mainly in the gene's promoter the DNA sequence upstream the gene's transcription start site (TSS)
- TFs can promote or repress transcription
- Other regulators: micro-RNAs (miRNAs)



Motif discovery: The typical two-step pipeline Promoter/3'UTR

Co-regulated gene set

Gene expression microarrays



Functional group (e.g., GO term)

Location analysis

(ChIP-chip, ...)





sequences





A Motif Algorithm for Detecting Enrichment in mUltiple Species

Supports diverse motif discovery tasks:

- 1. Finding over-represented motifs in one or more given sets of genes.
- 2. Identifying motifs with **global spatial features** given **only** the genomic **sequences**.
- Simultaneous inference of motifs and their associated expression profiles given genome-wide expression datasets.

How?

- A general **pipeline architecture** for enumerating motifs.
- Different statistical scoring schemes of motifs for different motif discovery tasks.

Task I: Over-represented motifs in given target set Input: Target set (T) = co-regulated genes Background (BG) set (B) = entire genome No sequence model is assumed! Motif scoring: Hypergeometric (HG) enrichment score b, t = BG/Target genes containing a hit



I BG set should be of the same "nature" as the target set, and much larger E.g., all genes on microarray

Drawback of the HG score

- Length/GC-content distribution in the target set might significantly differ from the distribution in the BG set
 - Very common in practice due to correlation between the expression/function of genes and the length/GC-content of their promoters and 3' UTRs
 - The HG score might fail to discover the correct motif or detect many spurious motifs
- → Use the binned enrichment score
 - Slightly less sensitive than HG score...
 - ... but takes into account length/GC-content biases

Test case: Human G2+M cell-cycle genes



These motifs form a module associated with G₂+M [Elkon et al. '03, Tabach et al. '05, Linhart et al. '05]

Benchmark: Real-life metazoan datasets

- We constructed the first motif discovery benchmark that is based on a large compendium of experimental studies
- Source: Various (expression, ChIP-chip, Gene Ontology, ...)
- Data: 42 target-sets of 26 TFs and 8 miRNAs from 29 publications
- Species: human, mouse,
- fly, worm
- Average set size:
- 400 genes (=383 Kbps)



d score





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Amadeus – Global spatial analysis



Task II: Global analyses

Scores for spatial features of motif occurrences Input: Sequences (no target-set / expression data)

Motif scoring:Localization w.r.t the TSS

Strand-bias

Chromosomal preference

"	>{		}} 3	N		5	
) 6	7	8	9	1 0	≱ % 11	1 2	
13	1 4	2 (15		16	1 7	18	
8 19	20		21	22		×	

TSS

Global analysis I: Localized human + mouse motifs

Input:

- All human & mouse promoters (2 x ~20,000)
- Score: localization



Name	Motif logo	Localization		Strand bias	Chrom pref.						
		peak	<i>p</i> -val	<i>p</i> -val	<i>p</i> -val						
A. Known TFs											
SP1	GC_CC=CCC=	-60	10 ⁻¹²⁹	-	-						
NF-Y		-9060	10 ⁻¹⁴⁵	-	10 ⁻⁴ (19)						
GABP	<u>GGAAGtg_</u>	-300	10 ⁻¹¹³	-	-						
ΤΑΤΑ		-30	10 ⁻⁶¹	10 ⁻¹⁵	10 ⁻³ (6)						
NRF1		-30	10 ⁻⁴⁸	-	-						
ATF/CREB	ACGTCACAGA	-60	10 ⁻²⁷	-	-						
MYC	<mark>GTCA</mark> CTG	-6030	10 ⁻²⁷	-	-						
RFX1	c_TGGcAACG	-60	10-9	-	-						
B. Novel											
ACTACAWYTC		-9060	10 ⁻²¹	10-8	10 ⁻⁴ (19)						
CTCGCGAGAT	ст <u>с</u> с <u>с</u> дадат	-6030	10-7	-	-						
C. Other											
Splice donor site	$= GGT_{\subseteq}AG_{=}$	+30	10 ⁻²³	10-8	-						

Amadeus is available at:

"Transcription factor and microRNA motif discovery: The Amadeus platform and a compendium of metazoan target sets", C. Linhart*, Y. Halperin*, R. Shamir, Genome Research 18:7, 2008

(*equal contribution)







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PRIMA – GOAL: 'Reverse engineering' of transcriptional networks

■ Co-expression → Co-regulation → common cisregulatory promoter elements

 Identification of co-expressed genes using microarray technology (clustering)

 Computational identification of cis-regulatory elements that are over-represented in promoters of the co-expressed genes

PRIMA – General description

PRIMA identifies transcription factors (TFs) whose binding sites (BSs) are enriched in a given 'target set' of promoters with respect to a 'background set' of promoters.

Required 'databases':

Promoter sequences on a genomic scale

'Models' for binding sites recognized by TFs

Implemented in Expander.

Amadeus - Allegro

Expressibatedagane set



Allegro: expression model

Discretization of expression values

 $e_1 = Up$ (U) ≥ 1.0 e_2 =Same (**S**) (-1.0, 1.0) $e_3 = Down(D)$ ≤-1.0

Expression pattern

C_1 C_2 Cm ... -2.3 -0.8 1.5 g

Discrete expression Pattern (DEP)



- Expression data should be (partially) pre-processed, e.g.:
 - Time series \rightarrow log ratio relative to time 0
 - Several tissues/mutations/... \rightarrow standardization
 - Do NOT filter out non-responsive genes

Expression model: **CWM** = Condition Weight Matrix

- Non-parametric, log-likelihood based model, analogous to PWM for sequence motifs
- Sensitive, robust against extreme values, performs well in practice





Human cell cycle [Whitfield et al., '02]



Large dataset: ~15,000 genes, 111 conditions, promoters region: -1000...200 bps



Allegro recovers the major regulators of the human cell cycle [Elkon et al. '03; Tabach et al. '05; Linhart et al. '05].

Yeast HOG pathway [O'Rourke et al. '04]

~6,000 genes, 133 conditions



(a) Sin1 branch

(b) Sho1 bran

- Allegro can discover multiple motifs with diverse expression patterns, even if the response is in a small fraction of the conditions
- Extant two-step techniques recovered only 4 of the above motifs:
 - K-means/CLICK + Amadeus/Weeder: RRPE, PAC, MBF, STRE
 - Iclust + FIRE: RRPE, PAC, Rap1, STRE

Amadeus/Allegro - Additional features



Allegro is available at:

"Allegro: Analyzing expression and sequence in concert to discover regulatory programs", Y. Halperin*, C. Linhart*, I. Ulitsky, R. Shamir, Nucleic Acids Research, 2009

(*equal contribution)



Summary

- Developed *Amadeus* motif discovery platform:
 - Broad range of applications:
 - Target gene set
 - Spatial features (sequence only)
 - Expression analysis Allegro
 - Sensitive & efficient



- Easy to use, feature-rich, informative
- New over-representation score to handle biases in length/GC-content of sequences
- Novel expression model CWM
- Constructed a large, real-life, heterogeneous benchmark for testing motif finding tools

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