Predicting RNA binding of new RNA-binding proteins

Project in
Deep Learning in
Computational Biology

Outline

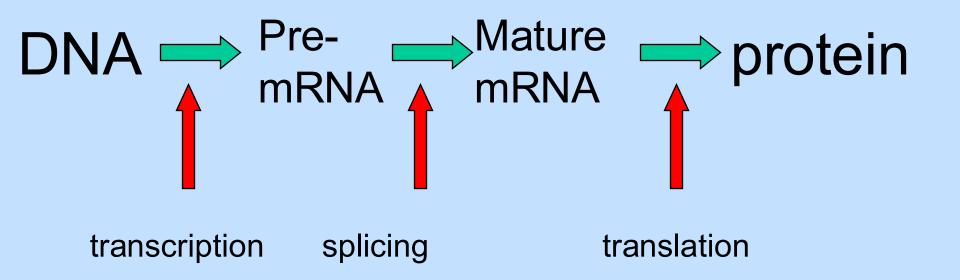
1. Some biological background again...

2. The project

1. Background

Slides with Ron Shamir and Chaim Linhart, Computational Genomics TAU

Gene: from DNA to protein



DNA

- DNA: a "string" over the alphabet of 4 bases (nucleotides): { A, C, G, T }
- Resides in chromosomes
- Complementary strands: A-T; C-G

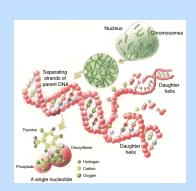
Forward/sense strand:

AACTTGÇĢ

Reverse-complement/anti-sense strand:

TTGAACGC

Directional: from 5' to 3':



RNA (Ribonucleic acid)

Bases:

- Adenine (A)
- Guanine (G)
- Cytosine (C)
- Uracil (U); replaces T
- Oriented from 5' to 3'.

 Single-stranded => flexible backbone => secondary structure => catalytic role.

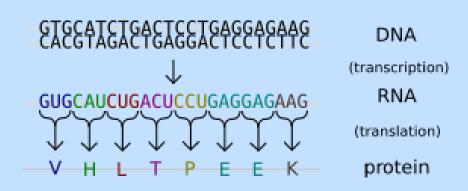


Translation

 Codon - a triplet of bases, codes a specific amino acid (except the stop codons); many-to-1 relation

Stop codons - signal termination of the protein

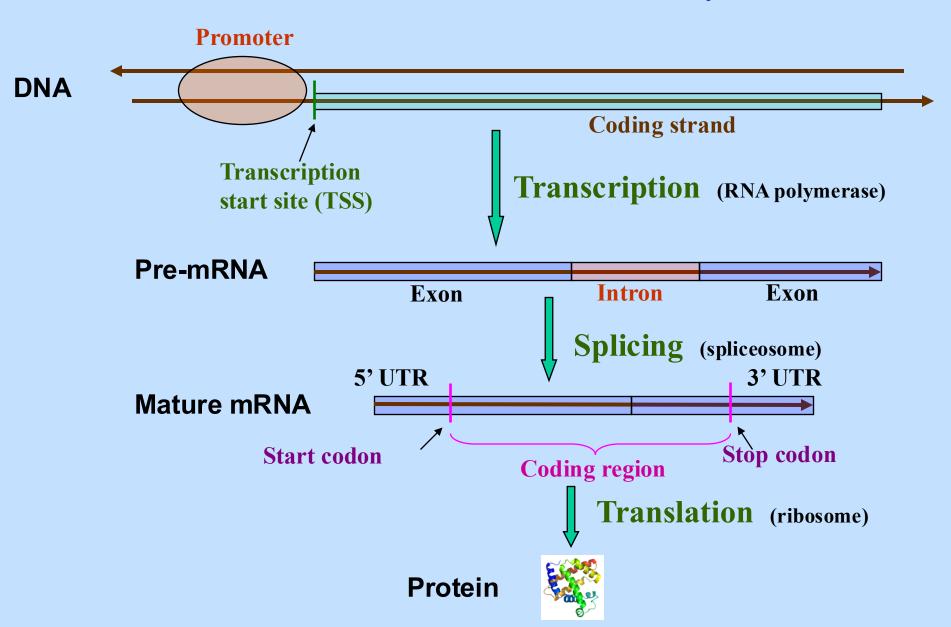
synthesis process



Second base of codon												
		U	С	A	G							
First base of codon	U	UUU UUC UUA UUG	UCU UCC UCA UCG	UAU UAC UAA UAG	UGU Cys UGC UGA UGG Trp	C A						
	С	CUU CUC CUA CUG	CCU CCA CCG	CAU His CAC GIN CAG GIN	CGU CGC CGA CGG	U Codon						
	A	AUU AUC AUA Met	ACU ACC ACA ACG	AAU Asn AAC Lys AAG Lys	AGU AGC AGA AGG	<mark>o≯ ∩⊂</mark> Third base						
	G	GUU GUC GUA GUG	GCU GCC GCA GCG	GAU GAC GAA CAG	GGU GGC GGA GGG	U C A						

The genetic code, written by convention in the form in which the Codons appear in mRNA. The three terminator codons, UAA, UAG, and UGA, are boxed in red; the AUG initiator codon is shown in green.

Gene structure (eukaryotes)



Genome sequences

 Many genomes have been sequenced, including those of viruses, microbes, plants and animals.

· Human:

- 23 pairs of chromosomes
- 3+Gbps (bps = base pairs), only ~3% are genes
- ~25,000 genes

Yeast:

- 16 chromosomes
- 20 Mbps
- 6,500 genes

Regulation of Expression

- Each cell contains an identical copy of the whole genome - but utilizes only a subset of the genes to perform diverse, unique tasks
- Most genes are highly regulated their expression is limited to specific
 tissues, developmental stages,
 physiological condition
- · Main regulatory mechanisms:
 - transcriptional regulation
 - post-transcriptional regulation

Post-transcriptional regulation

- Post-transcription is regulated primarily by RNAbinding proteins (RBPs) - proteins that bind to RNA subsequences, called binding sites (BSs)
- BSs of a particular RBP share a common pattern, or motif
- · Input:

0 5 10 15 20 25 30 35 40 45

TCTCATCCGGTGGGAATCACTGCCGCATTTGGAGCATAAACAATGGGGGG

TACGAAGGACAAACACTTTAGAGGTAATGGAAACACAACCGGCGCATAAA

ATACAAACGAAAGCGAGAAGCTCGCAGAAGCATGGGAGTGTAAATAAGTG

GGCGCCTCATTCTCGGTTTATAAGCCAAAACCTTGTCGAGGCAACTGTCA

TCAAATGATGCTAGCCGTCGGAATCTGGCGAGTGCATAAAAAGAGTCAAC

Output:

0 5 10 15 20 25 30 35 40 45

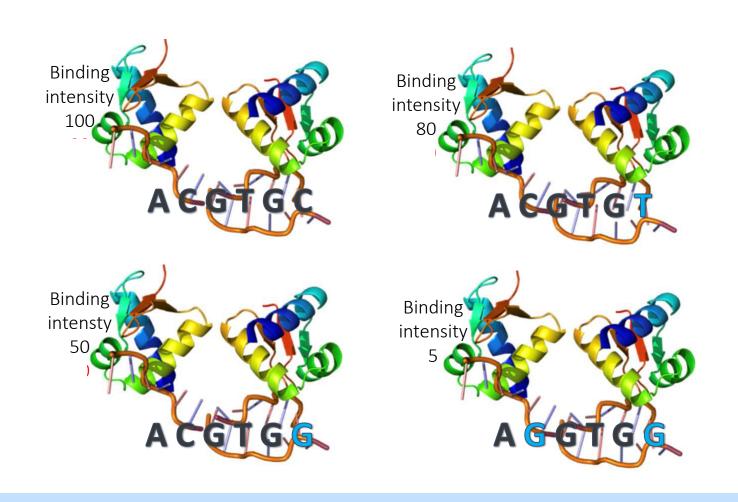
TCTCATCCGGTGGGAATCACTGCCGCATTT**GGAGCATAAA**CAATGGGGGG

TACGAAGGACAAACACTTTAGAGGTAATGGAAACACAACC**GGCGCATAAA**ATACAAACGAAAGCGAGAAGCTCGCAGAAGCATG**GGAGTGTAAA**TAAGTG

GGCGCCTCATTCTC**GGTTTATAAG**CCAAAACCTTGTCGAGGCAACTGTCA

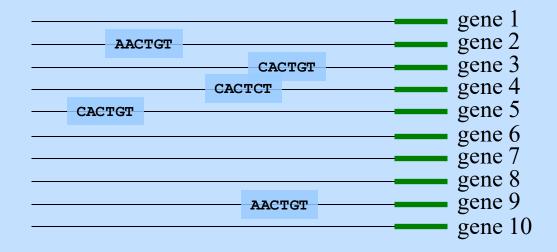
TCAAATGATGCTAGCCGTCGGAATCTGGCG**AGTGCATAAA**AAAGAGTCAAC

Protein RNA-binding



RBP BS motif models - strings

· Consensus ("degenerate") string:



List of k-mers (weighted or unweighted)

How can we use NN for these models?

RBP BS models - PWM

- Position weight matrix (PWM): each position has weights for the 4 possible letters (A, C, G, T)
- For example:

How can we use NN for this model?

	1	2	3	4	5	6
Α	0.1	8.0	0	0.7	0.2	0
С	0	0.1	0.5	0.1	0.4	0.6
G	0	0	0.5	0.1	0.4	0.1
Т	0.9	0.1	0	0.1	0	0.3

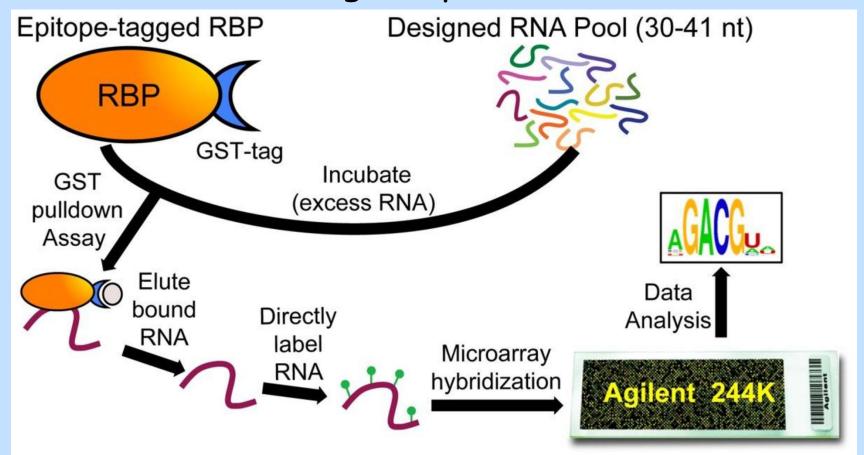
Logo format:



RNAcompete

Ray et al, Nat. Biotech 2009

- Generate an RNA pool covering all possible 9mers, each at least 16 times
- Detect RBP binding to specific 9-mers

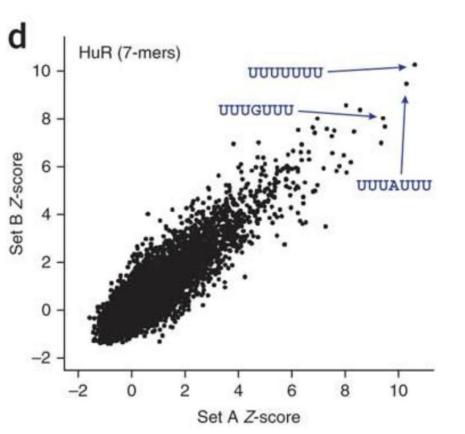


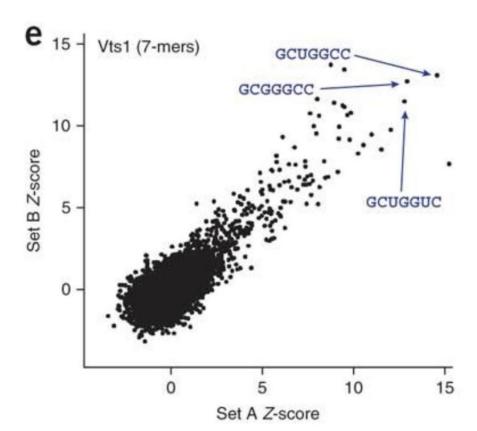
b RNAcompete

Vts1

AGACCUUCAUGUUCUGUUUCUUUUACUUUGGUUGU
AGAAGUUUGUUUUGGUUUUGGUUUUGCUUUGGUGCA
AGAUUUUUUUACUUUCAUGUCUAACUCACCUUUGGGGAA
AGGUUUCGUUUUGUGUACUUCCCCUAGUUUCUUUGGC
AGAAGUUAGUUUGUUUUUUUUGUGUUGAUUUAGUGGCCU
AGAAGUAGGGAAGUUUUUUUUUCGUUGAUUUAGUGCCCAC
AGAUUAUUUCUCGUUAUAAUAAUAGCGUUUAGUGCCCAC
AGAUUCAGAUUUAGCACCACUGUUGUAGUAUGUUUUGU
AGAAUUGAUUUAUUUUUGUCUCACCCAUCUGCGUU
AGACUGCUAACUUAUUUUUUGUCGUCCUAACGUUGCA

AGGGCCAAUGCGGUGCAGGCGCUGCGUUGGC
AGGUUCAAACGCAUGCGGGUCAAGCUACGAACGCCACU
AGAGAUGGAUUGGGCUGGCACCGGUCCAUC
AGGGCGUUUAACGGCGGGUCCCGUUAGGCGC
AGAGAGUGCAUAGCUGGCUUCUAUGCGCUC
AGGGCCAAUGCGGUGCAGGCGCUGCGUUGGC
AGAGUGUCAGGUACAUAAAGUAGCUCAAUGAGUUCGUU
AGAGUCCAGAACGGCAGGCACGUUUUGGAC
AGAGACAGGAGCUUAUUCAUUGAUCAGCGAUCGCG
AGAGGUUGAGACGGCAGGUCCCGUCUCGGCC





RNAcompete - implementation

- 30-41nt variable region
- A sequence library covering of all 9-mers, each at least 16 times

~240K probes

The computational challenge

- · Input:
 - Amino-acid sequence of a new protein (1 sequence file)
 - a list of RNA compete probes (1 sequence file)
- Goal: Predict RNA binding intensity for each RNAcompete probe
- Intuition: learning a binding model based on a protein's amino acid sequence

The project

General goals

- · Research
 - Learn about known solutions
 - Trial and error with training data
- Develop software from A-Z:
 - Design of deep neural networks
 - Implementation
 - Execution & analysis of test data
- A taste of bioinformatics
- Have fun
- · Get credit...

The computational task

 Given a set of RNAcompete data of different RBPs

 Learn a model that for a given RBP can score RNAcompete probes

- Main challenges:
 - Prediction performance
 - Training runtime

RBP Input

· 1 amino-acid sequence file

<sequence>\n

MSSKIFVGGNNVYIGYDNQKALGLILRRFETRKGFGFVTFDDQSRGKLIPYEVVGEVVKLGYRDQRG

RNAcompete Input

File with ~120K lines, each containing a probe sequence of length 30-41.

<sequence>\n

GTAATATTACATTCCGTTTAACCTGCGCGCCCTACGG
TTTCTCTATGCAGATTAGCTCAATTCCCAATACCTA
GTAATATTCATACGGCCTCGACTCATTACACGGCTT
CTATATGTGCATATATTACCCTACAGGTCGGATAGC
TTCGGATTGGGTGCATAATGAAGCAGGACTATTAAA
CCGAGGTATTTGCATTCAGTCGTCTTCTAATGAAG
CCCAGCCGGATTTGCATACCACGGACCGAGCAATGG
GAAGGATGGTATTAGCAGCGCGTTCGGGCCGTTCCAC
GTTACTATAATTAGCAGCGCGTTCGGGCCGTTCCAC

...

ACGTTTGCATTAATTTCCCCCACATTATTATGCGGA CCCAGTTAATTATGCGCCACGAGTCCGTCATAGTTA ACCTCATTAAATGTATTAAGGTTGTTTAAAGTTGGG

- The input sequence file is sorted lexicographically
- The output is a file with a binding intensity for input sequence (in the same order)

Solution example

 Train a model that receives a pair (amino acid RBP sequence, RNA probe sequence)

And outputs a binding intensity

- Protein representation can be using protein LLMs
- RNA representation by one-hot

Input schedule

You will be given:

200 development sets (RBP sequence + training RNA compete probes with binding intensities).

44 evaluation sets (RBP sequence + **test** RNAcompete probes). You have to assign a binding intensity to each RNAcompete probe.

Output

List of binding intensities - each corresponding to a an RNA probe sequence.

The goal

- To assign binding intensities to RNAcompete probes in the RNAcompete file
- Return a file with all binding intensities to all RNAcompete probes

 For this list we can compare to real values by Pearson correlation: cov(x,y)/(std(x)*std(y))

Implementation

- Input: the 1st argument is the RBP filename,
 and the 2nd is the RNA compete filename
- Output: a file with RNA binding intensities (in the same order of the RNA sequences)
- Training runtime will be measured
- Reasonable documentation

Submission

- · Electronic design document
- · Electronic code submission

- · 44 binding intensities files, e.g. RBP201.txt
- Executable / script for running time test

Design document

· 3-5 pages (pdf), Hebrew/English

 Briefly describe main goal, input and output of program

 Describe neural net architecture, algorithms, and scores

References

- Rapid and systematic analysis of the RNA recognition specificities of RNAbinding proteins
- · Ray et al. Nature Biotechnology, 2009

- ProbeRating: a recommender system to infer binding profiles for nucleic acidbinding proteins
- · Yang et al. Bioinformatics, 2020