

## Computational gene finding

Devika Subramanian  
Comp 470

## Outline (3 lectures)

- Lec 1
  - The biological context
  - Markov models and Hidden Markov models
- Lec 2
  - Ab-initio methods for gene finding
  - Comparative methods for gene finding
- Lec 3
  - Evaluating gene finding programs

(c) Devika Subramanian, 2007

2

## The biological context

- Introduction to the human genome and genes
- The central dogma: transcription and translation

(c) Devika Subramanian, 2007

3

## Facts about the human genome

- The human genome contains 3 billion chemical nucleotide bases (A, C, T, and G).
- About 30,000 genes are estimated to be in the human genome. Chromosome 1 (the largest human chromosome) has the most genes (2968), and the Y chromosome has the fewest (231).

(c) Devika Subramanian, 2007

4

## More facts

- The average gene consists of 3000 bases, but sizes vary greatly, with the largest known human gene being dystrophin at 2.4 million bases.

(c) Devika Subramanian, 2007

5

## More facts

- Genes appear to be concentrated in random areas along the genome, with vast expanses of non-coding DNA between.
- About 2% of the genome encodes instructions for the synthesis of proteins.
- We do not know the function of more than 50% of the discovered genes.

(c) Devika Subramanian, 2007

6

## More facts

- The human genome sequence is almost (99.9%) exactly the same in all people. There are about 3 million locations where single-base DNA differences occur in humans (Single Nucleotide Polymorphisms or SNPs).
- Over 40% of the predicted human proteins share similarity with fruit-fly or worm proteins.

(c) Devika Subramanian, 2007

7

## A great site to learn more

<http://www.dnai.org/index.htm>

(c) Devika Subramanian, 2007

8

## Genome sizes

Organism	Genome Size (Bases)	Estimated Genes
Human ( <i>Homo sapiens</i> )	3 billion	30,000
Laboratory mouse ( <i>M. musculus</i> )	2.6 billion	30,000
Mustard weed ( <i>A. thaliana</i> )	100 million	25,000
Roundworm ( <i>C. elegans</i> )	97 million	19,000
Fruit fly ( <i>D. melanogaster</i> )	137 million	13,000
Yeast ( <i>S. cerevisiae</i> )	12.1 million	6,000
Bacterium ( <i>E. coli</i> )	4.6 million	3,200
Human immunodeficiency virus (HIV)	9700	9

(c) Devika Subramanian, 2007

9

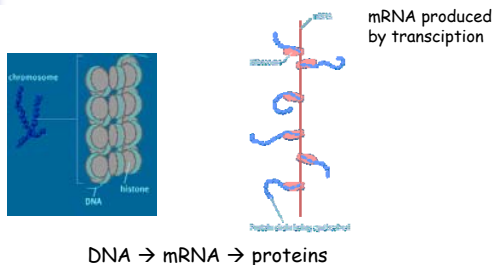
## Codons

- 3 consecutive DNA bases code for an amino acid. There are 64 possible codons, but only 20 amino acids (some amino acids have multiple codon representations).
- Four special codons: start codon (ATG) and three stop codons (TAG, TGA, TAA). They indicate the start and end of translation regions.

(c) Devika Subramanian, 2007

10

## The central dogma



(c) Devika Subramanian, 2007

11

## Transcription

- When a gene is "expressed" the sequence of nucleotides in the DNA is used to determine the sequence of amino acids in a protein in a two step process.
- First, the enzyme RNA polymerase uses one strand of the DNA as a template to synthesize a complementary strand of messenger RNA (mRNA) in a process called **transcription**. RNA is identical to DNA except that in RNA T is replaced with U (for uracil). Also, unlike DNA, RNA usually exists as a single stranded molecule.

(c) Devika Subramanian, 2007

12

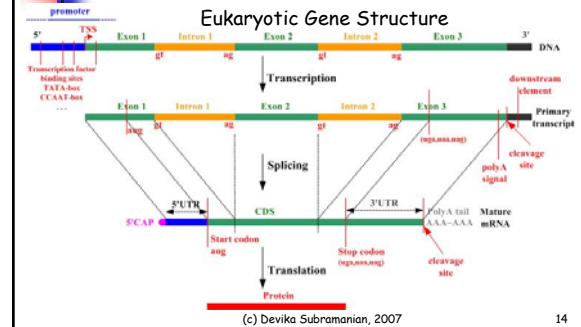
## Splicing and Translation

- In eukaryotes, after a gene is transcribed the introns are removed from the mRNA and the adjacent exons are **spliced** together in the nucleus prior to translation outside the nucleus.
- After the mRNA for a particular gene is made it is used as a template with which ribosomes synthesize the protein in a process called **translation**.

(c) Devika Subramanian, 2007

13

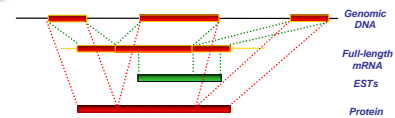
## The Biological Model



(c) Devika Subramanian, 2007

14

## How genes are validated



•cDNA - single-stranded DNA complementary to an RNA, synthesized from it by reverse transcription

•full-length mRNAs (GenBank *RefSeq*, ~16000 human sequences)

•ESTs - Expressed Sequence Tags

•relatively short, 500 bp long on average

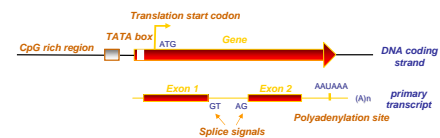
•span one or more exons

•large data sets required (GenBank *dbEST* - 4.3 M human sequences)

(c) Devika Subramanian, 2007

15

## Signals



•Upstream regulatory signals (TATA boxes)

•Translation start codon (ATG)

•Translation stop codon (e.g., TAA)

•Polyadenylation signal (~AATAAA)

•Splice recognition signals (e.g., GT-AG, branch point)

(c) Devika Subramanian, 2007

16

## Computational gene finding

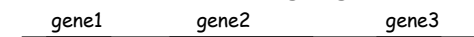
- Gene finding in prokaryotes
- Gene finding in eukaryotes
  - Ab initio
  - Comparative

(c) Devika Subramanian, 2007

17

## Finding genes in prokaryotes

- Prokaryotes are single-celled organisms without a nucleus (e.g., bacteria).
- Few introns in prokaryotic cells. Over 70% of *H. influenzae* genome codes for proteins.
- No introns in coding region.



(c) Devika Subramanian, 2007

18

## Finding genes in prokaryotes

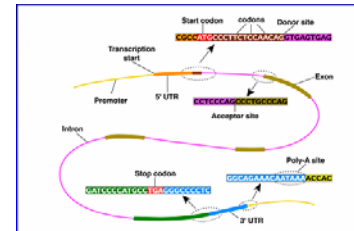
- Main idea: if bases were drawn uniformly at random, then a stop codon is expected once every 64/3 (about 21) bases. Since coding regions are terminated by stop codons, a simple technique to find genes is to look for long stretches of bases without a stop codon. Once a stop codon is found, we work backward to find the start codon corresponding to the gene.
- Main problems: misses short genes, overlapping ORFs.

(c) Devika Subramanian, 2007

19

## Computational gene finding

- Gene finding in eukaryotic DNA



(c) Devika Subramanian, 2007

20

## Ab initio methods

- Use information embedded in the genomic sequence *exclusively* to predict the gene structure.
- Find structure  $G$  representing gene boundaries + internal gene structure which maximizes the probability  $P(G|\text{genomic sequence})$ .
- Hidden Markov models are the predominant generative method for modeling the problem.

(c) Devika Subramanian, 2007

21

## Ab-initio methods

- Advantages
  - Intuitive, natural modeling
  - Prediction of 'novel' genes, *i.e.*, with no a priori known cDNA or protein evidence
- Caveats
  - Not effective in detecting alternatively spliced forms, interleaved or overlapping genes
  - Difficulties with gene boundary identification
  - Potentially large number of false positives with over-fitting

(c) Devika Subramanian, 2007

22