CS273A



Lecture 2: Protein Coding Genes

MW 1:30-2:50pm in Clark **S361*** (behind Peet's)

Profs: Serafim Batzoglou & Gill Bejerano

CAs: Karthik Jagadeesh & Johannes Birgmeier

* Handful of lectures/primers elsewhere: track on website/piazza

Announcements

- http://cs273a.stanford.edu/
 - Course guidelines, office hours, etc.
 - Lecture 1 is posted
 - Problem set 1 rolls out next week
- Course communications via Piazza
 - Auditors please sign up too



- The first tutorial this Friday in Beckman B-302 from 2:00pm-3:30pm.
 It's the only one some students should consider skipping. While they may be familiar with the first half of the Molecular Biology 101 lecture, we also cover gene regulation and genome rearrangements.
- CAs will be sending out a Doodle poll via Piazza to identify ideal times for office hours. Students can contact them via Piazza for questions.

Class Goals



- Meet your genome (learn to surf, learn the surf)
- Understand genomic tools (theory, applications)
- DIY (pose questions, write & run tools, understand answers)

Class Topics

- (0) Genome context: cells, DNA, central dogma
- (1) Genome content / genome function: genes, gene regulation, repeats, epigenetics
- (2) Genome sequencing: technologies, assembly/analysis, technology dependence
- (3) Genome evolution: evolution = mutation + selection, modes of evolution, comparative genomics, ultraconservation, exaptation
- (4) Population genomics:

 Tracking human migration patterns via neutral evolution
- (5) Genomics of human disease: disease susceptibility, cancer genomics, personal genomics
- (6) Genome "output" (organism) evolution: Evolutionary developmental biology ("evo-devo")

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CATA<del>CATGCTTCAACTACTTAATAAATGATTGTATGATAATGTTTTCAATGTAAGAGATTTCGATTATCCTTATAGTTCATA</del>CAT(
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Organism – Cell - Genome

10¹³ different cells in an adult human. The cell is the basic unit of life.

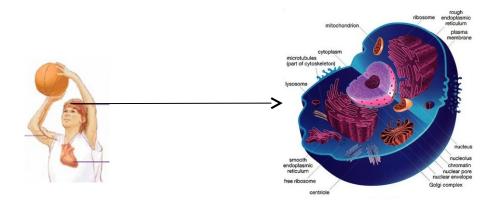
DNA = linear molecule inside the cell that carries instructions needed throughout the cell's life ~ long string(s) over a small alphabet

Alphabet (nucleotides/bases) {A,C,G,T}

Strings (chromosomes) of length 10⁴-10¹¹

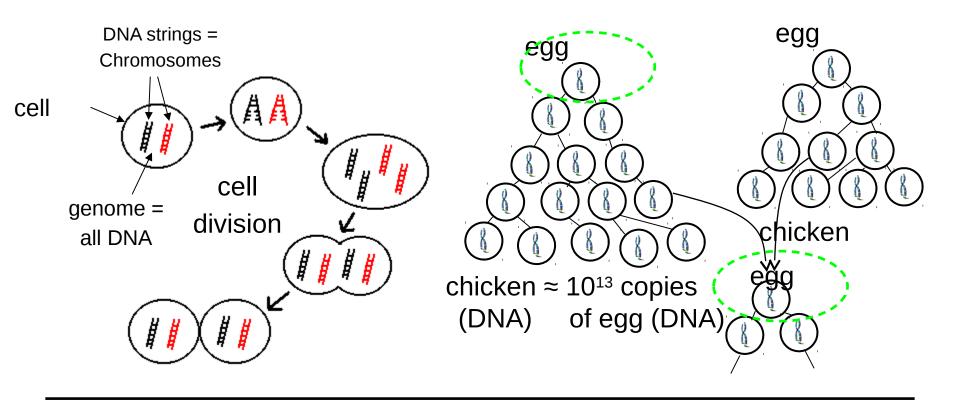
Genome:

"instruction"



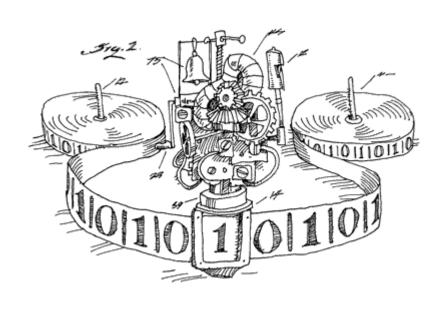
One Cell, One Genome, One Replication

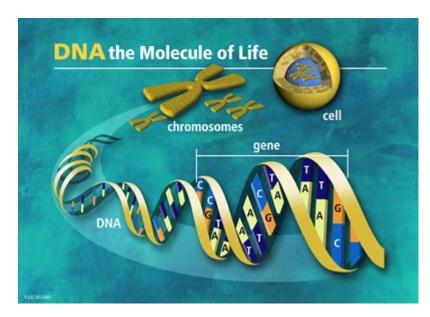
- •Every cell holds a copy of all its DNA = its genome.
- •The human body is made of $\sim 10^{13}$ cells.
- •All originate from a *single* cell through *repeated* cell divisions.



What will we study?

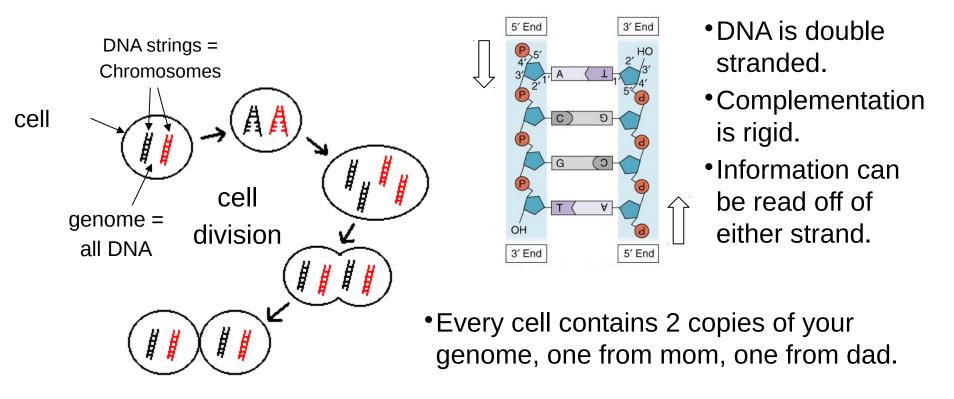
The most amazing "Turing tape" in existence, your genome.





How to Read The Genome

- •Genome = DNA.
- •Genome is broken up into several strings = chromosomes.
- •Humans: Females= (2*chr.1-22)+XX Males= (2*chr.1-22)+XY



The Biggest Challenge in Genomics...

... is <u>computational</u>:

How does this

encode this

GCTAGATCGCCTGGTA
GCTTTGCGCCGTCAAA
GTCTTGAAGGCTGTGA
ICAAGCTTCTTGCGAT
CCCGTTTGACCGGAGC
CTTGCCAATGAGTTCT
CAGCTGTCTATATGAA
ICACAAAAATACGCAAT





Program Output

This "coding" question has <u>profound</u> implications for our lives

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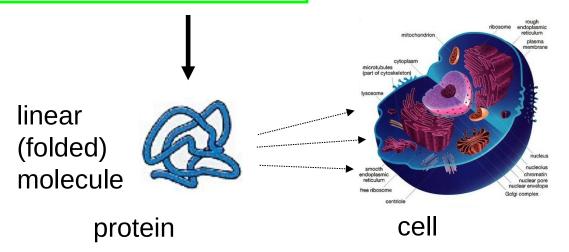
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Genomes, Genes & Proteins

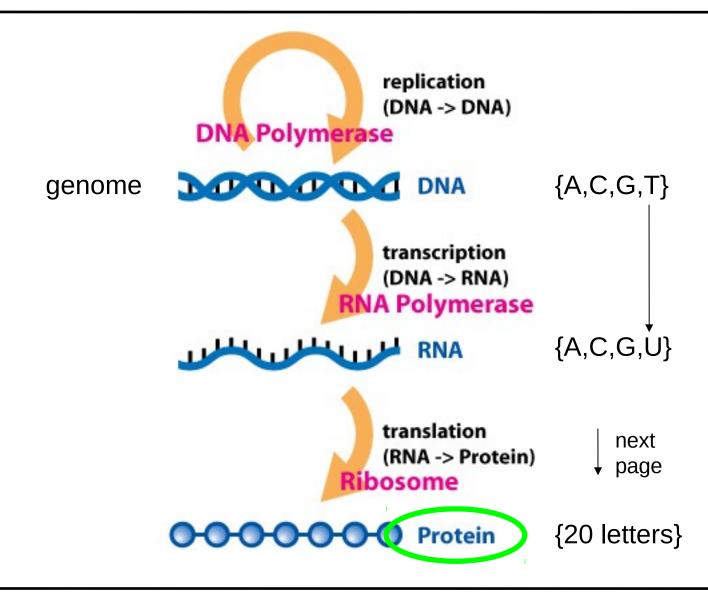
The most visible instructions in our genome are Genes. Genes explain exactly HOW to synthesize any protein. Proteins are the work horses of every living cell.

Genome: gene

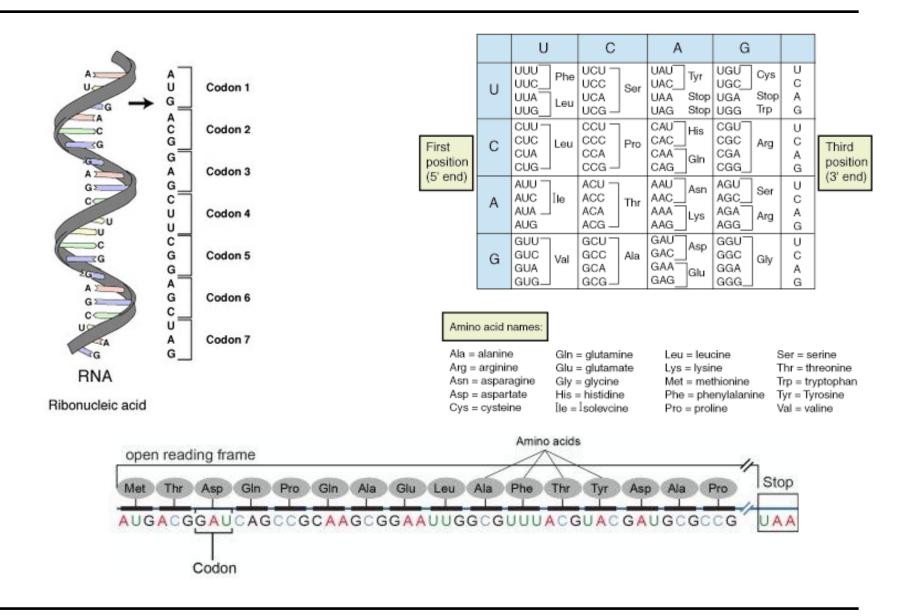
...ACGTACGACT GACTAGCATCGACTACGACTAGCAC...



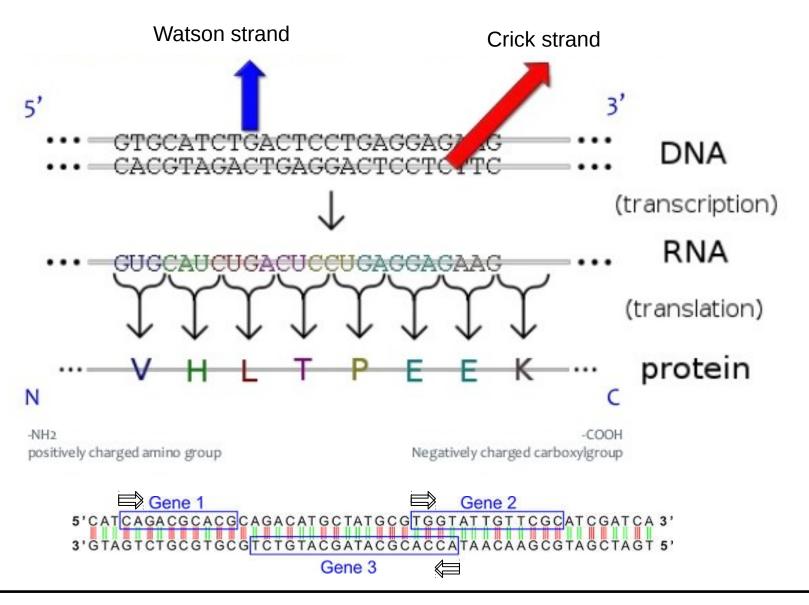
Central Dogma of Biology



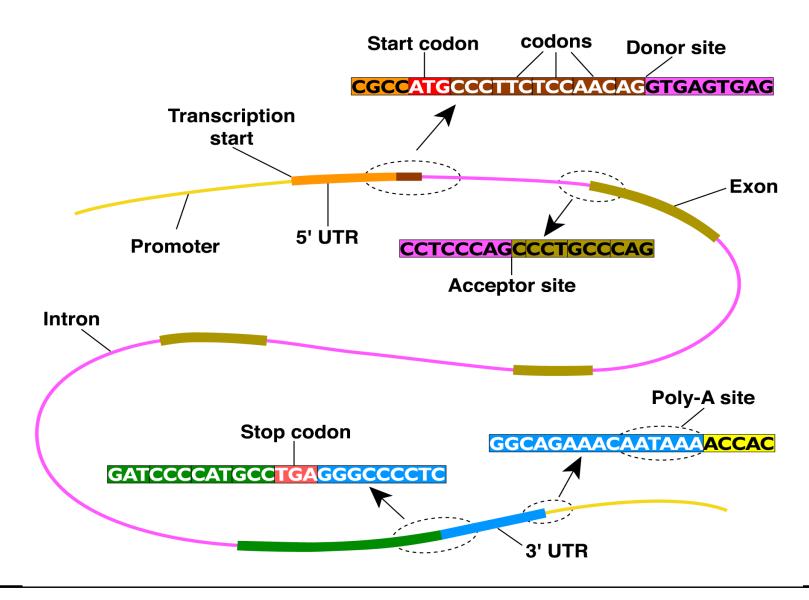
Translation: The Genetic Code



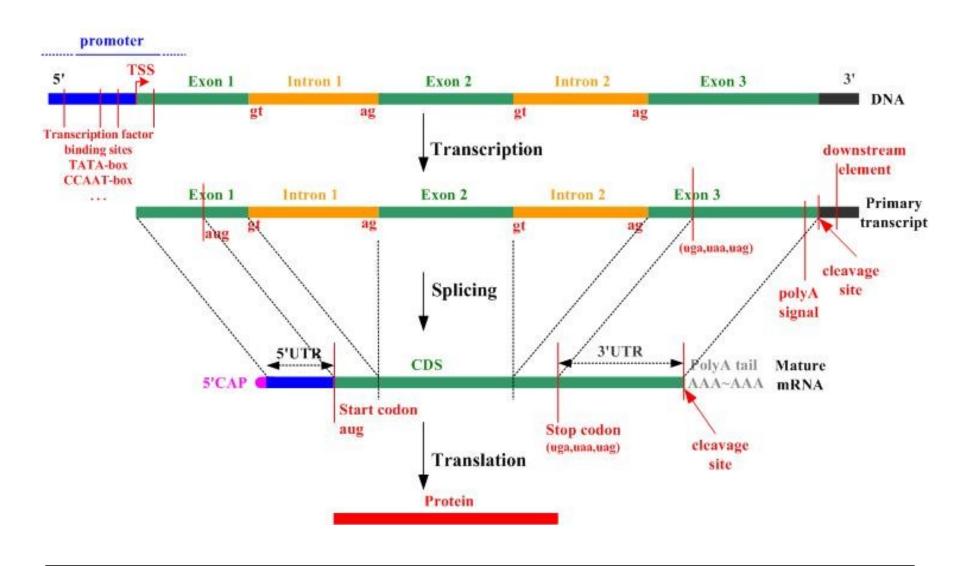
Genes Can Be Encoded on Either Strand



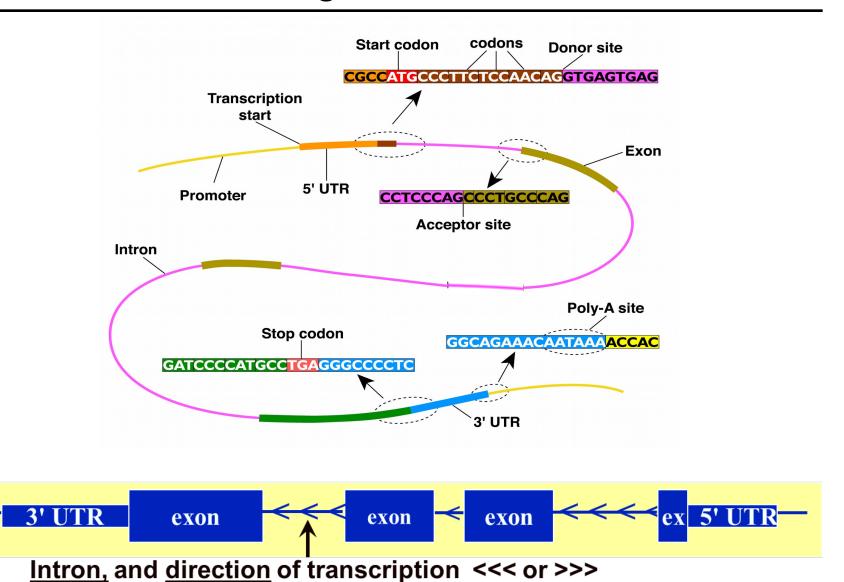
Gene Structure



Gene Splicing

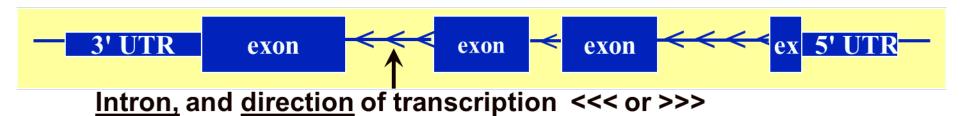


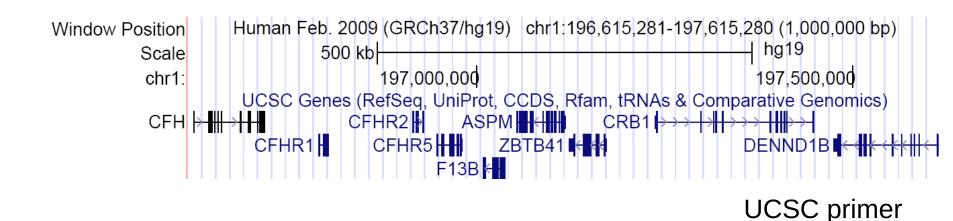
Visualizing Gene Structure



http://cs273a.stanford.edu [BejeranoFall15/16]

Genes in the Human Genome



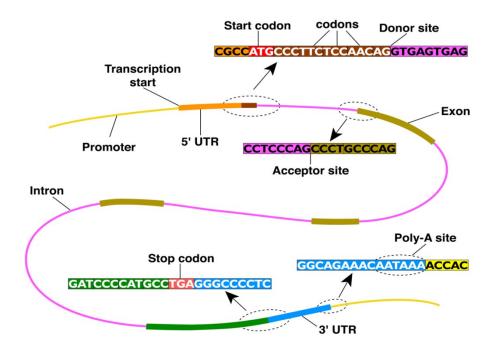


There are ~20,000 protein coding genes in the human genome. (Even half way through sequencing the human genome, Researchers thought there will be well over 100,000 genes).

Gene Finding

Computational Challenge:

"Find the genes, the whole genes, and nothing but the genes"



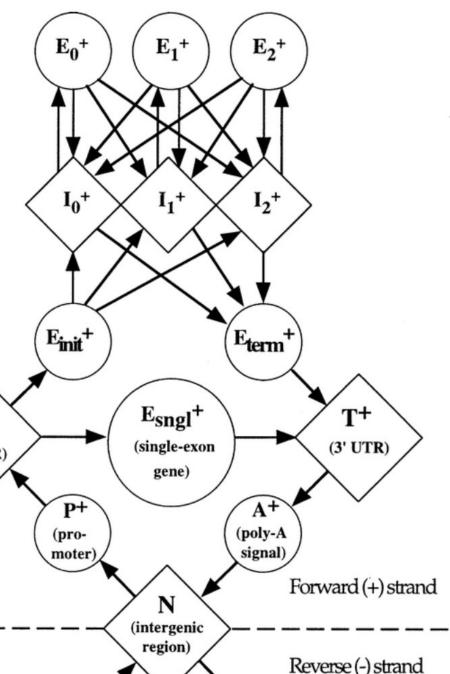
Gene prediction approachs

- Rule-based programs
 - Use explicit set of rules to make decisions.
 - Example: GeneFinder
- Neural Network-based programs
 - Use data set to build rules.
 - Examples: Grail, GrailEXP
- Hidden Markov Model-based programs
 - Use probabilities of states and transitions between these states to predict features.
 - Examples: Genscan, GenomeScan



GenScan States

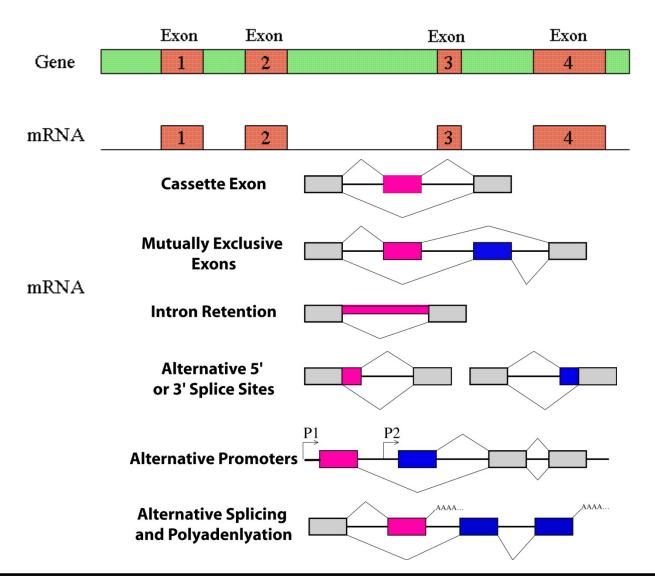
- N intergenic region
- P promoter
- F 5' untranslated region
- E_{sngl} single exon (intronless) (translation start -> stop codon)
- E_{init} initial exon (translation start -> donor splice site)
- \triangleright E_k phase k internal exon (acceptor splice site -> donor splice site)
- E_{term} terminal exon (acceptor splice site
 -> stop codon)
- I_k phase k intron: 0 between codons; 1
 after the first base of a codon
 the second base of a codon



Reverse (-) strand

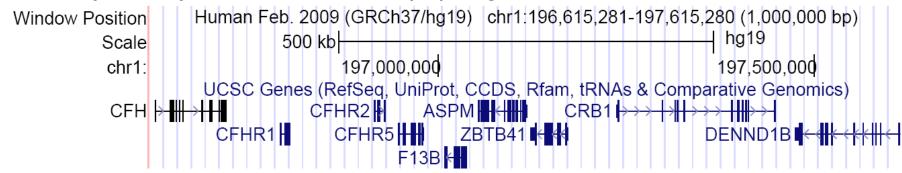
 F^+

Alternative Splicing

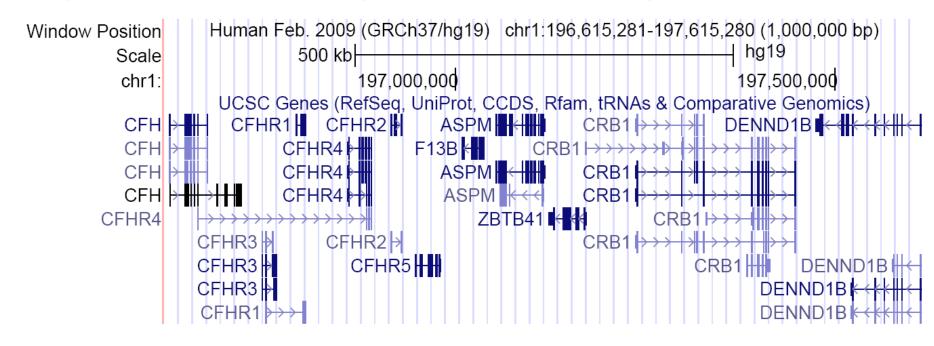


Genes in the Human Genome

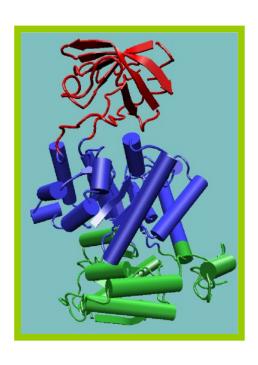
When you only show one transcript per gene locus:



If you ask the GUI to show you all well established gene variants:



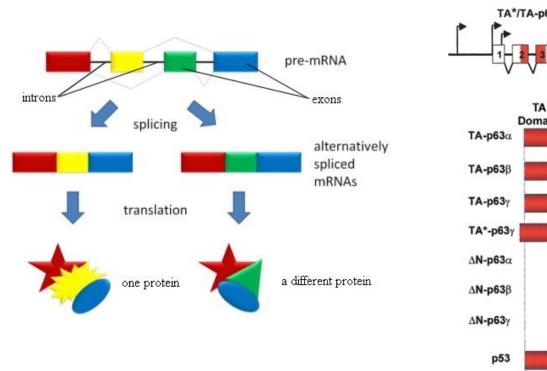
Protein Domains

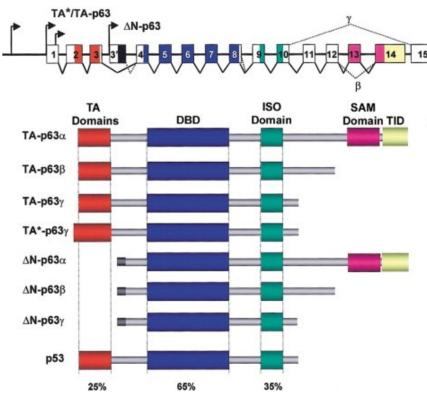


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NVRTATESFASDPILYRPVAVALDTKGPEIRTGLIKGSGTAEVE
LKKGATLKITLDNAYMAACDENILWLDYKNICKVVEVGSKVYVD
DGLISLQVKQKGPDFLVTEVENGGFLGSKKGVNLPGAAVDLPAV
SEKDIQDLKFGVDEDVDMVFASFIRKAADVHEVRKILGEKGKNI
KIISKIENHEGVRRFDEILEASDGIMVARGDLGIEIPAEKVFLA
QKMIIGRCNRAGKPVICATQMLESMIKKPRPTRAEGSDVANAVL
DGADCIMLSGETAKGDYPLEAVRMQHLIAREAEAAMFHRKLFEE
LARSSSHSTDLMEAMAMGSVEASYKCLAAALIVLTESGRSAHQV
ARYRPRAPIIAVTRNHQTARQAHLYRGIFPVVCKDPVQEAWAED
VDLRVNLAMNVGKAAGFFKKGDVVIVLTGWRPGSGFTNTMRVVP
VP

A protein domain is a subsequence of the protein that folds independently of the other portions of the sequence, and often confers to the protein one or more specific functions.

Alt. Splicing and Protein Repertoire





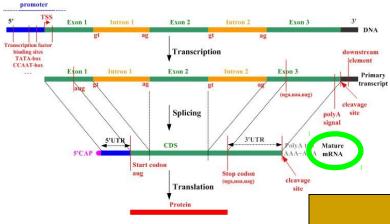
Alternative splicing often produces protein variants that have a different domain composition, and thus perform different functions. What if we want to predict all splice variants that are ever made?

Can we even do it from sequence alone?

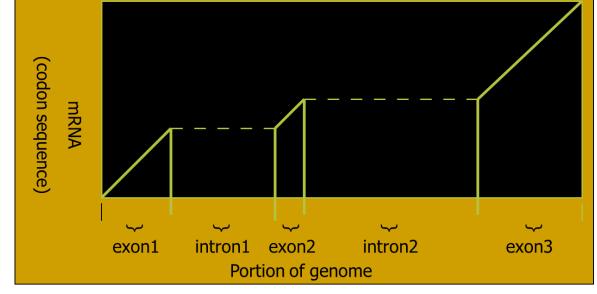
Common Problems

- Common problems with gene finders
 - Fusing neighboring genes
 - Spliting a single gene
 - Miss exons or entire genes
 - Overpredict exons or genes
- Other challenges
 - Nested genes
 - Noncanonical splice spites
 - Pseudogenes
 - Different isoforms of same gene

We can sequence all mRNA of a given cell

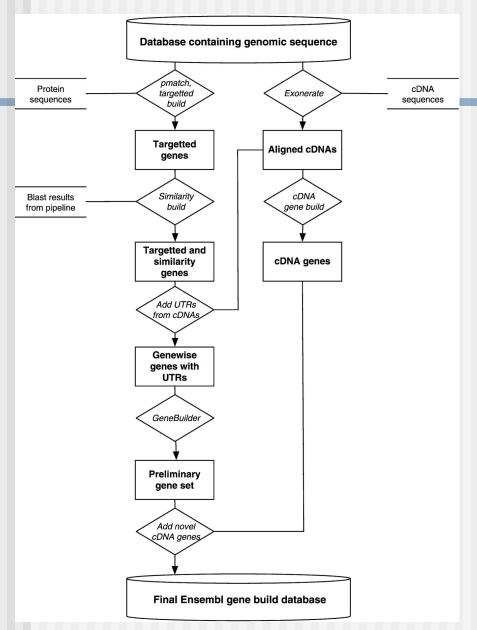


(Great, but not all genes/isoforms are expressed in all cells. Some are very exotic).



AUGGUG - - - - GGCCCUUUGGGA - - - - - CACUAA
GTGAGGATGGTAAATAGGGCAT - - - GGATTGAGCACUAATAA

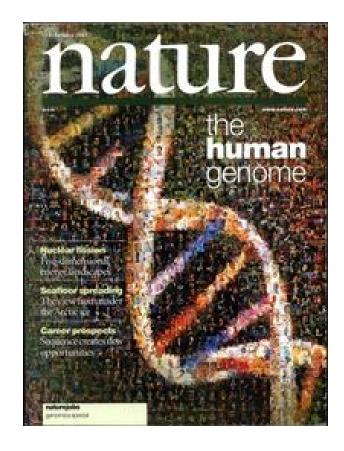
Gene Annotation System



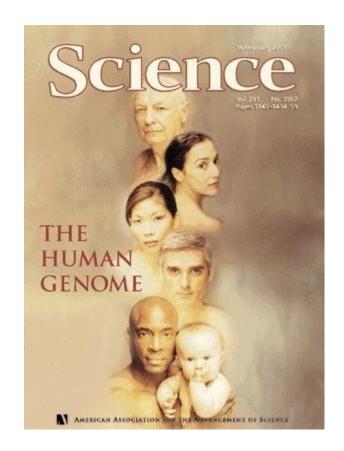
- All Ensembl gene predictions are based on experimental evidence
- Predictions based on manually curated Uniprot/Swissprot/Refseq databases
- UTRs are annotated only if they are supported by EMBL mRNA records

Val Curwen, et al. The Ensembl Automatic Gene Annotation System Genome Res., (2004) 14 942 - 950.

First full draft of the Human Genome



Human Genome Consortium (HGC)



Celera

2001

Everything in Genomics is a *Moving*Target

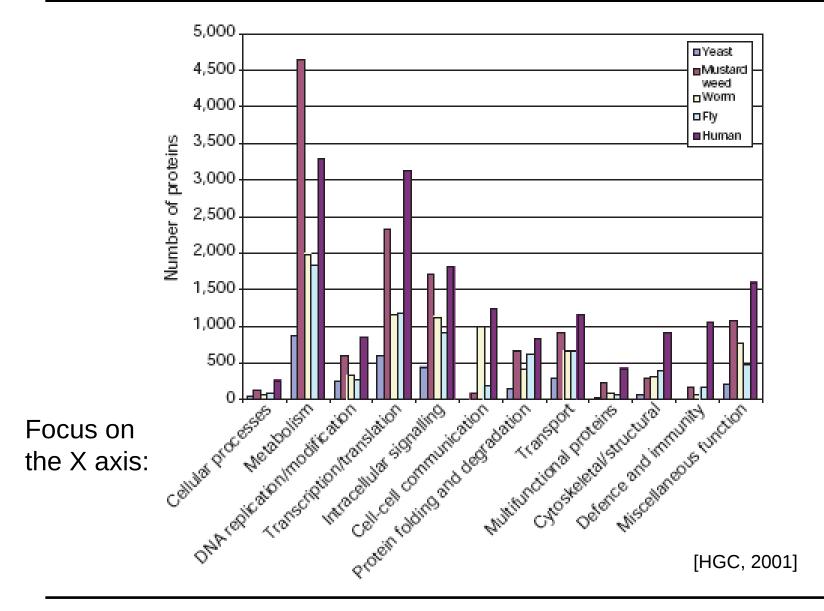
- The genomes (ie, assemblies)
- Their annotations
- Our understanding of Biology
- The portals

Conclusion: write code that can be run...



and rerun and rerun and rerun and rerun

Biological Functions of the Human Gene Set



Molecular Functions of the Human Gene Set

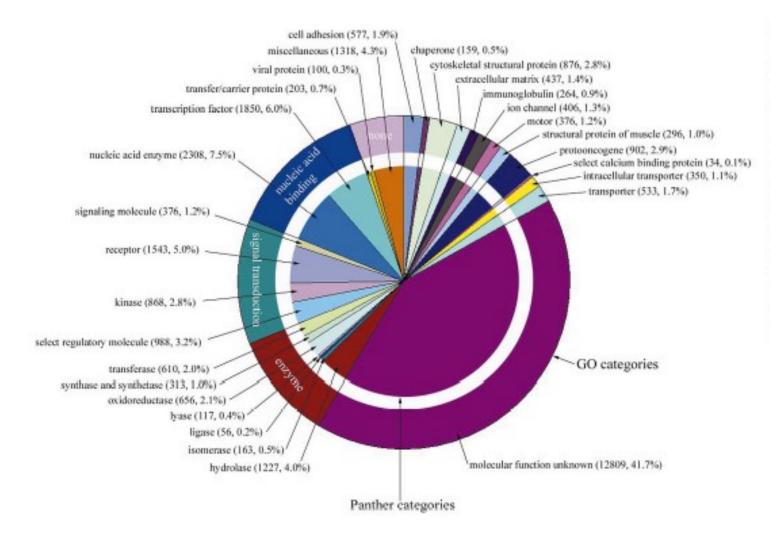
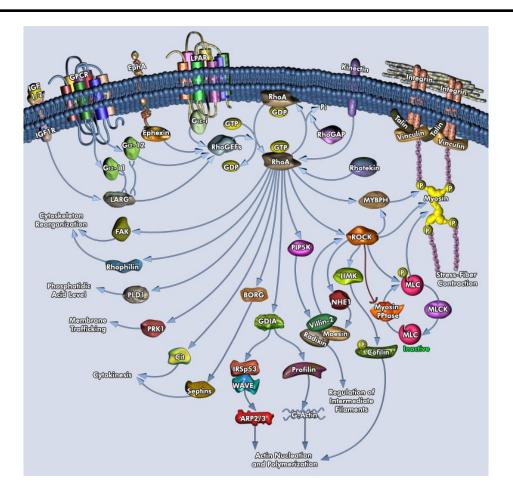


Fig. 15. Distribution of the molecular functions of 26,383 human genes. Each slice lists the numbers and percentages (in parentheses) of human gene functions assigned to a given category of molecular function. The outer circle shows the assignment to molecular function categories in the Gene Ontology (GO) (179), and the inner circle shows the assignment to Celera's Panther molecular function categories (116).

[Celera, 2001]

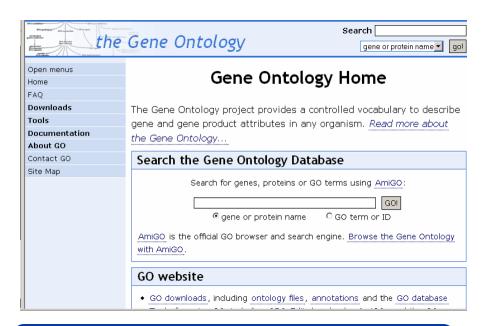
Biological vs. Molecular Function: Pathways

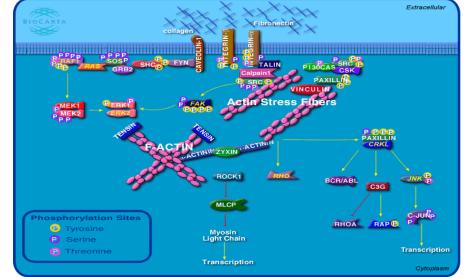


Proteins with very different molecular functions participate to manifest a single biological function, for example: a pathway.

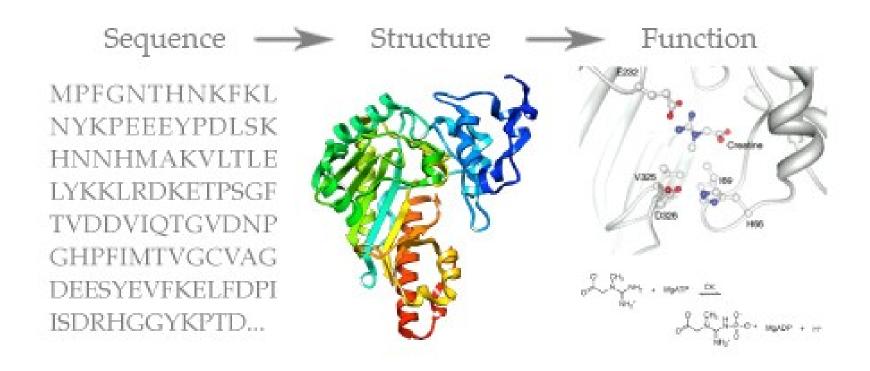
Gene Sets

- Gene Ontology ("GO")
 - Biological Process
 - Molecular Function
 - Cellular Location
- Pathway Databases
 - KEGG
 - BioCarta
 - Broad Institute
- Multiple others



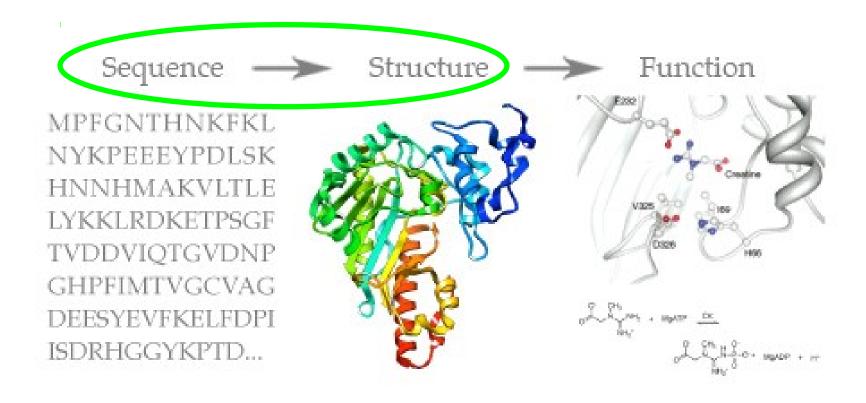


Genes & Their Functions



Gene (DNA) sequence determines protein (AA) sequence, which determines protein (3D) structure, which determines protein's function.

Protein Folding



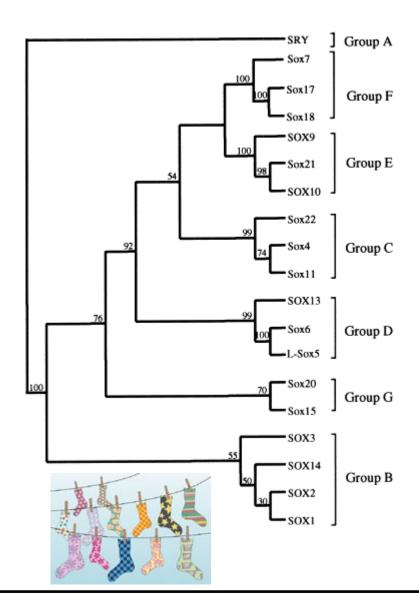
Protein folding is the challenge of deducing protein structure from protein sequence.

Gene Families, Gene Names

Genes (proteins) come in families. Genes of the same family have similar sequences.

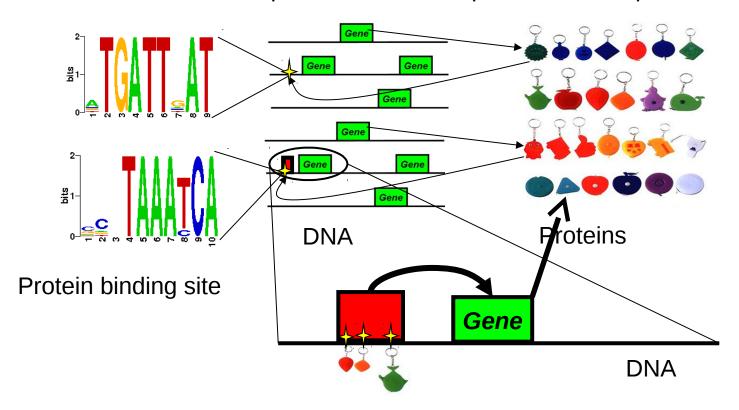
Which is why the fold into similar structure and perform similar functions.

Genes of the same family will typically have a "family name" followed by a (sequential) number or "first name".



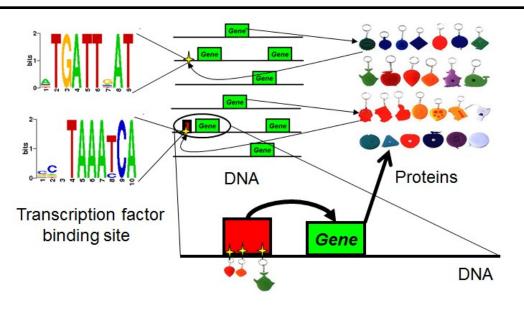
Some "Special" Functions: Gene Regulation

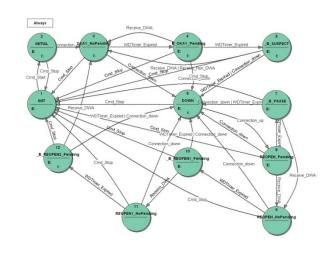
2,000 different proteins can bind specific DNA sequences.



Proteins that regulate the transcription of other proteins are called <u>transcription factors</u>.

The Importance of Gene Regulation



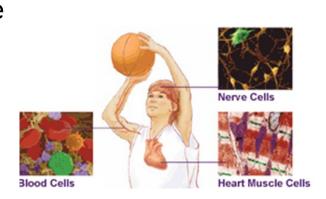


The looks & capabilities of different cells are determined by the subset of genes they express.

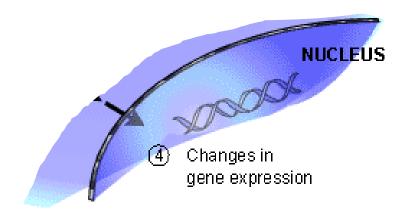
Different cell types express very different gene repertoires (from the same genome).

To change its behavior a cell can change its transcriptional program.

Think of it as a giant state machine...

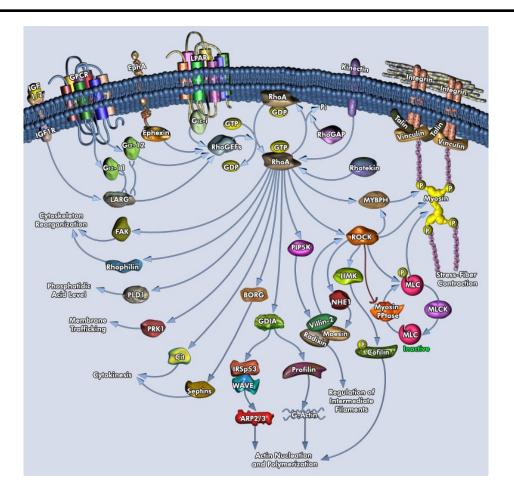


"Special" Function: Cell Signaling



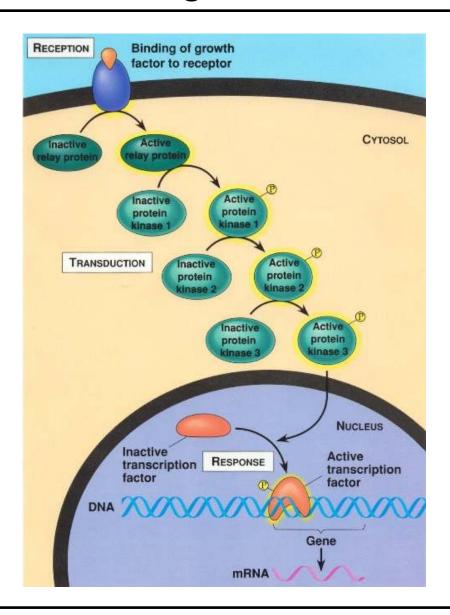
Cells also talk with each other. They send and receive messages, and change their behavior according to messages they receive.

Biological vs. Molecular Function: Pathways



Proteins with very different molecular functions participate to manifest a single biological function, for example: a pathway.

Signal Transduction



Now its an even bigger state machine of individual state machines (=cells) talking with each other, orchestrating their individual activities.

