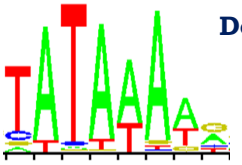
Algorithms in Bioinformatics

TWO

Motifs



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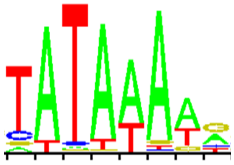
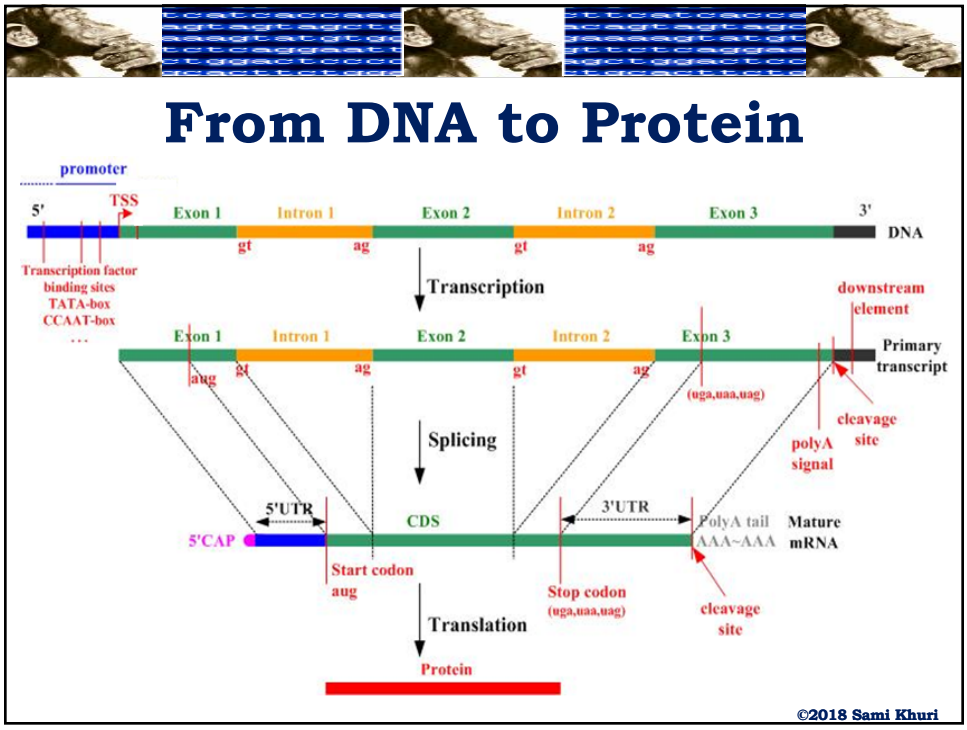
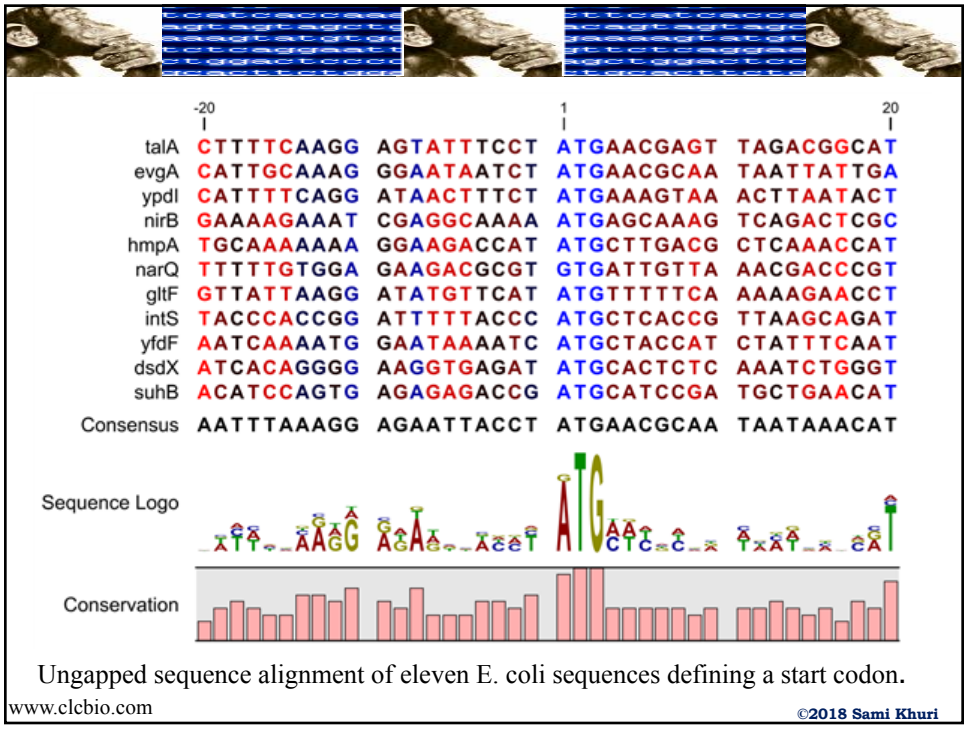


Importance and Abundance of Motifs

- DNA **motifs** are nucleotide sequence patterns of functional significance.
- **Examples**:
  - The **TATA box** is a motif that helps RNA polymerase find the transcription start site (TSS) in many eukaryotic genes.
  - The **CAT box** is another highly conserved region used for the initiation of transcription.





Ungapped sequence alignment of eleven E. coli sequences defining a start codon.

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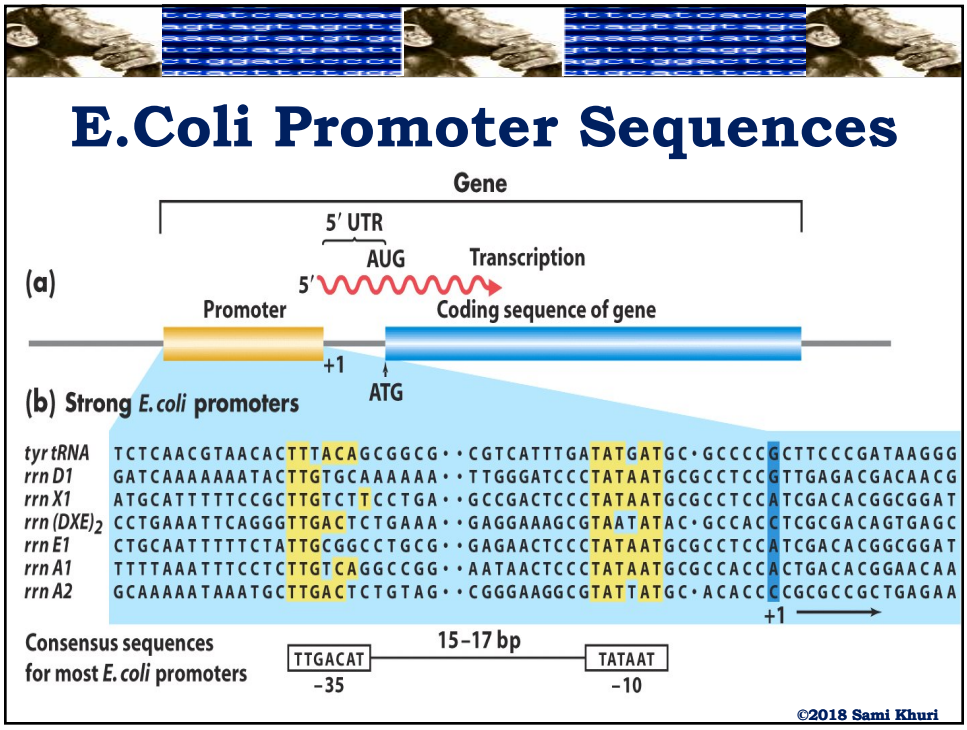
E.Coli Promoter Sequences
Gene
5′ UTR
AUG
Transcription
(a)
5′
Promoter
Coding sequence of gene
+1
ATG
(b) Strong E. coli promoters
tyr tRNA TCTCAACGTAACACTTTACAGCGGCG··CGTCATTTGATATGATGC·GCCCCGCTTCCCGATAAGGG
rrn D1 GATCAAAAAAATACTTGTGCAAAAAA··TTGGGATCCCTATAATGCGCCTCCGTTGAGACGACAACG
rrn X1 ATGCATTTTTCCGCTTGTCTTCCTGA··GCCGACTCCCTATAATGCGCCTCCATCGACACGGCGGAT
rrn (DXE)2 CCTGAAATTCAGGGTTGACTCTGAAA··GAGGAAAGCGTAATATAC·GCCACCTCGCGACAGTGAGC
rrn E1 CTGCAATTTTTCTATTGCGGCCTGCG··GAGAACTCCCTATAATGCGCCTCCATCGACACGGCGGAT
rrn A1 TTTTAAATTTCCTCTTGTCAGGCCGG··AATAACTCCCTATAATGCGCCACCACTGACACGGAACAA
rrn A2 GCAAAAATAAATGCTTGACTCTGTAG··CGGGAAGGCGTATTATGC·ACACCCGCGCCGCTGAGAA
+1
Consensus sequences for most E. coli promoters
TTGACAT
15–17 bp
TATAAT
–35
–10
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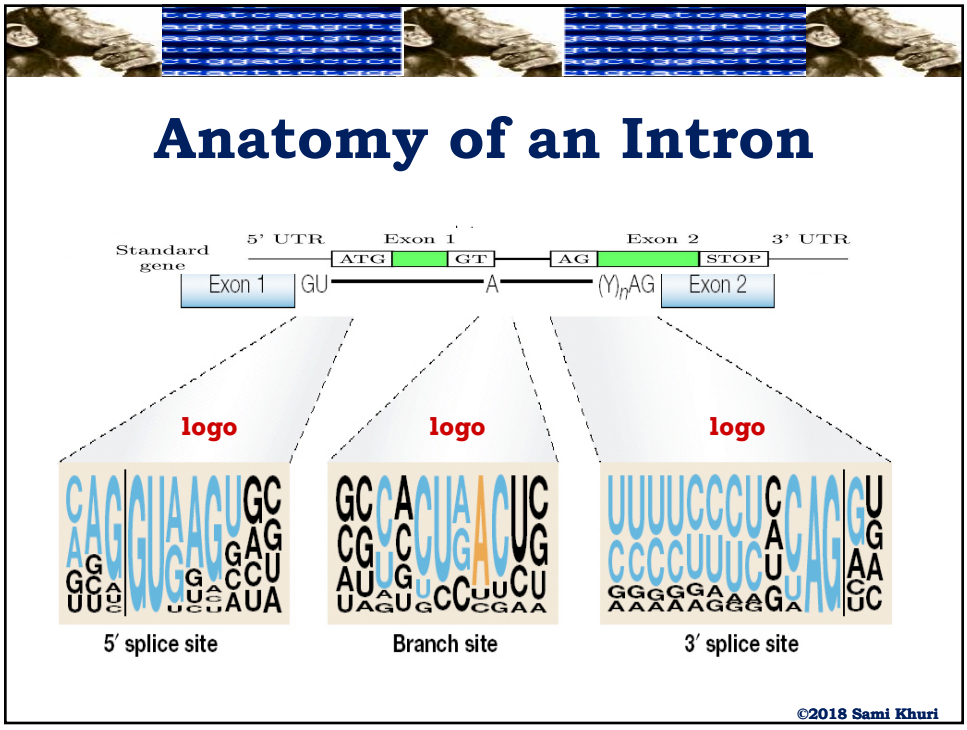
Anatomy of an Intron
Standard gene
5′ UTR
Exon 1
ATG
GT
AG
Exon 2
STOP
3′ UTR
Exon 1
GU
A
(Y)nAG
Exon 2
logo
logo
logo
5′ splice site
Branch site
3′ splice site
©2018 Sami Khuri

Sequence Motifs

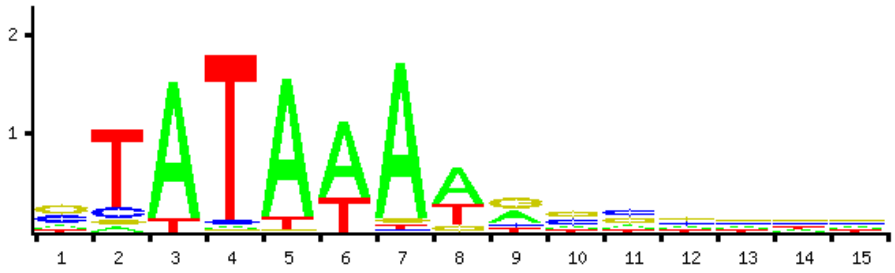

Table MM2.1 Nucleotide frequencies in 389 known TATA boxes.

Position	1	2	3	4	5	6	7	8	9	10	11	12	13	14	15
A	61	16	352	3	354	268	360	222	155	56	83	82	82	68	77
C	145	46	0	10	0	0	3	2	44	135	147	127	118	107	101
G	152	18	2	2	5	0	10	44	157	150	128	128	128	139	140
T	31	309	35	374	30	121	6	121	33	48	31	52	61	75	71

Detecting Motifs

A **motif** is a sequence pattern of functional significance.
Example: The **TATA box** is a motif that helps the polymerase find the transcription start site.

Table MM2.1 Nucleotide frequencies in 389 known TATA boxes.

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A	61	16	352	3	354	268	360	222	155	56	83	82	82	68	77
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G	152	18	2	2	5	0	10	44	157	150	128	128	128	139	140
T	31	309	35	374	30	121	6	121	33	48	31	52	61	75	71

Creating Tables of Frequencies

The probability of having an A in the first position is: 61/389 = 0.1568
The probability of a T in the second position is: 309/389 = 0.7943
Similarly for all 4 bases at all 15 positions.
We can thus create a table of frequencies.

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Position	1	2	3	4	5	6	7	8	9	10	11	12	13	14	15
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T	31	309	35	374	30	121	6	121	33	48	31	52	61	75	71

Creating Log-Odds Tables

Instead of creating a table of frequencies, we create a table of log-odds.
Suppose that the genome-wide average G and C content is 44%.
Then the probability of an A is 0.56/2 = 0.28.

$\log_2 (0.1568/0.28) = \log_2 (0.56) = - 0.84$.
Note that the base of the logarithm here is 2.
Similarly, $\log_2 (0.7943/0.28) = 1.5$.

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The Log-Odds Tables

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T	31	309	35	374	30	121	6	121	33	48	31	52	61	75	71

Table MM2.2 Position weight matrix.

A	−0.84	−2.77	1.69	−5.18	1.70	1.30	1.76	1.03	0.51	−0.96	−0.39	−0.41	−0.41	−0.68	−0.50
C	0.76	−0.90	−99.00	−3.10	−99.00	−99.00	−4.80	−5.42	−0.96	0.66	0.78	0.57	0.46	0.32	0.24
G	0.83	−2.25	−5.42	−5.42	−4.10	−99.00	−3.06	−0.96	0.88	0.81	0.58	0.58	0.58	0.70	0.71
T	−1.81	1.50	−1.64	1.78	−1.86	0.15	−4.14	0.15	−1.72	−1.18	−1.81	−1.07	−0.84	−0.54	−0.62

Taking Log-Odds

$$\frac{P(observed)}{P(expected)} \text{ is } \begin{cases} > 1 \\ = 1 \\ < 1 \end{cases}$$

$$\log_b \left(\frac{P(observed)}{P(expected)} \right) \text{ is } \begin{cases} > 0 \\ = 0 \\ < 0 \end{cases}$$

What is the Significance of Log-Odds

- If the nucleotide is **more likely** to occur at a given position than it is to occur overall, the ratio will be **bigger than 1.0** and the **log odds is positive**.
- If the nucleotide is **less likely** to occur at a certain position than it is to occur overall, then the ratio will be **smaller than 1.0** and the **log odds is negative**.

Using Log-Odds Tables (I)

Table MM2.2 Position weight matrix.

A	−0.84	−2.77	1.69	−5.18	1.70	1.30	1.76	1.03	0.51	−0.96	−0.39	−0.41	−0.41	−0.68	−0.50
C	0.76	−0.90	−99.00	−3.10	−99.00	−99.00	−4.80	−5.42	−0.96	0.66	0.78	0.57	0.46	0.32	0.24
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Table MM2.3 PWM score of the 15 bp sequence ACATATATAAGCTGG.

	A	C	A	T	A	T	A	T	A	A	G	C	T	G	G
A	−0.84	−2.77	1.69	−5.18	1.70	1.30	1.76	1.03	0.51	−0.96	−0.39	−0.41	−0.41	−0.68	−0.50
C	0.76	−0.90	−99.00	−3.10	−99.00	−99.00	−4.80	−5.42	−0.96	0.66	0.78	0.57	0.46	0.32	0.24
G	0.83	−2.25	−5.42	−5.42	−4.10	−99.00	−3.06	−0.96	0.88	0.81	0.58	0.58	0.58	0.70	0.71
T	−1.81	1.50	−1.64	1.78	−1.86	0.15	−4.14	0.15	−1.72	−1.18	−1.81	−1.07	−0.84	−0.54	−0.62

Table MM2.2 was constructed as explained in the previous slides; in other words, by taking the log of the ratio of the observed frequency over the expected frequency.

Using Log-Odds Tables (II)

Table MM2.2 Position weight matrix.

A	−0.84	−2.77	1.69	−5.18	1.70	1.30	1.76	1.03	0.51	−0.96	−0.39	−0.41	−0.41	−0.68	−0.50
C	0.76	−0.90	−99.00	−3.10	−99.00	−99.00	−4.80	−5.42	−0.96	0.66	0.78	0.57	0.46	0.32	0.24
G	0.83	−2.25	−5.42	−5.42	−4.10	−99.00	−3.06	−0.96	0.88	0.81	0.58	0.58	0.58	0.70	0.71
T	−1.81	1.50	−1.64	1.78	−1.86	0.15	−4.14	0.15	−1.72	−1.18	−1.81	−1.07	−0.84	−0.54	−0.62

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	A	C	A	T	A	T	A	T	A	A	G	C	T	G	G
A	−0.84	−2.77	1.69	−5.18	1.70	1.30	1.76	1.03	0.51	−0.96	−0.39	−0.41	−0.41	−0.68	−0.50
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T	−1.81	1.50	−1.64	1.78	−1.86	0.15	−4.14	0.15	−1.72	−1.18	−1.81	−1.07	−0.84	−0.54	−0.62

To see if a sequence of length 15 is a TATA box, we simply add the corresponding values from the PWM and see if we get a value above some threshhold.
In the example above, we add the 15 highlighted numbers to get 6.78.

Consensus Sequence and PWM

- All current methods for representing DNA motifs involve either consensus sequences or probabilistic models (such as PWM) of the motif.
- Consensus sequences do not adequately represent the variability seen in promoters or transcription factor binding sites.
- Both consensus sequences and PWM models assume positional independence. Neither method can accommodate correlations between positions.
- Probabilities calculated from PWM models can be highly misleading.

Classification Based Statistics

- Quantitative method to evaluate:
  - how well one can distinguish between cases and controls.
  - how well a diagnostic test performs in testing for some disease.

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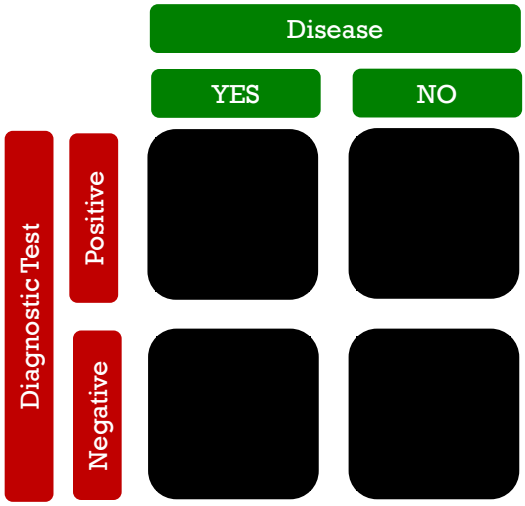

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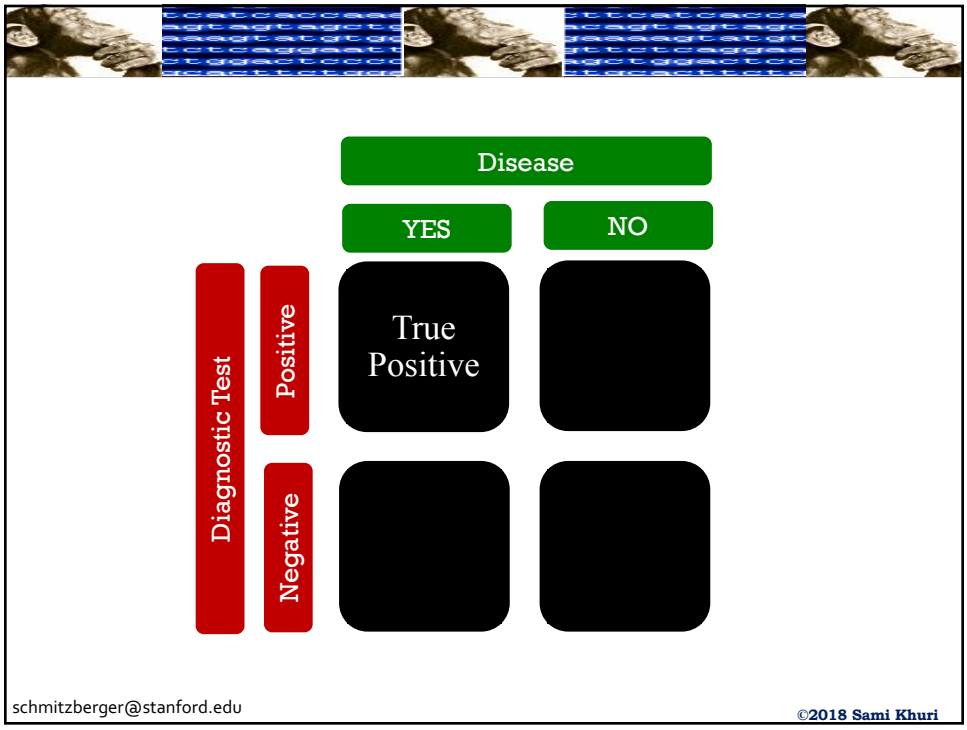
Disease
YES
NO
Diagnostic Test
Positive
Negative
True Positive
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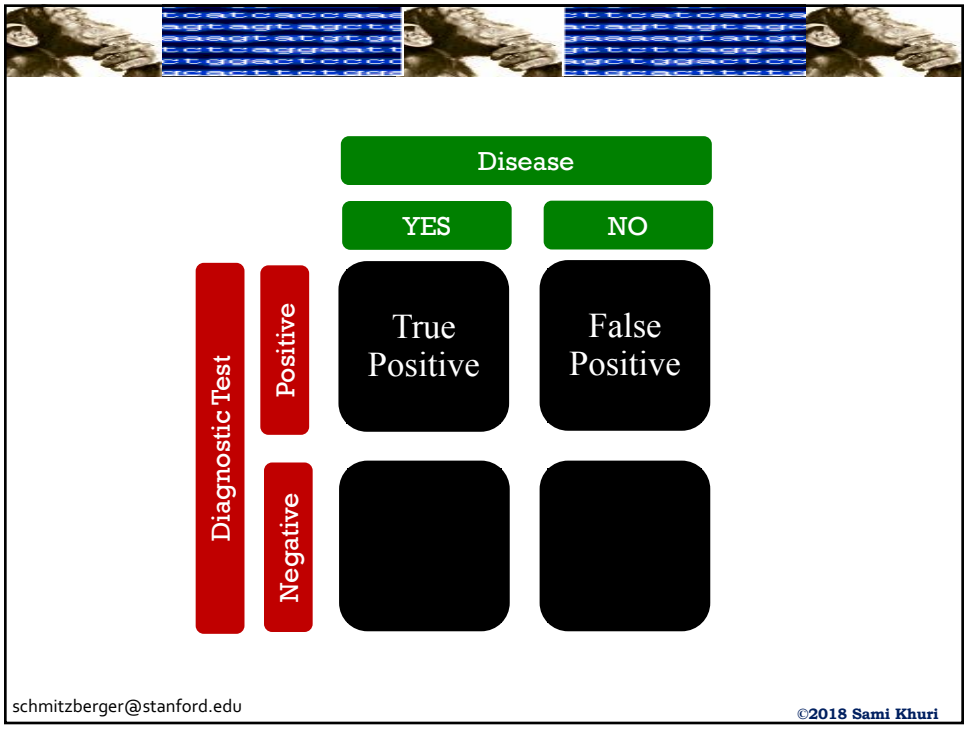
Disease
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Diagnostic Test
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True Positive
False Positive
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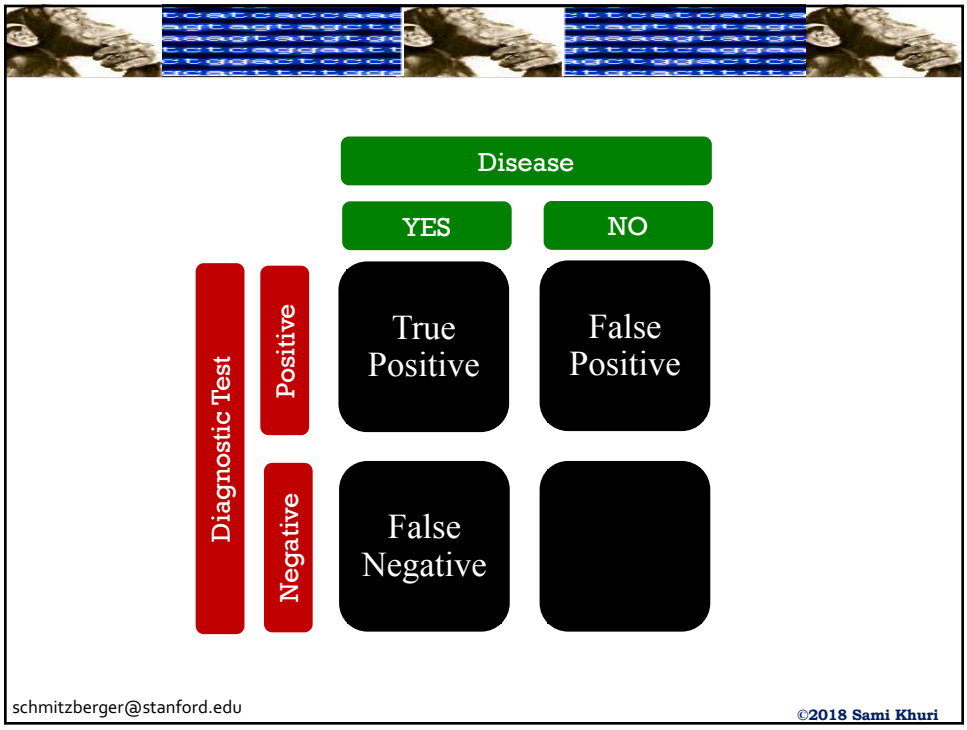
Disease
YES
NO
Diagnostic Test
Positive
Negative
True Positive
False Positive
False Negative
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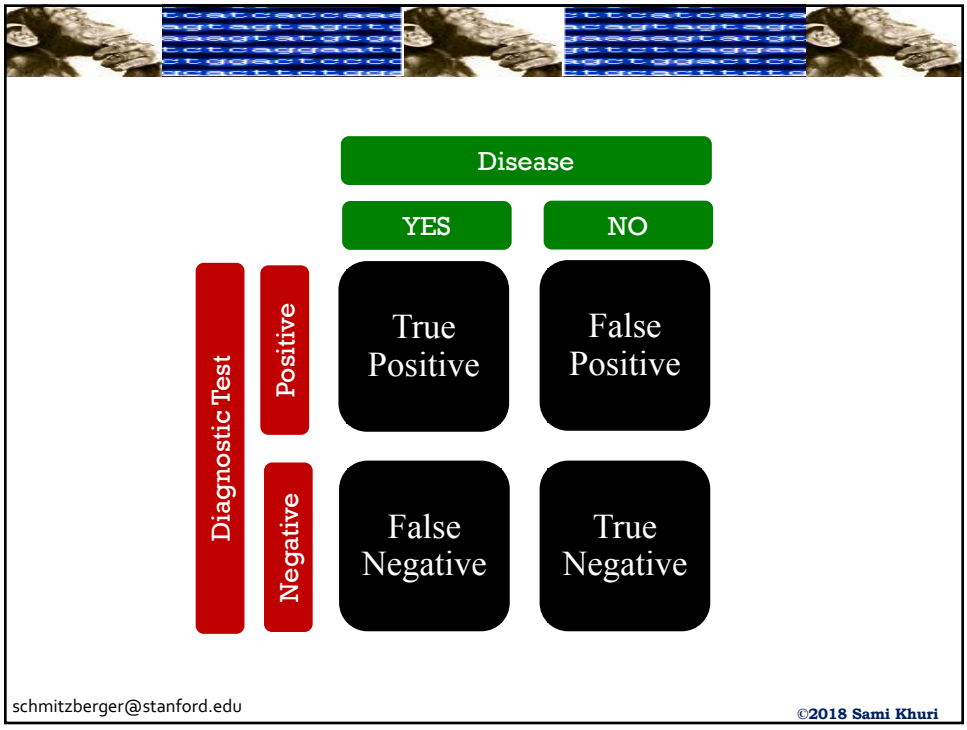
Disease
YES
NO
Diagnostic Test
Positive
Negative
True Positive
False Positive
False Negative
True Negative
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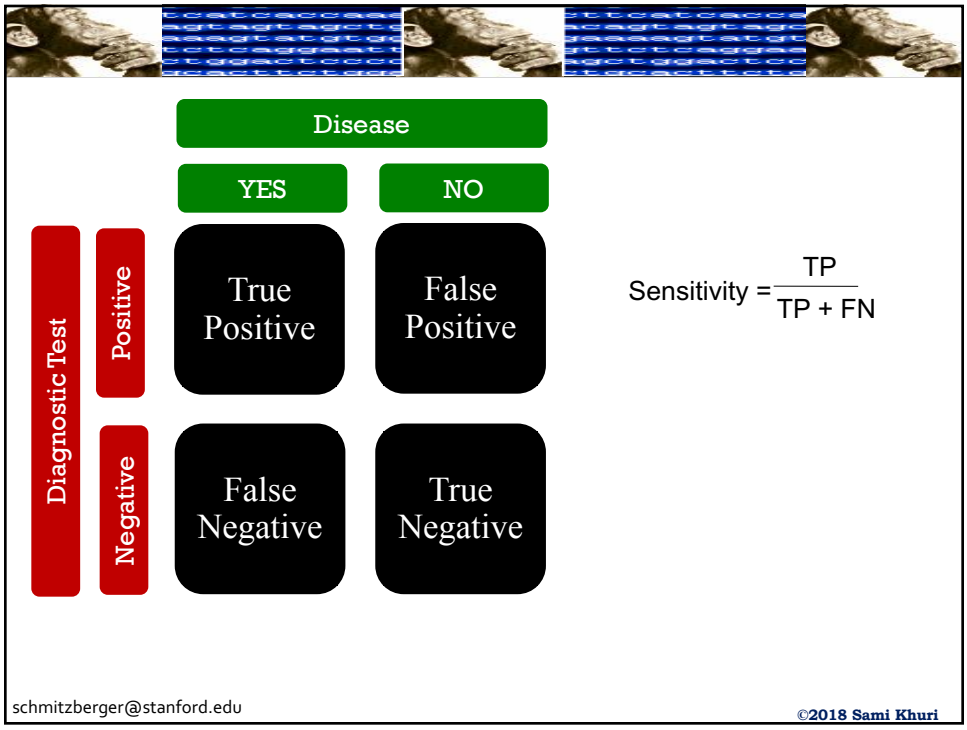
Disease
YES
NO
Diagnostic Test
Positive
Negative
True Positive
False Positive
False Negative
True Negative
Sensitivity = TP / TP + FN
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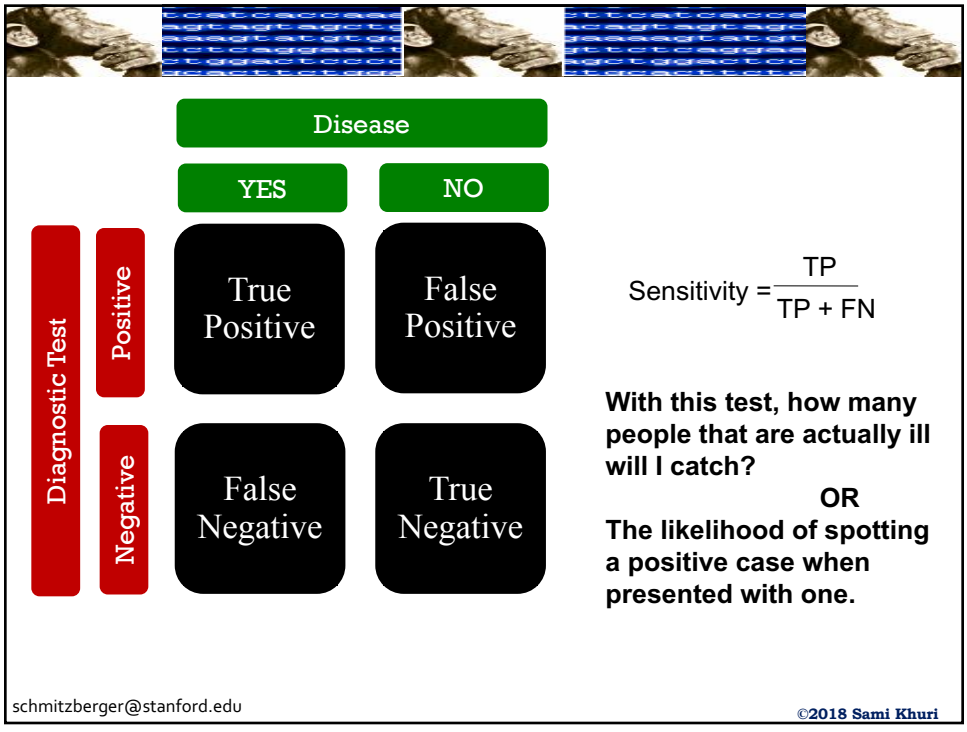
Disease
YES
NO
Diagnostic Test
Positive
Negative
True Positive
False Positive
False Negative
True Negative
Sensitivity = TP / TP + FN
With this test, how many people that are actually ill will I catch?
OR
The likelihood of spotting a positive case when presented with one.
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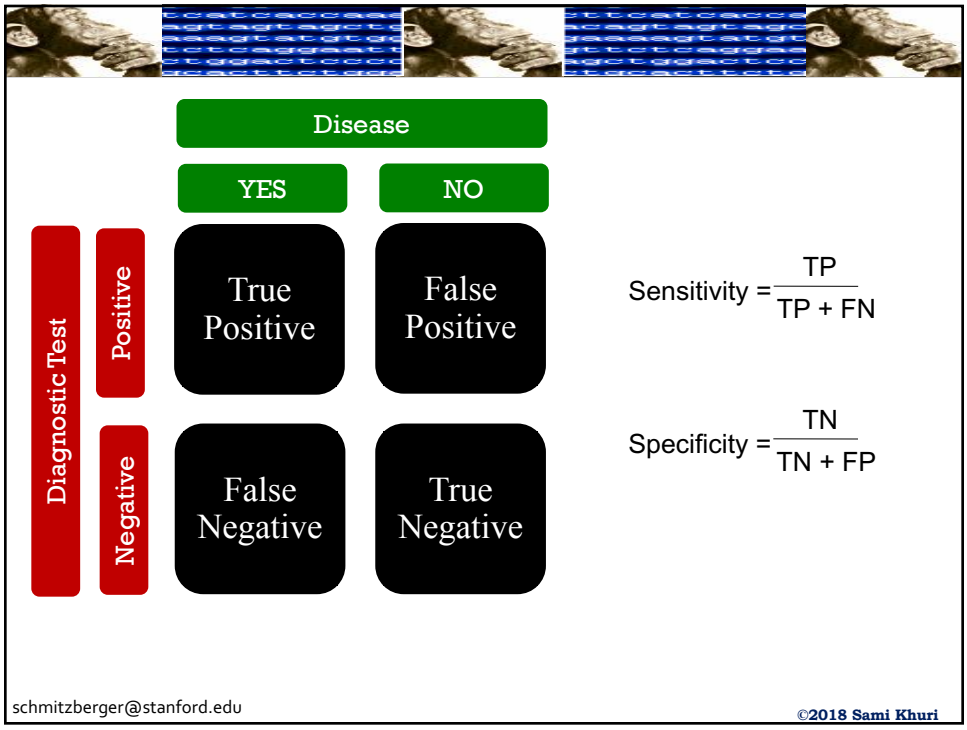
Disease
YES
NO
Diagnostic Test
Positive
Negative
True Positive
False Positive
False Negative
True Negative
Sensitivity = TP / TP + FN
Specificity = TN / TN + FP
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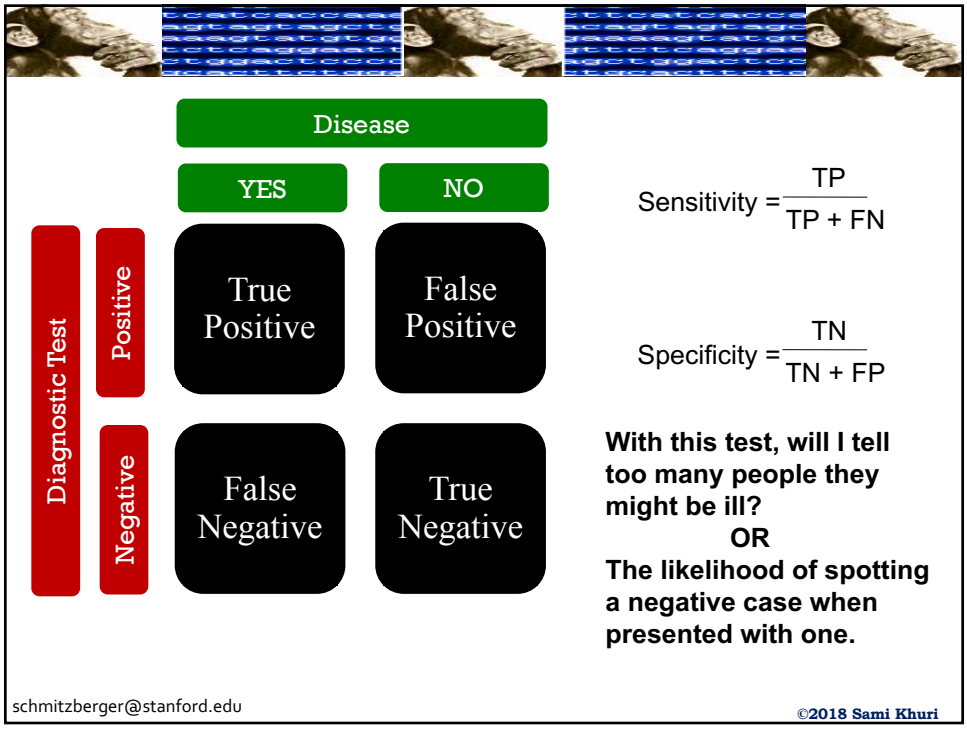
Disease
YES
NO
Diagnostic Test
Positive
Negative
True Positive
False Positive
False Negative
True Negative
Sensitivity = TP / TP + FN
Specificity = TN / TN + FP
With this test, will I tell too many people they might be ill?
OR
The likelihood of spotting a negative case when presented with one.
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Medical Test Evaluation

- True Positives = Test states you have the disease when you do have the disease
- True Negatives = Test states you do not have the disease when you do not have the disease
- False Positives = Test states you have the disease when you do not have the disease
- False Negatives = Test states you do not have the disease when you do

Evaluating Medical Tests

- Sensitivity =The probability of having a positive test result among those with a positive diagnosis for the disease
  - Sensitivity
    = True Positives / True Positives + False Negatives
- Specificity = The probability of having a negative test result among those with a negative diagnosis for the disease
  - Specificity
    = True Negatives / True Negatives + False Positives