Understanding Statisticians Lingo

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- Don't need to be statistician to appreciate & understand statistical results
- Good study design always trumps complicated statistics
- Statistics can be manipulated
- Statistics is a tool
- Common sense prevails



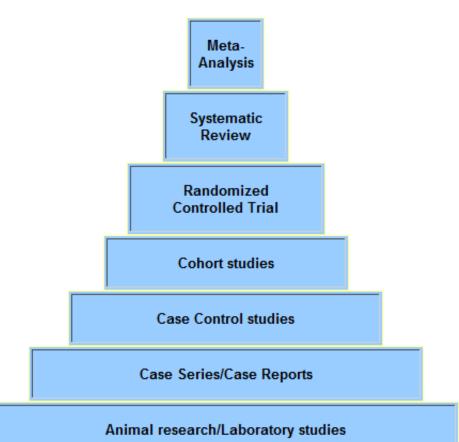




- Practically any study can be designed to show what you want it to show
- No single way to analyze data although some tests more *appropriate* than others
- Statistical significance does not always mean clinical significance
- Good tables & graphs often more informative than text



Study Types











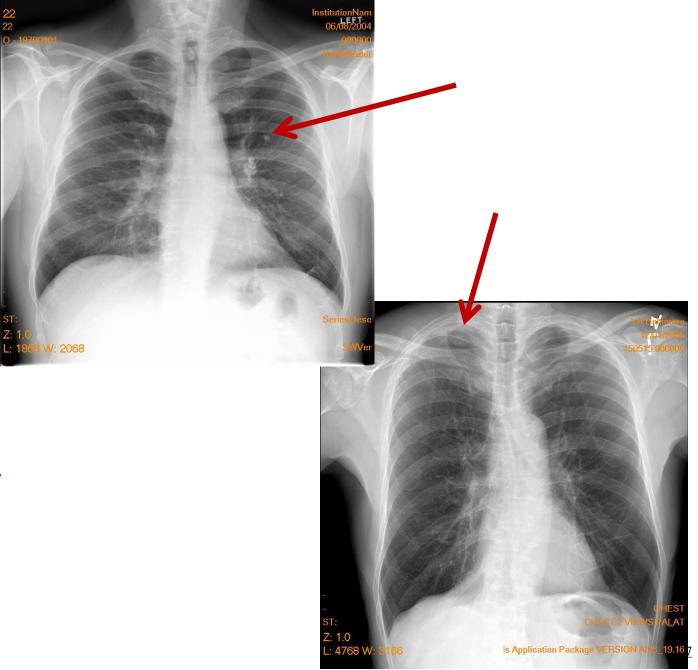
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- Data & study quality
 - Representative & appropriate sample
 - Large enough sample size
 - Appropriate control group
 - Randomization procedures in place
 - Proper blinding
 - Reader studies numbers, experience etc.
 - IRB, IACUC, HIPAA, COI
 - Could it be replicated

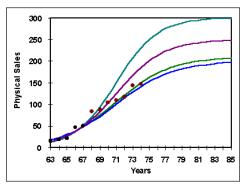








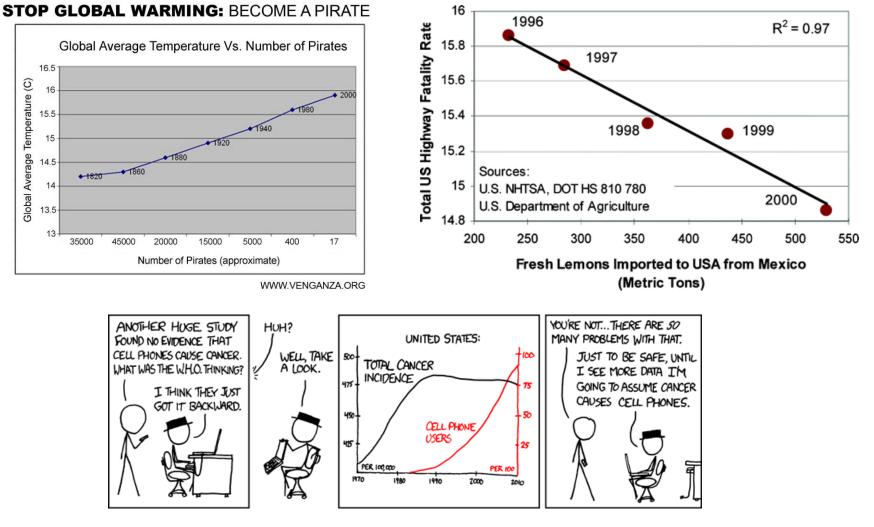
- Vital & pertinent data or information *left out*
 - Why are details left out
 - Would it affect results & conclusions
- Missing data
 - Why is it missing
 - How was it dealt with practically
 - How was it dealt with statistically
 - Would it affect results & conclusions





- How are the data *interpreted*
 - Correlation ≠ causation
 - P-values
 - Statistical power
 - Statistical vs clinical significance
 - In isolation or in context literature
 - Limitations noted



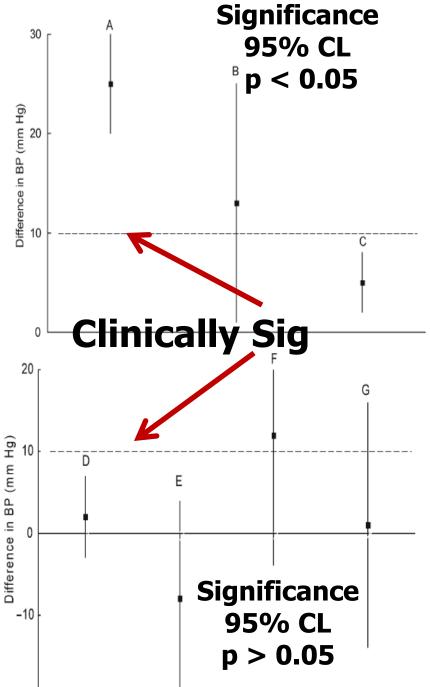




Confidence Limits

- Method for estimating population values based on what is known about sample values
- Diastolic BP = 88, s = 4.5, n = 72
- Sx = $4.5/\sqrt{72} = 0.53$
- 95% Upper = 88 + (1.96 x 0.53) = 89.04
- 95% Lower = 88 (1.96 x 0.53) = 86.96
- 5% probability range excludes population mean





<i>P</i> -value < 0.05	Is the difference between sample means clinically significant?		Interpretation	
	At the lower end of CI range	At the upper end of CI range	_	
Yes	Yes	Yes	A: There is a clinically important difference between the study groups	
Yes	No	Yes	B: Cannot reach a final conclusion—more data required	
Yes	No	No	C: There is a clinically unimportant difference between the sample groups	
No	No	No	D: There is no clinically important difference between the two groups	
No	Yes	No	E: Cannot reach a final conclusion—more data required	
No	No	Yes	F: Cannot reach a final conclusion—more data required	
No	Yes	Yes	G: Meaningless range of CI—more data required	

- Margins of error
- How margin error & CL interact with sample size
- To get same level precision (+/-3.2%) larger samples needed as CL increases
- If want to be certain that 95/100 times study repeated estimate will be +/- 3.2% need sample 950

	Confidence Level		
Sample	80%	90 %	95 %
	% Margin of Error + / -	% Margin of Error + / -	% Margin of Error + / -
100	6.4	8.3	9.8
150	5.3	6.7	8
200	4.5	5.8	6.9
250	4.1	5.2	6.2
300	3.7	4.8	5.7
350	3.4	4.4	5.2
400	3.2	4.1	4.9
450	3.0	3.9	4.6
500	2.9	3.7	4.4
550	2.7	3.5	4.2
600	2.6	3.4	4.0
650	2.5	3.2	3.8
700	2.4	3.1	3.7
750	2.3	3.0	3.6
800	2.3	2.9	3.5
850	2.2	2.8	3.4
900	2.1	2.7	3.3
950	2.1	2.7	3.2
1000	2.0	2.6	3.1





How Large is Enough?

- There will always be a difference
- You expect this by chance alone
- Step 1 = what difference is *clinically* or scientifically relevant?



THINGS GOT REALLY INTERESTING WHEN THE STATISTICIAN STARTED DOING WARD ROUNDS

- Statisticians can't help!
- Must be made on scientific or clinical grounds
- Typically define an *acceptable range* of treatment effects (difference in means)





Typical Effect Sizes

Test	Small	Medium	Large
T-test indep. Means	0.20	0.50	0.80
T-test correl. R	0.10	0.30	0.50
2 indep. R	0.10	0.30	0.50
Paired sign test	0.05	0.15	0.25
Indep. Prop. (z- test)	0.20	0.50	0.80
X ²	0.10	0.30	0.50
1-way ANOVA	0.10	0.25	0.40
Mult. Correl.	0.02	0.15	0.35





	True status Ho = True	True status Ho = False
Test = accept Ho	$\begin{array}{l} \textbf{Correct} \\ \textbf{p} = \textbf{1} - \alpha \end{array}$	Type II p = B
Test = reject Ho	Type I $\mathbf{p} = \alpha$	Correct p = 1 - B

Power(1 - B) = probability that test
significance will lead to correct reject null





What Affects Power

- 0.80 generally minimum acceptable
- Increases with sample size & lower population variability
- Increase by raising level significance (p 0.10 more powerful than p 0.05 = easier reject null)
- One-tailed > power two-tailed
- Larger effect size more power

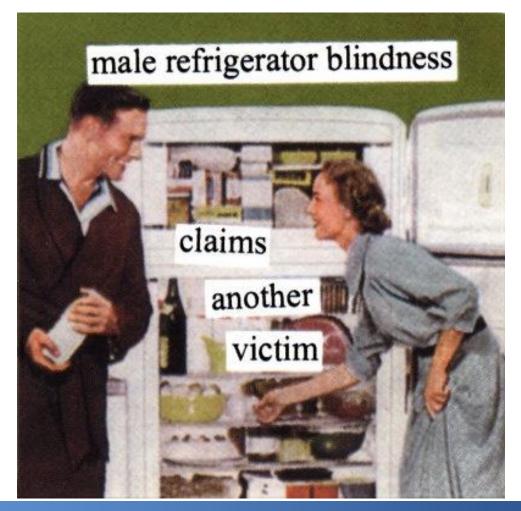
Typical Sample Sizes

Test (α = 0.05)	Small	Medium	Large
T-test indep. Means	393	64	26
T-test correl. R	783	85	28
2 indep. R	1573	177	66
Paired sign test	783	85	30
Indep. Prop. (z-test)	392	63	25
X ² for 1df/3df/5df	785/1090/1293	87/121/143	26/44/51
1-way ANOVA for 2/3 groups	393/322	64/52	26/21
Mult. Correl. For 2/3 variables	481/547	67/76	30/34



ANNI HII H

Power Analogy







NIN III K

The Fridge

- Is it there or not?
- Better If it really is there what is the probability would find it?
- How long spent looking? Longer = more likely find it
- How big is it? Gallon milk easier than a lime
- How messy is fridge? Messier less likely to find than organized

The Experiment

- Time = sample size, more data = more power
- Size = effect size, larger = more power
- Messiness = variability in data, lower = more power
- Use large sample with small sd & large effect size & get no significant difference can be confident in it

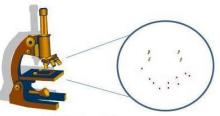


Validity Issues

- History, maturation, learning
- John Henry & Hawthorne Effects
- Experimental treatment diffusion
- Ecological validity
- Novelty & disruption effects
- Small samples, faulty randomization
- Intact groups
- Counterbalancing & memory



When No One is Watching



Add An Observer



Metrics of Performance







Definitions

- T = <u>Test</u> result (diagnosis)
 - -T+ = positive test (abnormal)
 - -T- = negative test (normal)
- D = <u>Disease</u> status (ground truth, gold standard)
 - D+ = patient actually has disease (abnormal, signal)
 - -D- = patient does not have disease (normal, noise)



Definitions

- TP = T+ | D+ ("hit")
- FN = T- | D+ ("miss")
- FP = T+ | D- ("false alarm")
- TN = T- | D-



Good & Bad Decisions

• TPF + FNF = 1 • TNF + FPF = 1 T + TPF FPF = TPF FPF TNF



Sensitivity & Specificity

 Sensitivity = fraction of diseased cases called diseased

Sensitivity =TP / (TP + FN)

 Specificity = fraction of non-diseased cases called normal

Specificity = TN / (TN + FP) = 1 - FPF



PPV & NPV

- PPV = fraction cases testing + that are diseased
 - -PPV = TP / (TP + FP)
- NPV = fraction cases testing that are not diseased

-NPV = TN / (TN + FN)

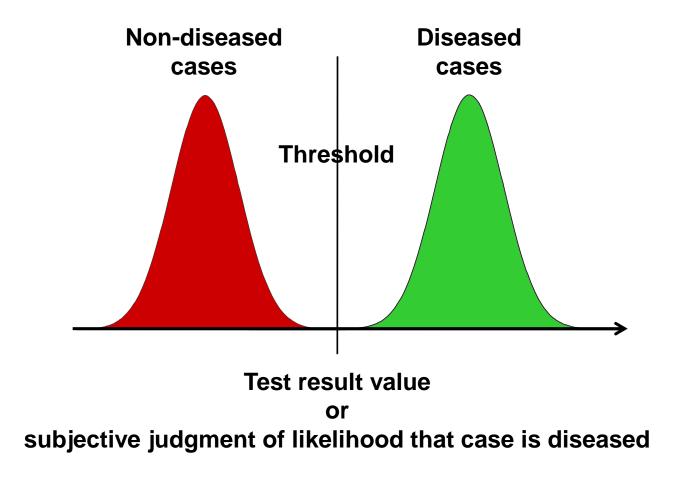




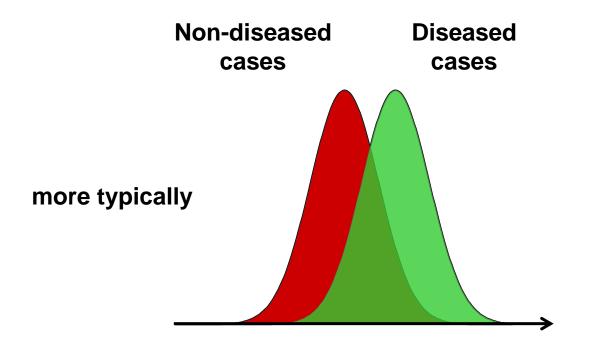
 Accuracy = sum correct outcomes divided by total number of tests done

$= \frac{TP + TN}{all tests} = \frac{TP + TN}{TP + TN + FP + FN}$



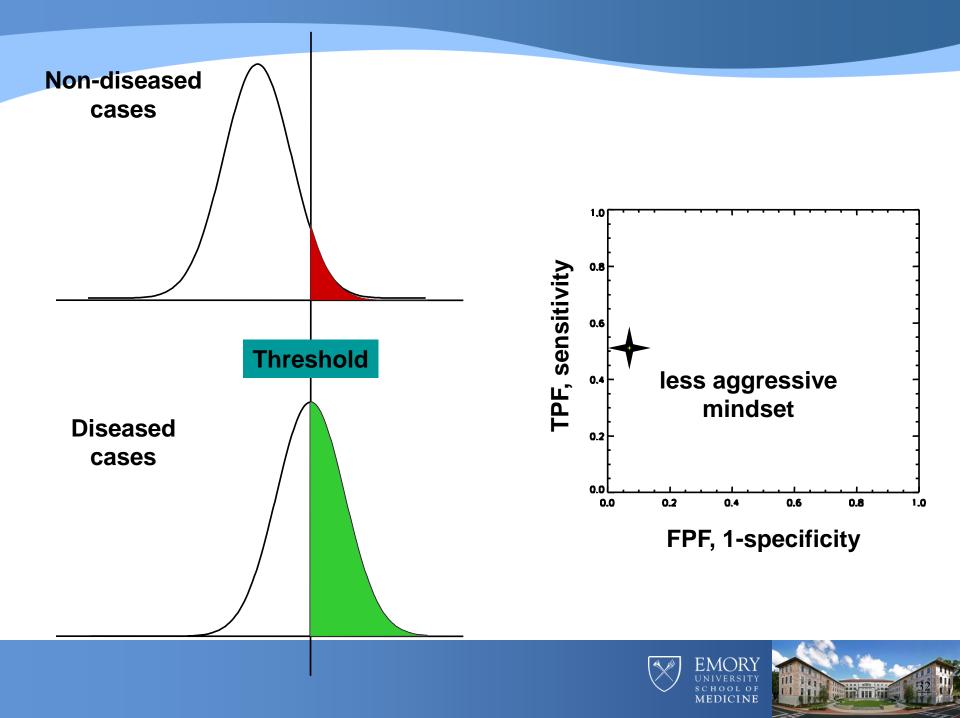


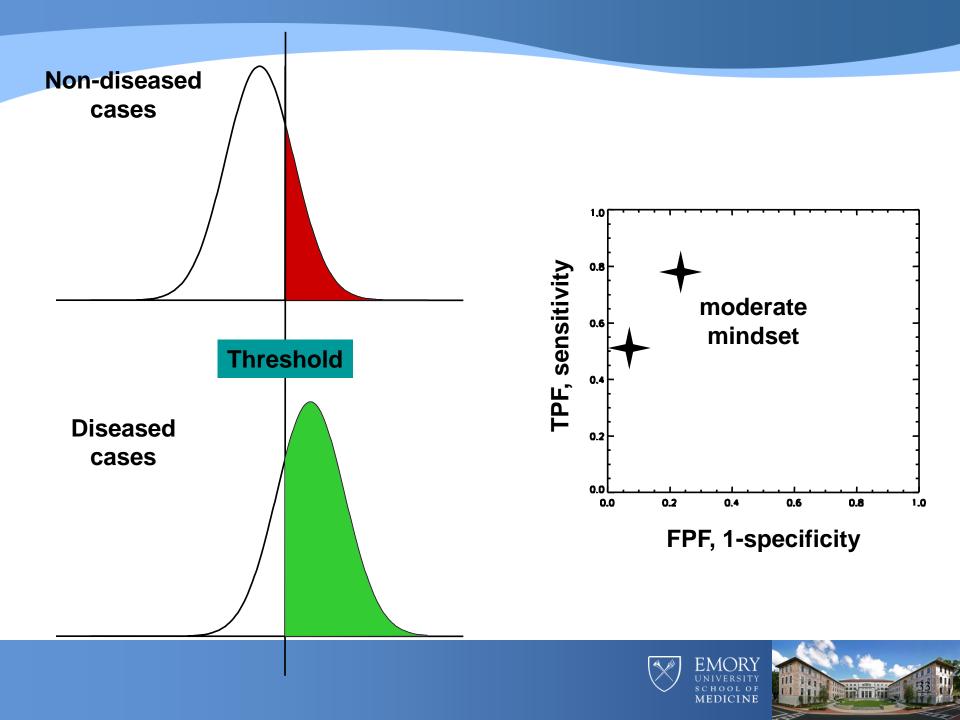


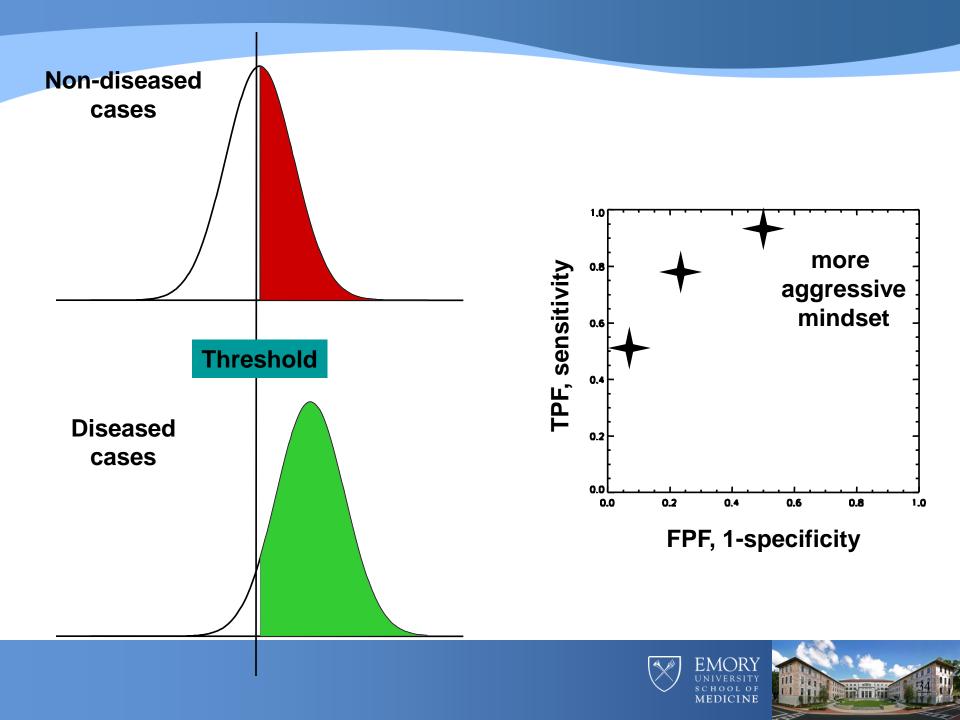


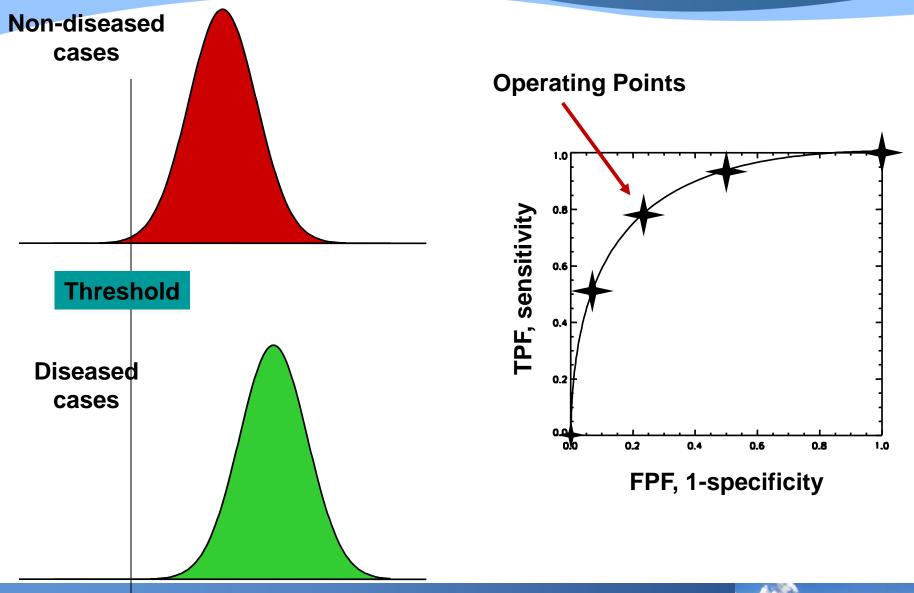
Test result value or subjective judgment of likelihood that case is diseased





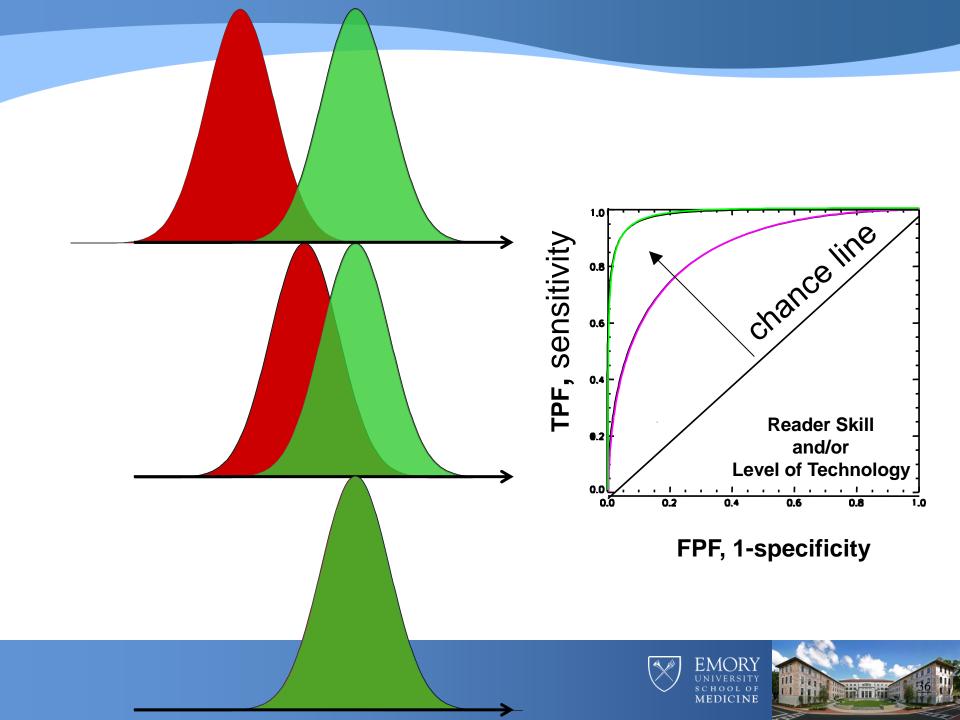












Summary

- Lots of good reviews papers available on various topics
- Lots of stats programs but not always good
 Same with graphing
- When in doubt ask!
- Am available to help with stats as needed!



Questions?

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