Chapter 14: Inference For Regression (Last Topic:) 14.1 Inference about the modal Day 1 OBJ: You will find a conf. int. for the true slope (B) of a LSRL. Look @ Bottom of pg 780 to Bottom of pg. 782. Y= a+bx & Slope b and yount are statistics - they are calculated from the sample Data. (Like X + B) To Do Inference we think of a and b as estimators of parameters of and B (Like u+p) Conditions for Regression Inference · Linearity (Check the scatter plot for signs of curva time) · Triclependence (Check the residual plot for randomness) · Consistency of Variance (Check Residual Plot to Be superior it is not for shoot of the serious of
Look @ Bottom of pg 780 to Bottom of pg 782. Q=a+bx & Slope b and unint are statistics - they are calculated from the sample Data (Like x + B) To Do Inference we think of a and b as estimators of parameters of and B (Like u+p) Conditions for Regression Inference Linearity (check the scatter plot for signs of curve time)
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· Independence (crock the residual plat for randomness) · Consistency of Variance / Chock Residual Plat to Re
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a construction of the cons
· Normality of Errors (Check the histogram of the residuals)
Inference
eThe first step is to estimate the unknown parameters
a, B, c.
· Calculate J= a+bx
· b of the LSRL is an unbiased estimator of the true
Slope P.
oa of the LSRL is an unbiased estimator of the true
Example 14.2 + 2 IP After 3 Residuals = 60s y- Exp. y
Standard Error About the LSRL
5= VE Residual = IN (1-7) } SON FS. (In Num. of So St. Day) N-2 Suse S to astimate the unknown of.
V n-2 I h-2 I use S to estimate the unknown 5.
df of s is n-2 (why n-2 instead of n-1?) blc how we are observing 2-variables!
DIC how we are observing d-variables!
Conf. Int. For B = b ± t* 5 Example 14.4
Conf. Int. For B = b ± ±* 5 [Example 14.4] [Example 14.1]

2007 & Ho: B=1 . The mean of y D's at the same rate as X

* Regression output (Statistical Software) Give t and Two sided p-value.

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Example 14.5

Example 14.6

HW: 14.6

OTher values For Ho: B=)

In other words that the slope is or is not zero.

I think it is easy to get lulled into that idea, especially b/c the model utility test is so ubiquitous and we rarely (it seems) have an occasion to test a non-zero slope. I think this rarity is probably due to a lack of knowledge of situations that might lead to hypothesis tests (a) other than zero, and (b) other than slope for regression.

Let me see if I can dredge something up...

Rooblem #1 Test Prep Company

Suppose I have a wiz-bang new method of test prep for the SAT, and I want to guarantee a 25 point increase across the board on the next test. Then I would think a reasonable model would be: NewScore = alpha + beta*OldScore + Error. My null hypotheses would seem to be: alpha = 25 AND beta = 1.

Problem:#2...The |cubic| scaling hypothesis/

Some biologist or other has told me that for ears of corn of a given hybrid, the volume is proportional to the cube of the length. (This is the standard assumption for organisms of the same species.)

OK, I reason that this means $V = kL^{(1/3)}$, and I do the appropriate transformation... Now I have:

LogV = alpha + (1/3)LogL + Error.

My null hypotheses would seem to be: alpha = 0 AND beta = 1/3.

Problem #3: Assortative mating \$

It has been commonly observed (at least among arthropod observers) that larger males tend to mate with larger females. Suppose that a theory is posited that males (being males, right ladies??) will pick females (or at least they THINK they are making the decision) smaller than themselves by 10%. Then, for the typical garden variety male Mexican Redknee tarantula (Brachypelma smithi) one would have a model similar to Problem #2 but w/o the transformation. For the male and female leg lengths, we would suppose the following model to be reasonable:

FLL = alpha + beta * MLL

My null hypotheses would seem to be: alpha = 0 AND beta = 0.90.