

this time: correlation
 &
 next time: regression

read: ch. 6 in
 lecture notes
 pp. L-214 - L-268

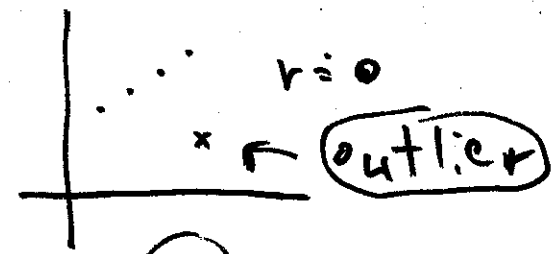
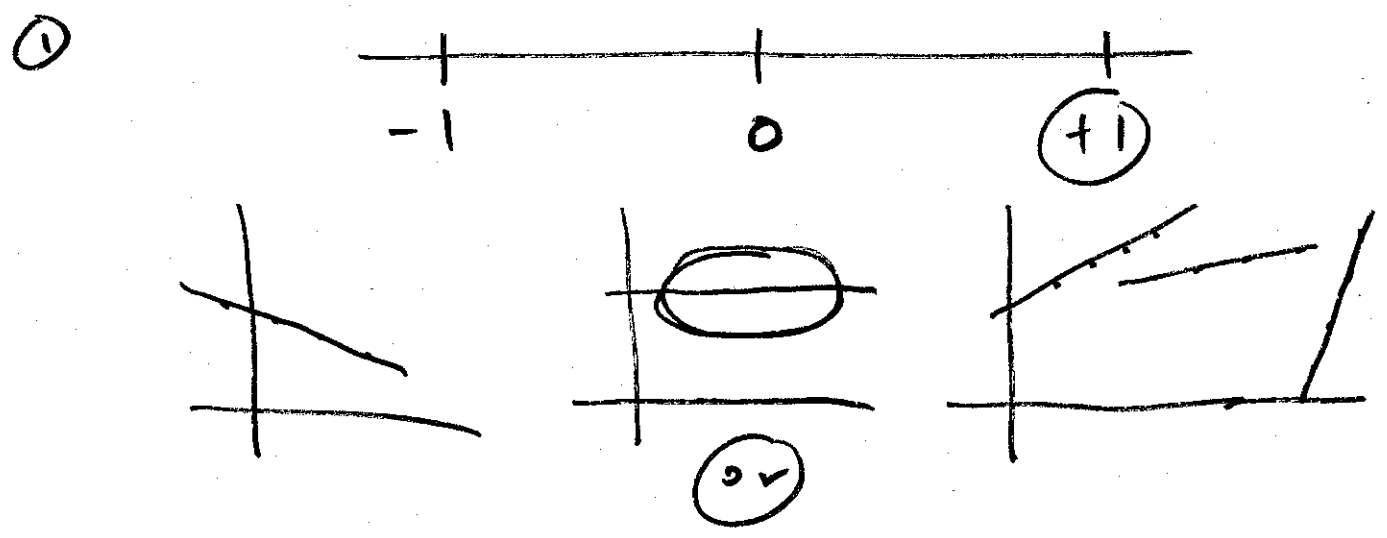
AM 57
 26 Feb
 09

lab 5 due by 5pm Fri 27 Feb

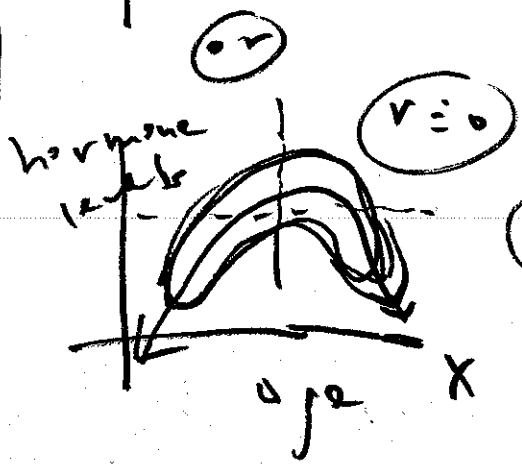
hwk 4 due Thu 5 Mar

correction: lab 6 due by 5pm Fri 6 Mar

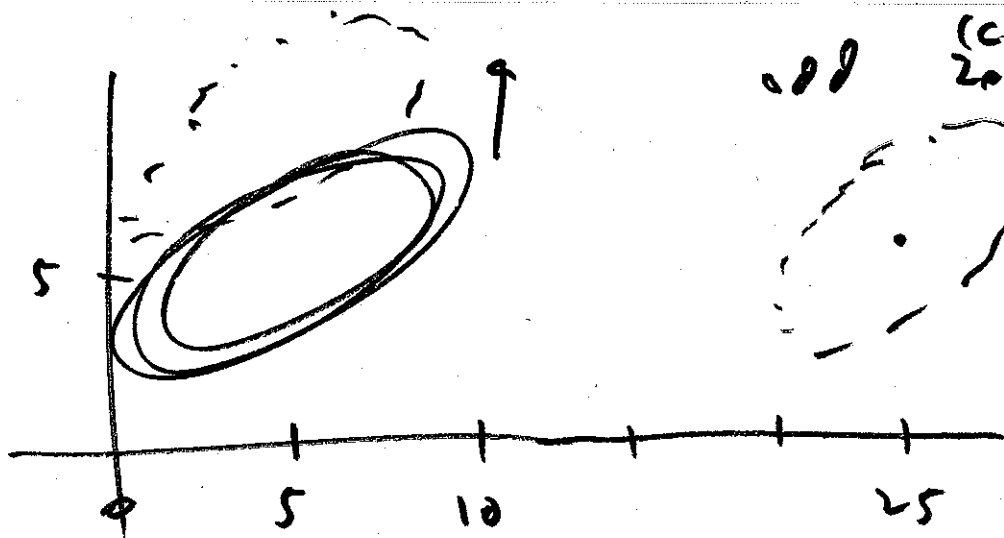
today: L-221 + L-2 facts about r



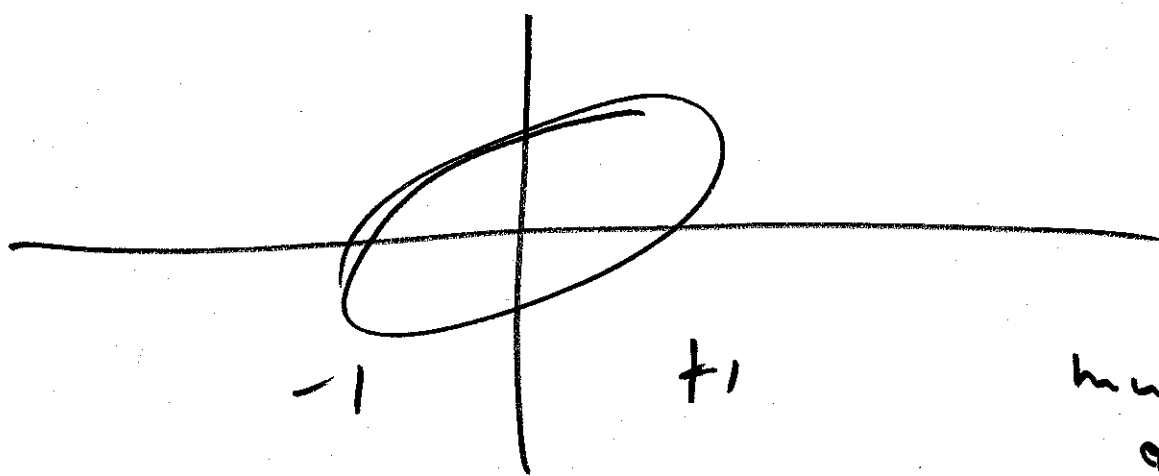
$$r = \frac{\sum_{i=1}^n (x_i - \bar{x}) (y_i - \bar{y})}{\sqrt{\sum_{i=1}^n (x_i - \bar{x})^2} \sqrt{\sum_{i=1}^n (y_i - \bar{y})^2}}$$



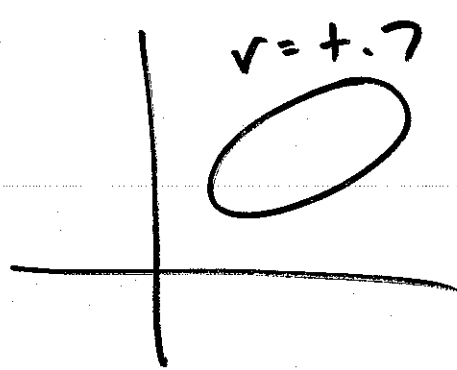
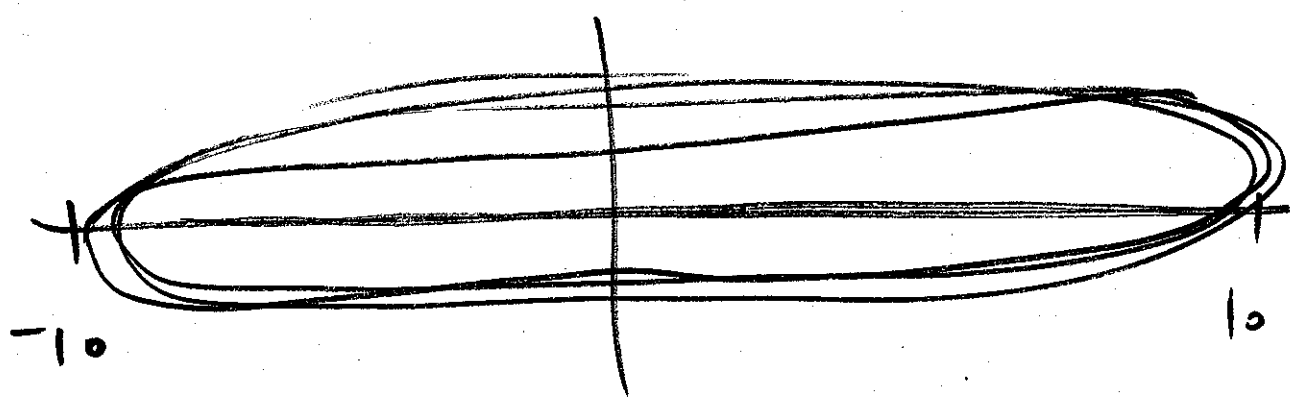
nonlinearity



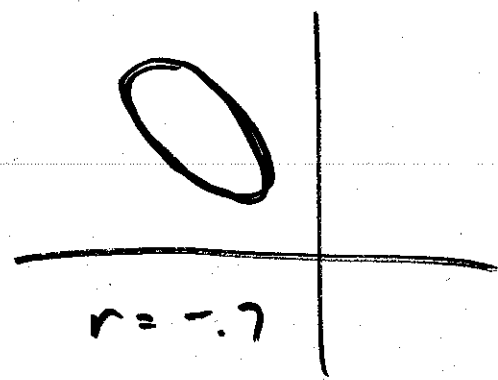
add (c) to all x -values
 leaving r unchanged

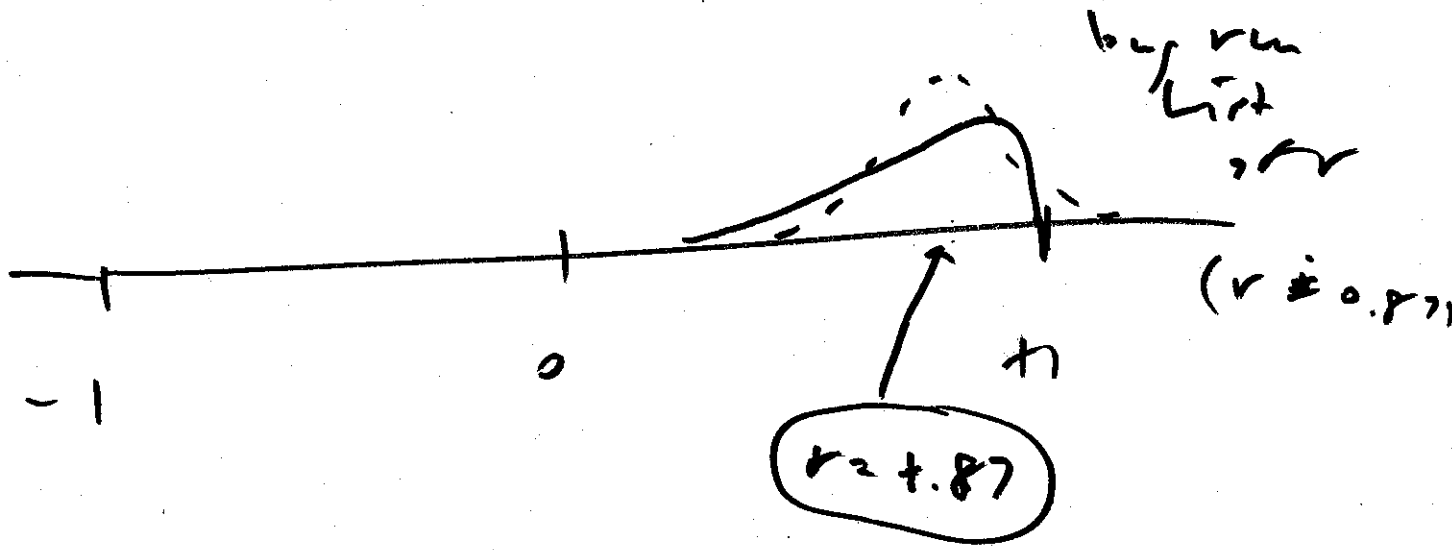
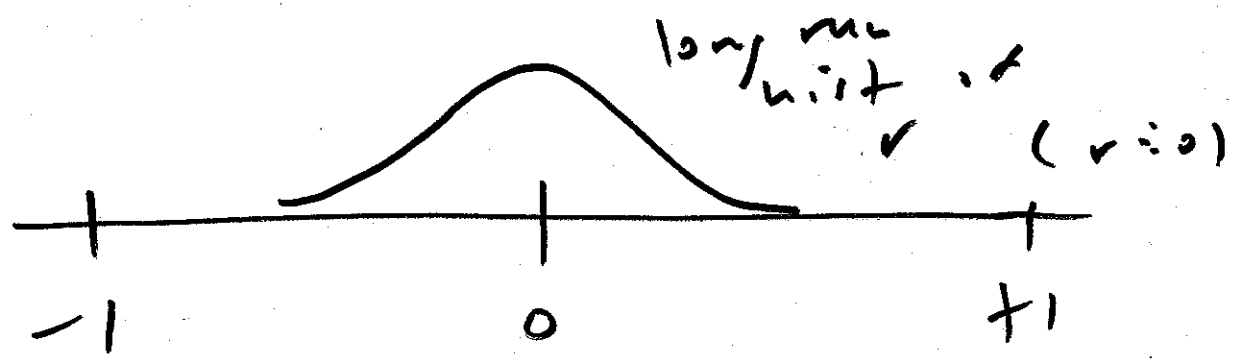


multiply all x -values by 10



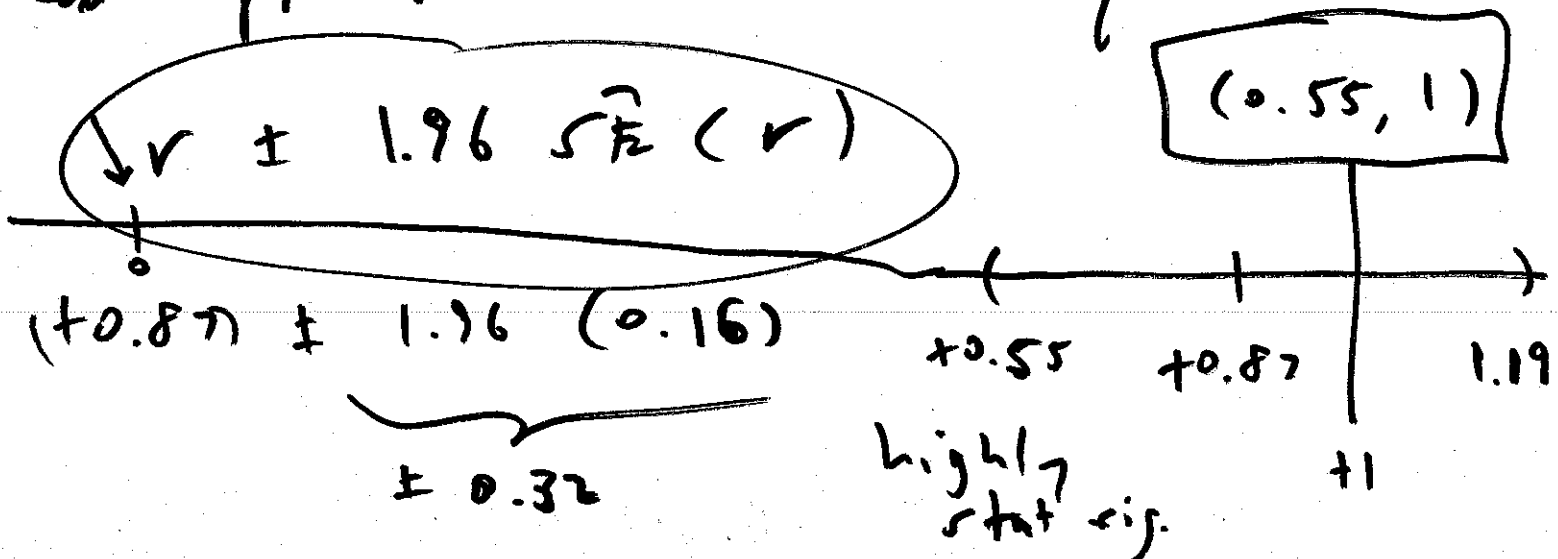
mult. x by -1



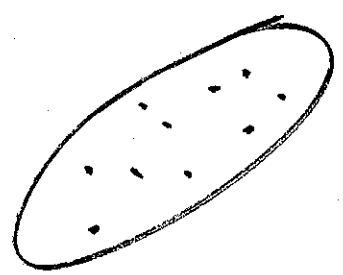


skip: $L-231 \rightarrow L-244$

approx 98% CI for ρ :



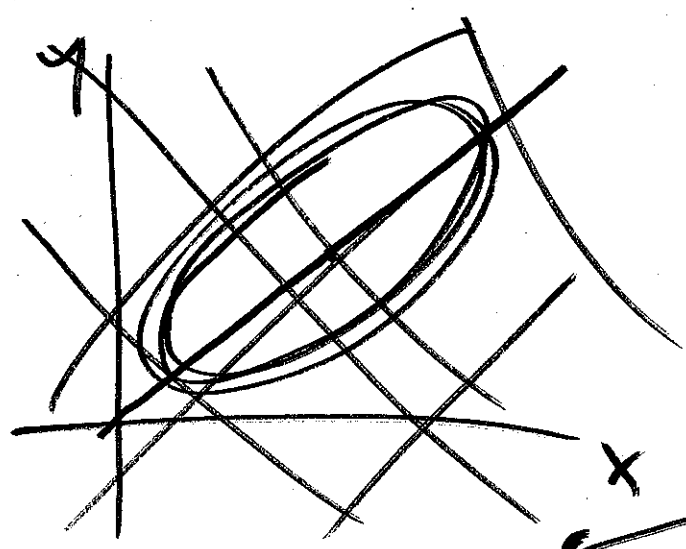
y
tail
length



wing length x

$r = +0.87$ q_1 : how
use x to predict y?

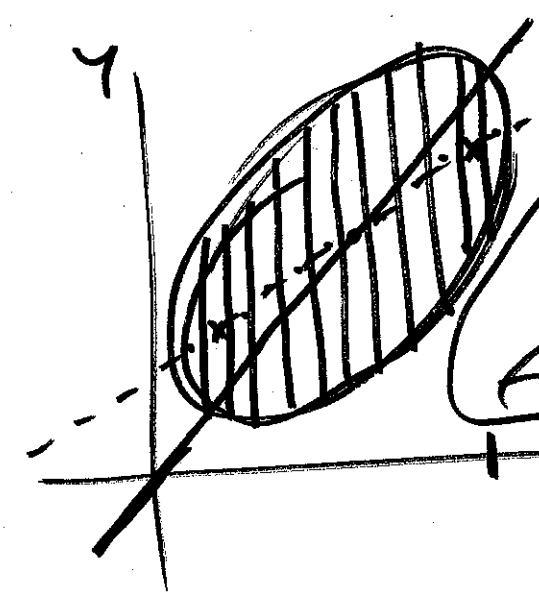
q_2 : what's the
equation of best
line for predicting
y from x?



(major
axis)

Galton
(1890s)

SD line:
slope $\frac{s_y}{s_x}$



regression
line
for predicting
y from x

slope $r \cdot \frac{s_y}{s_x}$

y-intercept

equation of
best line for predicting
y from x

$$\hat{y} = \hat{\beta}_0 + \hat{\beta}_1 x$$

interest! slope $\hat{\beta}_1 = r \cdot \frac{s_y}{s_x}$; line ⑤

go y through (\bar{x}, \bar{y}) :

$$\bar{y} = \hat{\beta}_0 + \hat{\beta}_1 \bar{x} \quad \text{so}$$

$$\hat{\beta}_0 = \bar{y} - \hat{\beta}_1 \bar{x} \quad \text{②}$$

