

1) Suppose a researcher desires to predict average female score from average male score from data on standardized science tests given to 8th graders in 39 countries. Assume that the correlation $r = 0.9889$. $= \hat{\rho}$, predict y from x

	variable	mean	standard deviation
y	female score	508.79	49.304
x	male score	525.38	49.525

a) Find the slope of the least square line.

$$\hat{\beta}_2 = \hat{\rho} \frac{s_y}{s_x} = 0.9889 \frac{49.304}{49.525} = 0.9845$$

b) Find the intercept of the least square line.

$$\hat{\beta}_1 = \bar{y} - \hat{\beta}_2 \bar{x} = 508.79 - 0.9845(525.38) = -8.4466$$

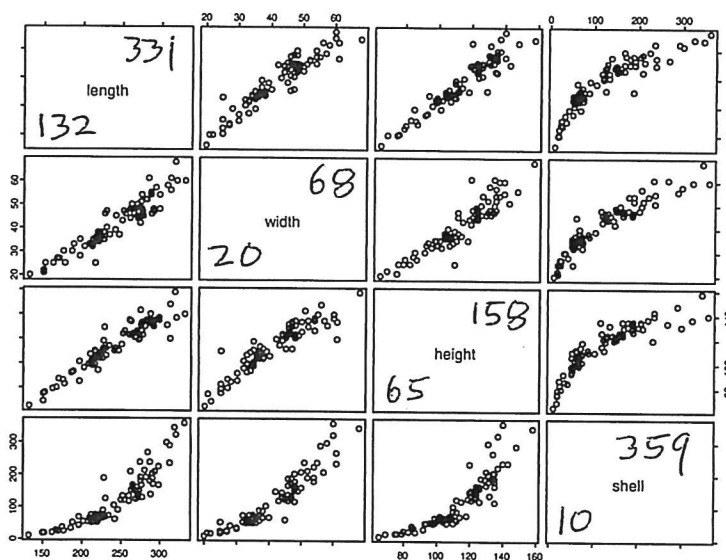


Figure 1: Scatterplot Matrix for Original Mussel Data Predictors

2) From the figure above, which transformations are suggested by the log rule?

1

$\log(\text{shell})$

since $\frac{359}{10} \approx 36$

3) From the figure below, which transformation should be used? Explain briefly.

$$y = \log(z)$$

since the plot is linear

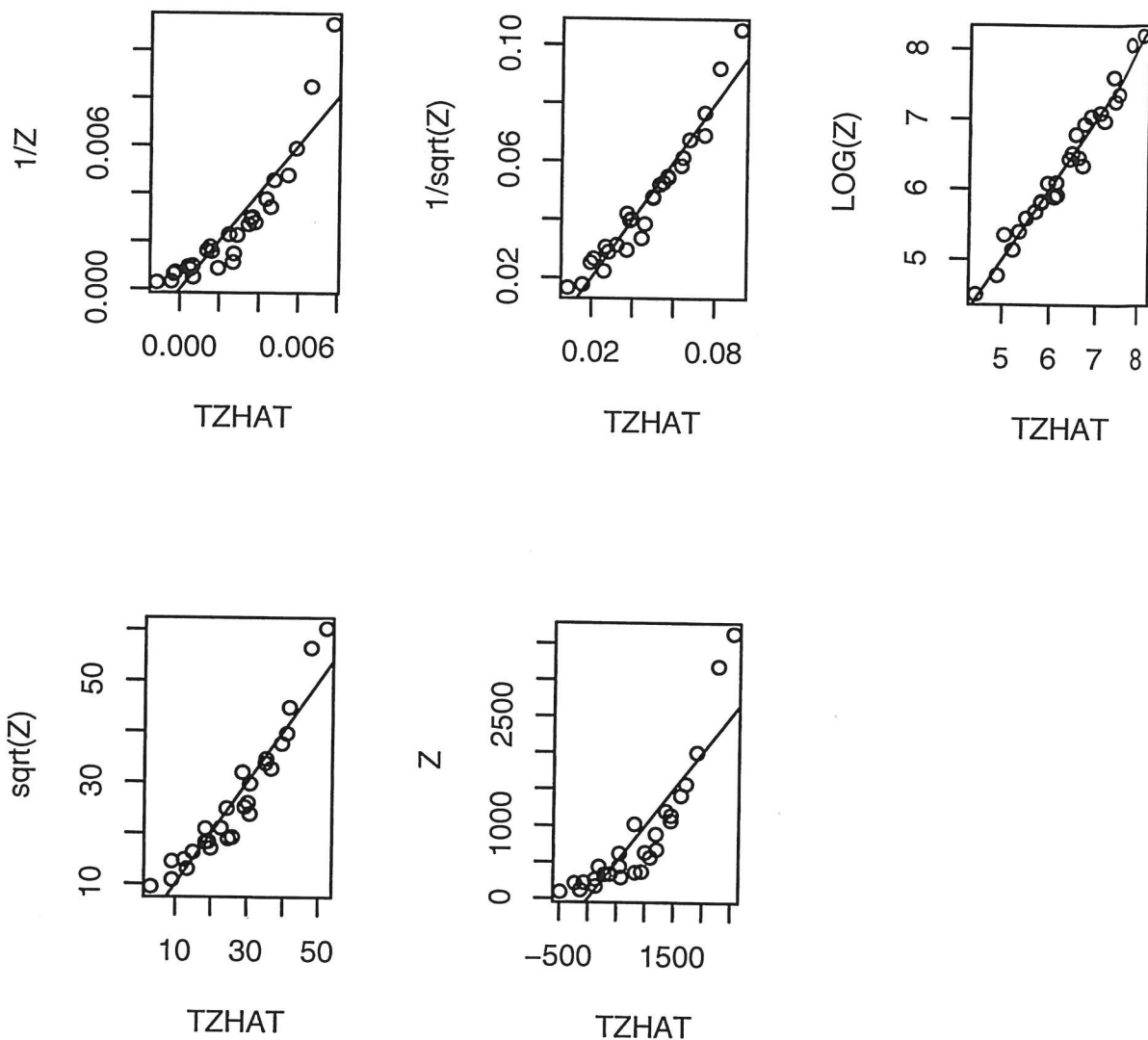


Figure 1:

Math 484 Exam 2 Fall 2016

	Estimate	Std.Err	t-value	Pr(> t)
X	0.9966	0.0016	626.6118	0

$X = \text{remembered score}$

Summary Analysis of Variance Table

Source	df	SS	MS	F	p-value
Regression	1			39264.2	0.0000
Residual	19				

4) Dr. Olive graded 2016 quiz 5, then entered the scores on the computer. He tried to remember the score x_i and then entered the actual score Y for the 20 Math 484 students. Regression through the origin was used. Assume least squares output can be used.

a) Predict Y if $x_i = 90$.

$$\hat{Y} = \hat{\beta}_1 X = 0.9966(90) = \boxed{89.694}$$

b) Perform the t test for $\beta_1 = 0$.

$$H_0 \beta_1 = 0 \quad H_A \beta_1 \neq 0$$

$$t_0 = 626.6118$$

$$p\text{-val} = 0$$

reject H_0 $X = \text{remembered score}$ is needed in the MLR model for $Y = \text{actual score}$

c) Perform the no intercept Anova F test.

$$H_0 \beta_1 = 0 \quad H_A \beta_1 \neq 0$$

$$F_0 = 39264.2$$

$$p\text{-val} = 0$$

reject H_0 there is an MLR relationship

between $Y = \text{actual score}$ and

$X = \text{remembered score}$

-2 \sum 34 \rightarrow 30 2.042 (500.57, 524.11)
(439.90, 584.71)

Label	Estimate	Std. Error	t-value	p-value
Constant	445.862	26.3328	16.932	0.0000
edaids	0.787074	0.287856	2.734	0.0099
fteach	0.578163	0.330786	1.748	0.0895

Prediction = 512.341, se(pred) = 35.4772, se = 5.76336

Q4) 22 5) Let Y = mean science score of female 8th graders in 37 countries. The predictors are $edaids$ = percent of 8th graders with dictionary, study table and computer and $fteach$ = percentage of 8th graders taught by female teachers. Suppose that it is desired to predict Y_f if $edaids = 47$ and $fteach = 51$, so that $x_f = (1, 47, 51)^T$. Assume that $\hat{Y}_f = 512.341$, $se(\hat{Y}_f) = 5.76336$ and that $se(pred) = 35.4772$.

a) If $x_f = (1, 47, 51)^T$ find a 95% confidence interval for $E(Y_f|x_f)$.

$\hat{Y}_f \pm t_{n-p, 1-\frac{\alpha}{2}} SE(\hat{Y}_f) = 512.341 \pm 1.96 (5.76336)$

$= 512.341 \pm 11.296 = [501.045, 523.637]$

b) If $x_f = (1, 47, 51)^T$ find a 95% prediction interval for Y_f .

$\hat{Y}_f \pm t_{n-p, 1-\frac{\alpha}{2}} SE(pred) = 512.341 \pm 1.96 (35.4772)$

$= 512.341 \pm 69.535 = [442.806, 581.876]$

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	L1	L2	L3	L4
# of predictors	13	6	5	4
# with $0.01 \leq p\text{-value} \leq 0.05$	0	0	0	0
# with $p\text{-value} > 0.05$	9	2	1	2
R_f^2	0.993	0.991	0.990	0.981
$\text{corr}(\hat{Y}, \hat{Y}_I)$	1.0	0.9990	0.9987	0.9948
$C_p(I)$	13.0	4.70	5.24	27.21
\sqrt{MSE}	4.024	3.933	4.040	5.322
p-value for partial F test	1.0	0.985	0.444	0.007

6) The above table gives summary statistics for 4 MLR models considered as final submodels after performing variable selection. The data set had 35 cases and assume that the full model L1 was good even though n is small. Which model should be used as the final submodel? Explain briefly why each of the other 3 submodels should not be used.

$L3 = I_X$ should be used $5.24 < 4.7 + 1 = 5.7$

L1 and L2 have too many predictors

L4 has $C_p > 2k$ and pval for partial F test is too small

7) Suppose that the regression model is $Y_i = \beta + 7X_i + \epsilon_i$ for $i = 1, \dots, n$ where the ϵ_i are iid $N(0, \sigma^2)$ random variables. The least squares criterion is $Q(\beta) = \sum_{i=1}^n (Y_i - \beta - 7X_i)^2$.

a) What is $E(Y_i)$?

$$\beta + 7x_i$$

b) Find the least squares estimator $\hat{\beta}$ of β by setting the first derivative $\frac{d}{d\beta} Q(\beta)$ equal to zero.

$$\frac{d}{d\eta} Q(\eta) = -2 \sum (y_i - \eta - 7x_i) \stackrel{\text{set}}{=} 0$$

$$\text{or } \sum (y_i - 7x_i) = n\eta$$

$$\text{so } \hat{\beta} = \frac{\sum (y_i - 7x_i)}{n} = \bar{y} - 7\bar{x}$$

c) Show that your $\hat{\beta}$ is the global minimizer of the least squares criterion Q by showing that the second derivative $\frac{d^2}{d\beta^2} Q(\beta) > 0$ for all values of β .

$$\frac{d^2}{d\eta^2} Q(\eta) = \frac{d}{d\eta} [-2 \sum (y_i - \eta - 7x_i)]$$

$$= \frac{d}{d\eta} [-2 \sum (y_i - 7x_i) + 2n\eta]$$

$$= 2n > 0$$

Current terms: (brate gnp idrate mlife)

	df	RSS	k	C_I
Delete: gnp	93	242.17	4	2.677
Delete: brate	93	280.647	4	17.243
Delete: idrate	93	315.003	4	30.248
Delete: mlife	93	735.697	4	189.508

8) The above output for Rouncefield (1995) data is for predicting $Y = \text{female life expectancy}$ from $x_2 = \text{brate}$ (birth rate), $x_3 = \text{drate}$ (death rate), $x_4 = \text{GNP}$, $x_5 = \text{idrate}$ (infant death rate), and $x_6 = \text{mlife}$ (male life expectancy). What is the I_{min} model that has the smallest $C_p(I)$? Do not forget the constant.

constant, brate, idrate, mlife

9) For the figure below, consider the ladder rule for making a transformation $t_\lambda(w)$ to increase the linearity of the plot.

a) If $w = Y$, should λ be made larger or smaller?

smaller

b) If $w = X$, should λ be made larger or smaller?

smaller

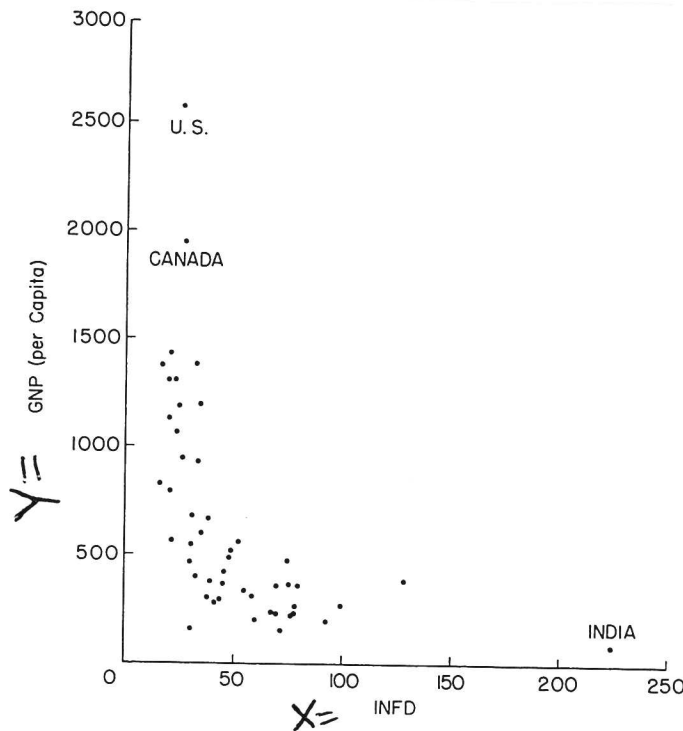


Figure 2.4 Gross national product (GNP) per capita by infant death rate (INFD) in 49 countries in the world.

Gungtar and Mason (1980, p.31)