

Question 1

- Best known equations for $HR_{max} = (220 - \text{age})$ bpm
- Experiment to ascertain validity of the $(220 - \text{age})$ equation
- Healthy volunteers aged 18-80 were recruited and had HR_{max} determined by exercise phys.

$N = \text{sample size}$

$A_i = \text{age of } i\text{th subject}$

$H_i = HR_{max}$

$$N = 40$$

$$\sum A_i = 1880$$

$$\sum A_i^2 = 101,374$$

$$\sum H_i = 7040$$

$$\sum H_i^2 = 1,249,420$$

$$\sum A_i \cdot H_i = 322,284$$

$$\hat{\sigma}^2 = S^2_{yx} = 123.74 \left(\frac{SSE}{df_E} \right)$$

a) Write down a linear model relating HR to age.

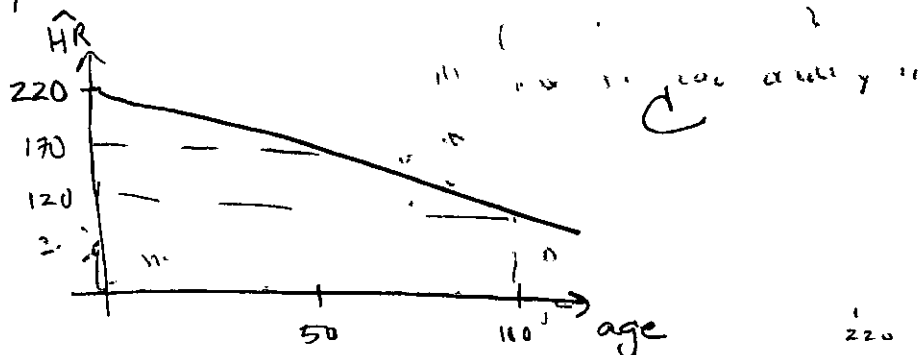
$$E(HR_{max}) = \text{intercept} + \beta_1 \text{ age} \quad \leftarrow \text{age in years}$$

\uparrow predicted HR based on age

$$= \beta_0 + \beta_1 \text{ age}$$

when age = 0 \leftarrow drop in HR_{max} w/ each addl year
 (birth)

b) Assuming $(220 - \text{age})$ is correct, state values of the parameters in your linear model.



c) fit model to data \Rightarrow assuming model from A

$$\hat{HR} = \beta_0 + \beta_1 \text{ age}$$

$$X = \begin{pmatrix} 1 & a_1 \\ \vdots & \vdots \\ 1 & a_{40} \end{pmatrix}$$

$$(X'X) = \begin{pmatrix} 1 & \dots & 1 \\ a_1 & \dots & a_{40} \end{pmatrix}$$

$$Xy = \begin{pmatrix} 1 & \dots & 1 \\ a_1 & \dots & a_{40} \end{pmatrix} \begin{pmatrix} hr \\ \vdots \\ hr \end{pmatrix}$$

$$\hat{\beta} = (X'X)^{-1} X'y$$

$$= \begin{pmatrix} 40 & \sum a_i \\ \sum a_i & \sum a_i^2 \end{pmatrix}^{-1}$$

$$= \begin{pmatrix} \sum hr \\ \sum a_i \cdot hr \end{pmatrix}$$

$$\hat{\beta} = \begin{pmatrix} 207.0444 \\ -0.6605 \end{pmatrix} = \begin{pmatrix} \hat{\beta}_0 \\ \hat{\beta}_1 \end{pmatrix}$$

$$\hat{\beta}_1 = \frac{\sum a_i H_i - N \bar{A} \bar{H}}{\sum a_i^2 - N \bar{A}^2}$$

d) Test whether data are consistent
w/ $\beta_1 = -1$ (Δ implied w/ formula)

$$\hat{\beta}_0 = \bar{H} - \beta_1 \bar{A}$$

$$H_0: \beta_1 = -1$$

$$C = (0 \ 1) \quad \theta_0 = (-1)$$

$$\hat{\theta} = -0.6605$$

$$\frac{\{(\hat{\theta} - \theta_0)' M^{-1} (\hat{\theta} - \theta_0)\} / 1}{MSE} \sim F_{1, 38}$$

or

$$\frac{\beta_1 - -1}{\sqrt{\text{Var} \beta_1}} =$$

} either way, test
stat is large so
reject the null.
Inferent from formula.

50 yr old $\rightarrow HR_{max} = 160$
different from expectation?

$$\text{expected} = 207.044 - 0.6605(50) \\ = 167.414$$

$$H_0: \hat{\beta}_0 + \hat{\beta}_1(50) = 160$$

$$C = (1 \ 50)$$

$$H_0: HR_{50} = 160$$

$$\hat{H} = 167.414$$

← craigs is wrong...

$$t = \frac{160 - 167.414}{\sqrt{\dots}}$$

← don't like this way.

$$C = (1 \ 50) \quad \theta_0 = 160$$

$$C\hat{\beta} = \hat{\theta} = 167.414$$

$F = 11.69$ 1,50 reject, it is diff

f) add interaction

$$g) \text{ new } \sum A_i = 1880 + 47$$

$$\sum A_i^2 = 101374 + 47^2$$

$$\sum A \cdot H = 322284 + (47 \cdot 196)$$

$$\sum H = 7040 + 196$$

then recalc as in c)

for the others, do it and then explain.

d) $E\hat{y} = \beta_0 + \beta_1 \text{age} + \beta_2 \text{educ}$ $\hat{\sigma}^2 = \text{MSE} = \frac{\text{SSS}}{\text{df}_E} = .5159$

$(X'X)^{-1} =$ $\begin{pmatrix} & & \\ & & \\ & & \end{pmatrix}_{3 \times 3}$

$H_0: \beta_0 = \beta_1 = \beta_2$ using GLH approach $\rightarrow \theta = C\beta = \theta_0$

$C = \begin{pmatrix} 1 & -1 & 0 \\ 1 & 0 & -1 \end{pmatrix}$ $\theta_0 = \begin{pmatrix} 0 \\ 0 \end{pmatrix}$ $\beta_0 - \beta_1 = 0$
 $\beta_1 - \beta_2 = 0$

don't have $\vec{\beta}$

$\hat{\theta} = C\hat{\beta} =$

e) $\text{corr}(\hat{\beta}_1, \hat{\beta}_2) = \frac{\text{cov}(\hat{\beta}_1, \hat{\beta}_2)}{\sqrt{\text{var}\hat{\beta}_1, \text{var}\hat{\beta}_2}}$ cell (1,2) of $((X'X)^{-1} \sigma^2)$
 \uparrow cell 11 \uparrow cell 22

same approach
diff. #'s

f) the x's don't need to be normal.

The normality assumption of HILS Gauss only applies to the conditional distributions and error

Question 3 Treatment efficacy

$$= \beta_0 + \text{trt} + \text{age} + \text{gender} + \text{trt} \times \text{age} + \text{trt} \times \text{gender}$$

n=100, 25 per gender / trt combo (2 genders, 2 treatments) med or counsel

a) Write the relation b/w age and trt efficacy for each combo

TRT	Gender	Relation
med	female	$\beta_0 + \beta_{\text{age}}$
med	male	$\beta_0 + \beta_{\text{age}} + \beta_{\text{gender}}$
counsel	female	$\beta_0 + \beta_{\text{trt}} + \beta_{\text{age}} + \beta_{\text{trt} \times \text{age}}$
counsel	male	$\beta_0 + \beta_{\text{trt}} + \beta_{\text{age}} + \beta_{\text{gender}} + \beta_{\text{trt} \times \text{age}} + \beta_{\text{trt} \times \text{gender}}$

b) Reg. coeffs obtained.

Interpret coeffs:

ME = Δ in efficacy for (non-ref or 1 unit increase) in x
G1/G2 INT = add effect of an increase in age / being male for those on trt=1 (add effect of BOTH)

To test whether there are gender diff in trt eff given they are the same age, we should test whether

$H_0: \beta_{\text{trt} \times \text{gender}} = 0$ $C = (0 \ 0 \ 0 \ 0 \ 0 \ 1)$ $\theta_0 = 0$ $\hat{\theta} = C\hat{\beta}$

$F = \frac{(\hat{\theta} - \theta_0)^2}{MSE}$ or $\frac{\beta_6}{\sqrt{\text{Var}(\beta_6)}} \sim t_{94}$

$F \sim 1, 100 - 6 = 94$

c) added last. (Wald test of coeff=0)

non-sign terms are trt x age, age (alone), and trt.
probably want to keep trt, given trt x gender is sign.

d) Fill in ANCOVA table

Source	df	Type I	MS	F / Type I / MSE	P
trt	1		45.72234	48.5824	< 0...
age	1		0.5817	0.6181	0.43
gender	1		167.163	113.8701	< 0..
trt x age	1		0.3369	0.3580 = $\frac{MS}{MSE}$	0.55
trt x gender	1	13.649	13.649	12.845056	
error	94	88.46328	$MSE = \frac{SSE}{94}$ = 0.9410987	\uparrow t ² from b) last added in order = added last	
total	99				

F-stat, and df, for testing

$$E(y) = \beta_0 + \beta_1 \text{trt} + \beta_2 \text{age} + \beta_3 \text{gender} + \beta_4 \text{trt} \times \text{age} + \beta_5 \text{trt} \times \text{g}$$

vs.

$$E(y) = \beta_0 + \beta_1 \text{trt} + \beta_2 \text{age} + \beta_3 \text{gender}$$

cray SS = $\frac{(SSH) - (SS\beta_4 + SS\beta_5)/2}{2}$

$$\Rightarrow H_0: \beta_4 = \beta_5 = 0$$

$$C = \begin{pmatrix} 0 & 0 & 0 & 0 & 1 & 0 \\ 0 & 0 & 0 & 0 & 0 & 1 \end{pmatrix} \quad \theta_0 = \begin{pmatrix} 0 \\ 0 \end{pmatrix}$$

$$\hat{\theta} = C\hat{\beta} = \begin{pmatrix} -0.004731 \\ 1.529518 \end{pmatrix}$$

or use SSH of β_4 and β_5 w/ MSE etc. testing.

$$F = \frac{(\hat{\theta} - \theta_0)' M^{-1} (\hat{\theta} - \theta_0) / df_H}{MSE} = \text{etc.}$$

~ 2.9

where

$$M^{-1} = C(X'X)^{-1}C'$$

μ_m = mean med when $g=0$ age=33

μ_c = mean consel when $g=0$ age=33

$H_0: \mu_m = \mu_c$

$$\beta_0 + \beta_2(33) = \beta_0 + \beta_1 + \beta_2(33) + \beta_4(33)$$

$$\Rightarrow \beta_2 + \beta_4(33) = 0$$

$$C = (0 \ 1 \ 0 \ 0 \ 33 \ 0)$$

$$\theta_0 = 0$$

$$F \sim 1, 94$$

\perp age when age=0 (?)

(4) current smoking status, Y/N cough that day

a) Two separate logistic reg. models are specified. Provide the algebraic expressions for each as well as interpretations.

i. $\text{logit}(\text{cough}) = \beta_0 + \beta_{\text{smoke}} \text{smoke}$ the model with β_{smoke} is
 $\beta_0 = \log \text{ odds for having a cough (if non-smoker)}$
 $\beta_{\text{smoke}} = 0.5415$ (for every 1 unit increase in smoking freq, the log odds of having a cough increases by 0.5415 \hookrightarrow treating as linear)

ii $\text{logit}(\text{cough}) = \beta_0 + \beta_{\text{occas.}} + \beta_{\text{regular}}$
 $\beta_0 = \log \text{ odds for coughing when } x_1 = x_2 = 0 \text{ (non-smoker)}$
 $\beta_{\text{occ.}} = \log \text{ OR for having a cough if an occas smoker compared w/ a non-smoker}$
 $\beta_{\text{reg.}} = \text{ " " regular smoker --}$

b) Interpret the model w/ smoke as a covariate but without an intercept term. (this model wasn't run... correct?)

$$\text{logit}\left(\frac{Y}{n}\right) = \beta_1 (\text{smoke})$$

increase in prob. of smoking assuming no coughing among non-smokers.

- c) adv interval \rightarrow easier to understand (simpler)
 disadv interval \rightarrow assumes the jump from each level of smoking is =
- adv cat \rightarrow allows the effect of each successive 'level' of coughing to differ
 disadv cat \rightarrow might be harder to interpret / use
- ha, almost exactly the same as Craig

d) estimate $P(\text{coughing})$ for occasional smokers

$$i. P(\text{cough}) = \frac{\exp(\beta_0 + \beta_1(1))}{1 + \exp(\beta_0 + \beta_1(1))} = 0.298615$$

$$ii. P(\text{cough}) = \frac{\exp(\beta_0 + \beta_1)}{1 + \exp(\beta_0 + \beta_1)} = 0.287901$$

e) Pick preferred model, with 2-3 sentences summarizing