

# Optional Assignment Week 9

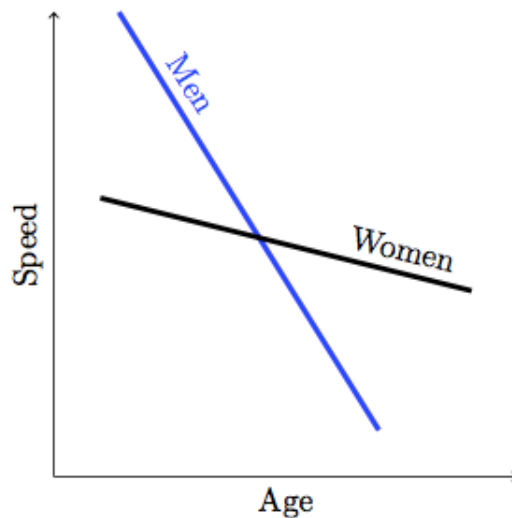
PIED 2711 | Analysing Data | 2018

November 26, 2018

## Question 1

The figure below shows a hypothesized (i.e. predicted) relationship between how fast a person runs, the person's age, and sex. Consider the model formula:

$$\widehat{\text{Speed}}_i = \alpha + \beta_1 \text{Age}_i + \beta_2 \text{Female}_i + \beta_3 \text{Age}_i \times \text{Female}_i$$



1. According to the prediction, what do we expect the sign of the coefficient on  $\text{Age}_i$  to be?
  - (a) Negative
  - (b) Zero
  - (c) Positive
  - (d) No way to tell from the information given

2. According to the prediction, what do we expect the sign of the coefficient on  $\text{Female}_i$  to be?
  - (a) Negative
  - (b) Zero
  - (c) Positive
  - (d) No way to tell from the information given
  
3. According to the prediction, what do we expect the sign of the coefficient on the interaction term ( $\text{Age}_i \times \text{Female}_i$ ) to be?
  - (a) Negative
  - (b) Zero
  - (c) Positive
  - (d) No way to tell from the information given

## Question 2

A researcher has data from the 67 Florida counties on  $y$  = crime rate (number per residents),  $x_1$  = median income (thousands of dollars), and  $x_2$  = percentage in urban environment.

1. The first table shows the results when she regressed  $y$  on  $x_1$ . Write down the model equation from the output and interpret the slope coefficient.
2. The second table shows the results when she regressed  $y$  on  $x_1$  and  $x_2$ . Write down the model equation from the output and interpret the slope coefficients.
3. Take a look at the correlation matrix. Use these correlations to explain why the  $x_1$  effect seems so different in the first model and second model.
4. Write down the model equations relating crime rate to income at urbanization levels of 1) 0, 2) 50, and 3) 100. Interpret.

```
. reg Crime Income, noheader
```

Crime	Coef.	Std. Err.	t	P> t	[95% Conf. Interval]	
Income	<b>2.611494</b>	<b>.6728733</b>	<b>3.88</b>	<b>0.000</b>	<b>1.267673</b>	<b>3.955315</b>
_cons	<b>-11.6059</b>	<b>16.78631</b>	<b>-0.69</b>	<b>0.492</b>	<b>-45.13048</b>	<b>21.91867</b>

```
. reg Crime Income Urban, noheader
```

Crime	Coef.	Std. Err.	t	P> t	[95% Conf. Interval]	
Income	<b>-.7905853</b>	<b>.8049117</b>	<b>-0.98</b>	<b>0.330</b>	<b>-2.398581</b>	<b>.8174107</b>
Urban	<b>.6418376</b>	<b>.1109602</b>	<b>5.78</b>	<b>0.000</b>	<b>.4201692</b>	<b>.863506</b>
_cons	<b>39.97226</b>	<b>16.35362</b>	<b>2.44</b>	<b>0.017</b>	<b>7.302154</b>	<b>72.64238</b>

```
. corr Crime Income Urban
(obs=67)
```

	Crime	Income	Urban
Crime	<b>1.0000</b>		
Income	<b>0.4338</b>	<b>1.0000</b>	
Urban	<b>0.6774</b>	<b>0.7307</b>	<b>1.0000</b>

### Question 3

Table 11.12 shows Stata output from fitting the multiple regression model to recent statewide data, excluding D.C., on  $y$  = violent crime rate (per 100,000 people),  $x_1$  = poverty rate (percentage with income below the poverty line), and  $x_2$  = percentage living in urban areas.

**TABLE 11.14**

```
. regress violent poverty urban
```

Source	SS	df	MS	Number of obs =	50
Model	2448368.07	2	1224184.	F(2, 47)	= 31.249
Residual	1841257.15	47	39175.68	Prob > F	= 0.0001
Total	4289625.22	49	87543.37	R-squared	= 0.5708
				Adj R-squared	= 0.5525
				Root MSE	= 197.928

	Coef.	Std. Err.	t	P> t
violent	32.622	6.677	4.885	0.0001
poverty	9.112	1.321	6.900	0.0001
_cons	-498.683	140.988	-3.537	0.0009

```
. corr violent poverty urban
```

	violent	poverty	urban
violent	1.0000		
poverty	.3688	1.0000	
urban	.5940	-.1556	1.0000

1. Write down the model equation (with the estimated coefficients).
2. Massachusetts had  $y = 805$ ,  $x_1 = 10.7$ , and  $x_2 = 96.1$ . Find its predicted violent crime rate. Find the residual, and interpret.

3. Interpret the fit by showing the model equation relating  $y$  and  $x_1$  for states with 1)  $x_2=0$ , 2)  $x_2=100$ . Interpret.
4. Conduct a hypothesis testing when the null hypothesis is that there is no relationship between violent crime and poverty.
5. When we add  $x_3$  = percentage of single-parent family to the model, we get the results in the table below. Why do you think the effect of poverty rate is much lower after  $x_3$  is added to the model?

**TABLE 11.15**

Variable	Coefficient	Std. Error
Intercept	-1197.538	
Poverty	18.283	6.136
Urban	7.712	1.109
Single parent	89.401	17.836
$R^2$	0.722	