

HW#3

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5/15/2019

% Table created by stargazer v.5.2.2 by Marek Hlavac, Harvard University. E-mail: hlavac at fas.harvard.edu
% Date and time: Tue, May 21, 2019 - 21:58:51

####Explain what the correlation coefficients mean

For every unit increase in level of risk reduction, the respondent is more likely to vote yes by 0.004169 units.

For every unit increase in annual payment for the household, the respondent is less likely to vote yes by 0.005865 units.

For every unit increase in measure of environmental concern, the respondent is more likely to vote yes by 0.083213 units.

Earning an income in the one-percent makes the respondent 0.059366 units more likely to vote yes.

Earning a poor income makes the respondent 0.040298 units more likely to vote yes.

Earning a rich income makes the respondent 0.060219 units more likely to vote yes.

Earning a very rich income makes the respondent 0.277879 units more likely to vote yes.

Each unit increase in age until 30 makes the respondent 0.092936 units more likely to vote yes.

Each unit increase in age until 40 makes the respondent 0.127045 units less likely to vote yes.

Each unit increase in age until 60 makes the respondent 0.107886 units less likely to vote yes.

$$Odds(VotingYes) = 0.1197 + \beta_1(Risk) + \beta_2(BID) + \beta_3 + (NEP) + \beta_4(Income) + \beta_5(Age)$$

Table 1:

	<i>Dependent variable:</i>
	vote
agetofifty	0.010 (0.063)
agetoforty	−0.020 (0.062)
agetosixty	−0.016 (0.060)
agetothirty	0.020 (0.058)
incomeone_percent	0.009 (0.060)
incomepoor	0.003 (0.065)
incomerich	0.007 (0.068)
incomeevery__rich	0.047 (0.067)
NEP	0.016*** (0.002)
risk	0.001 (0.001)
bid	−0.001 (0.001)
Constant	0.120 (0.120)
Observations	500
R ²	0.120
Adjusted R ²	0.100
Residual Std. Error	0.429 (df = 488)
F Statistic	6.055*** (df = 11; 488)
<i>Note:</i>	*p<0.1; **p<0.05; ***p<0.01

$$Odds(Voting\ Yes) = 0.11970 + 0.02044(Age30) - 0.02012(Age40) + 0.00998(Age50) - 0.01623(Age60) + 0.00274(Poor) + 0.00749$$

- Based on this regression, what can you say about the value of a single prevented whale death? (Hint: Think about how risk reduction for all whales translates into the number of whale deaths avoided) (4% reduction in risk = 1 whale saved)

“{r}

A 4% reduction in risk = 1 whale saved

to find this out I am changing the risk of the 3 random people looked at in #3 and changing the risk from 60-64

their responding change in bid should equal their WTP for 1 whale

```
bid_7(60) # 446.7733 bid_7(64) # 449.5568
wtp_7<- bid_7(64) - bid_7(60) wtp_7 #2.783438
wtp_10<- bid_10(64) - bid_10(60) wtp_10 # 2.783438
wtp_1<- bid_1(64) - bid_1(60) wtp_1 # 2.783438
```

the overall willingness to pay for a whale in SB is 2.783438, this will be consistent among all people because we are using a linear model (lm) not a (glm)

- Pick three arbitrary respondents. For each of those respondents, estimate their WTP for a VSR program offering 60% risk reduction.

```
# not sure whos model to use but for now I will use my regression

#####
# For this portion put age, NEP, Income, risk (60), and age into the equation
# next set the equation = nd solve for the bid vlaue
# once yo have the bid value put all of the inputs into the regression eqaution and solve for vote

# if they are "middle income" put nothing in for income becasue this is the refrence value and is what
#####

# risk = 0.6

# look at person "#7" and plug their info into the regression equation
# bid = ?
# NEP = 50
# income = middle
```

```
# age = 30
# income = middle (so set = 0)

# 1 = 0.1196977 + 0.0204401*1 + 0.0158639*50 + 0.0007445*60 - 0.0010699*x

bid_7<- function(x){(0.1196977 + 0.0204401*1 + 0.0158639*50 + 0.0007445*x - 0.5)/0.0010699}
bid_7(60)
```

```
## [1] 446.7733
```

```
# bid = 446.7733
#####
#Look at person #10
# bid = ?
# NEP = 32
# Income = poor
# age = 30
# risk = 60

# 1 = 0.1196977 + 0.0204401*1 + 0.0027386*1 + 0.0158639*32 + 0.0007445*60 - 0.0010699*x

# 1 + 0.0010699*x = 0.1196977 + 0.0204401*1 + 0.0027386*1 + 0.0158639*32 + 0.0007445*60

# x = (0.1196977 + 0.0204401*1 + 0.0027386*1 + 0.0158639*32 + 0.0007445*x - 1)/0.0010699

bid_10<- function(x){ (0.1196977 + 0.0204401*1 + 0.0027386*1 + 0.0158639*32 + 0.0007445*x - 0.5)/0.0010699}

bid_10(60)
```

```
## [1] 182.4387
```

```
# bid = 182.4387
#####

# person 1
# bid = ?
# NEP = 15
# Income = very_rich
# age = 30
# risk = 60

# 1 = 0.1196977 + 0.0204401*1 + 0.0467922*1 + 0.0158639*15 + 0.0007445*60 - 0.0010699*x

# 1 - 0.0010699*x = 0.1196977 + 0.0204401*1 + 0.0467922*1 + 0.0158639*15 + 0.0007445*60

# x = (0.1196977 + 0.0204401*1 + 0.0467922*1 + 0.0158639*15 + 0.0007445*60 - 1)/0.0010699

bid_1<- function(x){(0.1196977 + 0.0204401*1 + 0.0467922*1 + 0.0158639*15 + 0.0007445*x - 0.5)/0.0010699}

bid_1(60)
```

```
## [1] -28.45266
```

```
# bid = -28.45266
```

Based on this regression, what can you say about the value of a single prevented whale death? (Hint: Think about how risk reduction for all whales translates into the number of whale deaths avoided) (4% reduction in risk = 1 whale saved)

```
# A 4% reduction in risk = 1 whale saved  
# to find this out I am changing the risk of the 3 random people looked at in #3 and changing the risk f  
# their responding change in bid should equal their WTP for 1 whale
```

```
bid_7(60) # 446.7733
```

```
## [1] 446.7733
```

```
bid_7(64) # 449.5568
```

```
## [1] 449.5568
```

```
wtp_7<- bid_7(64) - bid_7(60)  
wtp_7 #2.783438
```

```
## [1] 2.783438
```

```
wtp_10<- bid_10(64) - bid_10(60)  
wtp_10 # 2.783438
```

```
## [1] 2.783438
```

```
wtp_1<- bid_1(64) - bid_1(60)  
wtp_1 # 2.783438
```

```
## [1] 2.783438
```

```
# the overall willingness to pay for a whale in SB is 2.783438, this will be consistant among all peopl
```

4. Now repeat this process and estimate the mean WTP for a VSR program offering 60% risk reduction among Santa Barbara County households.

```
# take the average amongst all of the values, average NEP, average income = medium, have to see the mode
```

```
# create a equation for bid using these values and find the bid value for a 60% reduction
```

```
# make a table of the averages
```

```
# make income and age number so that their modes show up in the summary table  
data_new<- mutate(data, "income_num" = case_when(income == "poor" ~ 0,  
                                                    income == "middle" ~ 1,  
                                                    income == "rich" ~ 2,  
                                                    income == "very_rich" ~ 3,  
                                                    income == "one_percent" ~ 4),  
                  "age_num" = case_when(age == "tothirty" ~ 30,  
                                         age == "toforty" ~ 40,
```

```

age == "tofifty" ~ 50,
age == "tosixty" ~ 6))

data_new

## # A tibble: 500 x 8
##   risk    bid vote   NEP income    age    income_num age_num
##   <int> <int> <int> <int> <chr>    <chr>      <dbl>    <dbl>
## 1     20     80     0    15 very_rich tothirty      3     30
## 2     40    100     1    37 middle   tothirty      1     30
## 3     80     40     1    32 one_percent tothirty      4     30
## 4     20     40     0    32 middle   tothirty      1     30
## 5     80     80     1    40 middle   toforty       1     40
## 6     80     80     1    29 poor     tothirty      0     30
## 7     60     20     1    50 middle   tothirty      1     30
## 8     60     40     1    46 poor     toforty       0     40
## 9     20     40     0    27 very_rich toforty       3     40
## 10    60     40     1    32 poor     tothirty      0     30
## # ... with 490 more rows

#names(data_new)<- c("risk", "bid", "vote", "nep", "income", "age", "income_num", "age_num")

data_table_sum <- data_new %>%
  summarise(
    mean_nep = mean(NEP),
    mode_age = mode(age_num),
    mode_income = mode(income_num)
  )

data_table_sum

## # A tibble: 1 x 3
##   mean_nep mode_age mode_income
##   <dbl> <chr>    <chr>
## 1    38.4 numeric  numeric

# mean NEP = 38.366
# most common age = 30
# average income = medium

# plug in these values and find the WTP for a 60% reduction

# vote = 0.1196977 + 0.0204401(age_30) - 0.0201190(age_40) + 0.0099816(age_50) - 0.0162261(age_60) + 0.0
# 0.5 + 0.0010699(bid) = 0.1196977 + 0.0204401(age_30) - 0.0201190(age_40) + 0.0099816(age_50) - 0.01622
# (bid) = (0.1196977 + 0.0204401(age_30) - 0.0201190(age_40) + 0.0099816(age_50) - 0.0162261(age_60) + 0
bid_mean<- function(x){(0.1196977 + 0.0204401*1 + 0.0158639*38.366 + 0.0007445*x - 0.5)/0.0010699}

bid_mean(60)

## [1] 274.2707

# avergae WTP = 274.2707

```

5. If there are 150,000 households in Santa Barbara County, estimate the total benefit of a VSR program to these households

```
# a 60% reduction would lead to 15 whales saved  
# Before we solved that each person values each whale worth 2.783438  
# multiply the wtp per whale by the population
```

```
total_benefit<- 2.783438*15*150000  
total_benefit # total benefit to sb for a 60% reduction is 6262736
```

```
## [1] 6262736
```

6. Based only on these benefits, do the benefits of the VSR program outweigh the costs?

```
# take the wtp and multiply it by the number of people  
# then compare it to the total benefit solved for in problem 5
```

```
total_cost<- 150000*274.2707  
total_cost
```

```
## [1] 41140605
```

```
total_cost - total_benefit # the benefits outweigh the costs where costs - benefits = 34877870
```

```
## [1] 34877870
```

7. Suppose the price of a carbon credit is \$Z per ton, at what price Z* will the shipping industry decide to adopt the VSR for purely self-interested reasons? (Note: if ships voluntarily reduce speed, then the \$7 million cost (noted above) of implementing the VSR would be avoided)

```
# at a 60% reduction will cost the ship $1000 but also result in 20 fewer tons of CO2 per trip  
# 20 tons = 1000 in costs
```

```
cost_per_tonn<- 1000/20  
cost_per_tonn # 50 dollars per ton reduced
```

```
## [1] 50
```

```
# If the price of the carbon credit Z were anything above $50 then the shipping company will adopt the
```

8. Now suppose the carbon credit price is exactly Z*, so all ships voluntarily reduce speed to achieve the 60% risk reduction. Approximately how many whales would be saved by this? What would be the social value of allowing ships to enter the carbon trading market?

```
# the number of whales saved a year with a 60% reduction would entail 15 whales saved
```

```
# to find the social value of the carbon program in the channel I think you would take the value we got
```

```
# 15 whales would be saved from a 60% reduction
```

```
total_benefit # 6262736 total social benefit from the program
```

```
## [1] 6262736
```