# Core Econometrics III: Problem Set 1

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04/06/2022

# Question 1

```
data = read.csv("data-001.csv")
```

### Question 2

Table 1: Regression Results

	Dependent variable:
	income_black_2010
pop_enslaved_1860	$-0.267^{**}$
	(0.129)
pop_total_1860	0.056
	(0.063)
pop_total_2010	0.011***
	(0.002)
Constant	28,951.560***
	(688.725)
Observations	710
$\mathbb{R}^2$	0.061
Adjusted R <sup>2</sup>	0.057
Residual Std. Error	11,789.970 (df = 706)
F Statistic	$15.214^{***} (df = 3; 706)$
Note:	*p<0.1; **p<0.05; ***p<0.01

The coefficient on "pop\_enslaved\_1860" is -0.2670247 This tells us that as the population of enslaved individuals of a county in 1860 increases by 1, the median income for black households in 2010 in that county changes by \$-0.2670247

#### Question 3

The coefficient on "pop\_enslaved\_1860" is -0.2670247 which is the same as in question 2.

#### Question 4

```
# Regression function
reg_fun = function(y, x) {

# linear algebra equation
coef = solve(t(x) %*% x) %*% t(x) %*% y

return(coef)
}
```

Results:

```
reg_fun(Y, X)

## [,1]

## [1,] 2.895156e+04

## [2,] -2.670247e-01

## [3,] 5.592848e-02

## [4,] 1.107846e-02

Success!
```

#### Question 5

```
reg_fun2 = function(y, x) {

# coefficient equation
coef = solve(t(x) %*% x) %*% t(x) %*% y
```

```
# standard errors
    # error (residuals)
e = (y - x %*% coef)
    # variance sigma estimate calculation
s_sq = (1/(dim(x)[1] - dim(x)[2]-1))*sum(e^2)

# calculate variance matrix
variance_matrix = s_sq * solve(t(x) %*% x)

# arrange the results
stnd_errors = sqrt(diag(variance_matrix))

# Combine results
results = cbind(coef, stnd_errors)
return(results)
}
```

Results:

```
reg_fun2(Y, X)

## stnd_errors

## [1,] 2.895156e+04 6.892133e+02

## [2,] -2.670247e-01 1.292888e-01

## [3,] 5.592848e-02 6.353467e-02

## [4,] 1.107846e-02 1.804909e-03
```

My function reports the coefficients and standard errors correctly.

### Question 6

To be approximately correct, the standard errors reported from my function rely on the assumptions of homoskedasticity, nonautocorrelation, and normally distributed errors:  $\epsilon | X \sim N(0, \sigma^2 I)$ 

## Question 7

In order for my coefficients to be interpretable as causal, one needs to assume that the model we're estimating is the true model, that there are no omitted relevant variables, that the exogenous variables are in fact exogenous.

#### Extra Credit

```
coef = solve(t(x) %*% x) %*% t(x) %*% y

# standard errors
    # error (residuals)
e = (y - x %*% coef)
    # standard error calculation
s_sq = (1/(dim(x)[1] - dim(x)[2]-1))*sum(e^2)

# calculate variance matrix
variance_matrix = s_sq * solve(t(x) %*% x)

# arrange the results
stnd_errors = sqrt(diag(variance_matrix))
results = cbind(coef, stnd_errors)
return(results)
}
```

#### **Results:**