1. Plot the Vote

```
Question 1:
```

```
def proportions in resamples():
    prop c = make array()
    for i in np.arange(5000):
        bootstrap = votes.sample()
        single_proportion = np.count nonzero(bootstrap.column(0) == 'C')
        / bootstrap.num rows
        prop c = np.append(prop c, single proportion)
    return prop c
Ouestion 2:
c lower bound = percentile(2.5, sampled proportions)
c upper bound = percentile(97.5, sampled proportions)
Question 3:
bins = np.arange(-0.2, 0.2, 0.01)
def leads in resamples():
    leads = make array()
    for i in np.arange(5000):
        bootstrap = votes.sample()
        c = np.count_nonzero(bootstrap.column(0) == 'C') / bootstrap.num_rows
        t = np.count nonzero(bootstrap.column(0) == 'T') / bootstrap.num rows
        leads = np.append(leads, c-t)
    return leads
sampled leads = leads in resamples()
Table().with column('leads', sampled leads).hist(bins=bins)
```

2. Interpreting Confidence Intervals

diff_lower_bound = percentile(2.5, sampled_leads)
diff upper bound = percentile(97.5, sampled leads)

Question 1:

```
correct option = 2
```

Ouestion 2:

```
true proportion intervals = 9500
```

Question 4:

```
confidence_interval_80 = interval_2
confidence_interval_90 = interval_1
confidence_interval_99 = interval_3
```

```
Question 5:
candidates tied = 2
Ouestion 6:
cutoff one percent = 2
Question 7:
cutoff ten percent = 3
3.
      Triple Jump Distances vs. Vertical Jump Heights
Note:
For the following questions we are using the functions f_standard_units,
f_correlation, f_slope and f_intercept. These functions were not given and you
had to define them yourself.
def f standard units(any numbers):
    return (any numbers - np.mean(any numbers))/np.std(any numbers)
def f correlation(t, label x, label y):
    return np.mean(f standard units(t.column(label x)) *
      f standard units(t.column(label y)))
def f_slope(t, label_x, label_y):
    r = f_correlation(t, label_x, label_y)
    return r*np.std(t.column(label y))/np.std(t.column(label x))
def f intercept(t, label x, label y):
    return np.mean(t.column(label y)) - f slope(t,label x,label y) *
      np.mean(t.column(label x))
Question 1:
jumps.scatter("triple")
linear correlation = True
Question 2:
r = 0.5
```

print('r:', parameters.item(0), '; slope:', parameters.item(1), '; intercept:',

Question 3:

parameters.item(2))

def regression_parameters(t):

r = f_correlation(jumps, 0, 1)
slope = f slope(jumps, 0, 1)

intercept = f_intercept(jumps, 0, 1)
return make_array(r, slope, intercept)

parameters = regression parameters(jumps)

Question 4:

```
triple record vert est = f slope(jumps, 0, 1) * 1829 + f intercept(jumps, 0, 1)
print("Predicted vertical jump distance: {:f}
centimeters".format(triple record vert est))
```

Ouestion 5:

estimation accurate = False

4. Cryptocurrencies

Ouestion 1:

```
## Using functions introduced in section 3
r = f_correlation(Table().with_columns("btc", btc.column("open"), "eth",
eth.\overline{column}("open")), 0, 1)
## Without fucntions introduced in section 3
def std units(arr):
    return (arr - np.mean(arr))/np.std(arr)
standard btc = std units(btc.column('open'))
standard eth = std units(eth.column('open'))
r = np.mean(standard_btc * standard_eth)
```

Question 2:

```
def eth_predictor(btc_price):
    table = Table().with columns("btc", btc.column("open"), "eth",
      eth.column("open"))
    parameters = regression parameters(table)
    slope = parameters.item(1)
    intercept = parameters.item(2)
    return slope * btc price + intercept
```

Ouestion 3:

```
predictions = eth predictor(btc.column('open'))
prices with predictions = Table().with columns(
    "btc", btc.column("open"),
"eth", eth.column("open"),
    "eth prediction", predictions)
```

Question 6:

```
regression changes = [False, True, True]
```