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```
In [2]:
          import numpy as np
          import pandas as pd
          import torch
          import pyro
          import pyro.distributions as dist
          pyro.set rng seed(101)
          pyro.enable validation(True)
In [3]:
         df = pd.read_csv('./driver.csv', index_col=0)
In [4]:
          df.head()
Out[4]:
             X
                      y z
          1 0 1.490090 0
          2 1 5.170279 1
          3 1 7.434170 1
             0 1.531446 0
             0 0.935765 0
In [5]:
          df.describe()
Out[5]:
                         X
                                     У
                                                Z
          count 1000.000000 1000.000000 1000.000000
                   0.498000
                               2.531169
                                          0.491000
           mean
            std
                   0.500246
                               2.535948
                                          0.500169
            min
                   0.000000
                              -3.353001
                                          0.000000
           25%
                   0.000000
                               0.244929
                                          0.000000
           50%
                   0.000000
                               2.321081
                                          0.000000
           75%
                   1.000000
                               4.864423
                                          1.000000
           max
                   1.000000
                               7.895047
                                          1.000000
```

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```
In [39]: # Calculating do operation using valid adjustment set formula

def estimate_y(x):
    est_sum = 0
    for z in df['z'].unique():
        filter_df = df.query('x == @x & z == @z')
        mean_y = np.mean(filter_df['y'])

    prob_z = df.query('z == @z').size/df.size

    est_sum += (mean_y*prob_z)

return est_sum
```

```
In [37]: # E[Y_X=x | X=cx]
def estimate_counterfactual(x, cx):
    est_sum = 0
    for z in df['z'].unique():
        filter_df = df.query('x == @x & z == @z')
        mean_y = np.mean(filter_df['y'])

    filter_df = df.query('x == @cx')
        prob_z = filter_df.query('z == @z').size/filter_df.size
        est_sum += (mean_y*prob_z)

    return est_sum
```

```
In [38]: print('E(Y_X=0 | X=1) = %.3f' % (estimate_counterfactual(0, 1)))
    print('ETT = %.3f' % (estimate_counterfactual(1, 1) - estimate_counter
    factual(0, 1)))
    print('E(Y_X=1 - Y_X=0) = %.3f - %.3f = %.3f' % (estimate_y(1), estimate_y(0), estimate_y(1) - estimate_y(0)))

E(Y_X=0 | X=1) = 3.091
    ETT = 1.346
    E(Y_X=1 - Y_X=0) = 3.075 - 1.861 = 1.214
```

Analysis

It is clear from the results that the ETT is slight higher than the total effect. This indicates that the training program must have had an impact on the motivation of the drivers results in higher revenues. Having done the ETT analysis, we can safely say that the training program in general is going to result in higher revenues despite the original motivation of the drivers.

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In []: