Activity Course 4 Automatidata project lab

November 1, 2023

1 Automatidata project

Course 4 - The Power of Statistics

You are a data professional in a data consulting firm, called Automatidata. The current project for their newest client, the New York City Taxi & Limousine Commission (New York City TLC) is reaching its midpoint, having completed a project proposal, Python coding work, and exploratory data analysis.

You receive a new email from Uli King, Automatidata's project manager. Uli tells your team about a new request from the New York City TLC: to analyze the relationship between fare amount and payment type. A follow-up email from Luana includes your specific assignment: to conduct an A/B test.

A notebook was structured and prepared to help you in this project. Please complete the following questions.

2 Course 4 End-of-course project: Statistical analysis

In this activity, you will practice using statistics to analyze and interpret data. The activity covers fundamental concepts such as descriptive statistics and hypothesis testing. You will explore the data provided and conduct A/B and hypothesis testing.

The purpose of this project is to demostrate knowledge of how to prepare, create, and analyze A/B tests. Your A/B test results should aim to find ways to generate more revenue for taxi cab drivers.

Note: For the purpose of this exercise, assume that the sample data comes from an experiment in which customers are randomly selected and divided into two groups: 1) customers who are required to pay with credit card, 2) customers who are required to pay with cash. Without this assumption, we cannot draw causal conclusions about how payment method affects fare amount.

The goal is to apply descriptive statistics and hypothesis testing in Python. The goal for this A/B test is to sample data and analyze whether there is a relationship between payment type and fare amount. For example: discover if customers who use credit cards pay higher fare amounts than customers who use cash.

This activity has four parts:

Part 1: Imports and data loading * What data packages will be necessary for hypothesis testing?

Part 2: Conduct EDA and hypothesis testing * How did computing descriptive statistics help you analyze your data?

• How did you formulate your null hypothesis and alternative hypothesis?

Part 3: Communicate insights with stakeholders

- What key business insight(s) emerged from your A/B test?
- What business recommendations do you propose based on your results?

Follow the instructions and answer the questions below to complete the activity. Then, you will complete an Executive Summary using the questions listed on the PACE Strategy Document.

Be sure to complete this activity before moving on. The next course item will provide you with a completed exemplar to compare to your own work.

3 Conduct an A/B test

4 PACE stages

Throughout these project notebooks, you'll see references to the problem-solving framework PACE. The following notebook components are labeled with the respective PACE stage: Plan, Analyze, Construct, and Execute.

4.1 PACE: Plan

In this stage, consider the following questions where applicable to complete your code response: 1. What is your research question for this data project? Later on, you will need to formulate the null and alternative hypotheses as the first step of your hypothesis test. Consider your research question now, at the start of this task.

The main purpose of this part of the project is to perform A/B testing to analyze the relationship between fare amount and payment type.

Complete the following steps to perform statistical analysis of your data:

4.1.1 Task 1. Imports and data loading

Import packages and libraries needed to compute descriptive statistics and conduct a hypothesis test.

Hint:

Before you begin, recall the following Python packages and functions that may be useful:

Main functions: stats.ttest_ind(a, b, equal_var)

Other functions: mean()

Packages: pandas, stats.scipy

```
[1]: import pandas as pd import numpy as np from scipy import stats
```

Note: As shown in this cell, the dataset has been automatically loaded in for you. You do not need to download the .csv file, or provide more code, in order to access the dataset and proceed with this lab. Please continue with this activity by completing the following instructions.

```
[2]: # Load dataset into dataframe taxi_data = pd.read_csv("2017_Yellow_Taxi_Trip_Data.csv", index_col = 0)
```

4.2 PACE: Analyze and Construct

In this stage, consider the following questions where applicable to complete your code response: 1. Data professionals use descriptive statistics for Exploratory Data Analysis. How can computing descriptive statistics help you learn more about your data in this stage of your analysis?

Descriptive stats are important because it summarizes the main features of datasets, and computing those stats uncover important information about the data that you may not catch at first glance.

4.2.1 Task 2. Data exploration

2

Use descriptive statistics to conduct Exploratory Data Analysis (EDA).

Hint:

38942136

Refer back to Self Review Descriptive Statistics for this step-by-step process.

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Note: In the dataset, payment_type is encoded in integers: * 1: Credit card * 2: Cash * 3: No charge * 4: Dispute * 5: Unknown

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```
[5]: taxi_data.head()
[5]:
                VendorID
                             tpep_pickup_datetime
                                                     tpep_dropoff_datetime
     24870114
                       2
                           03/25/2017 8:55:43 AM
                                                     03/25/2017 9:09:47 AM
                           04/11/2017 2:53:28 PM
     35634249
                       1
                                                     04/11/2017 3:19:58 PM
                            12/15/2017 7:26:56 AM
                                                     12/15/2017 7:34:08 AM
     106203690
                       1
```

30841670	2 04/15/	′2017 11:32:20 P	M 04/15/2017 11:	49:03 PM	
	passenger_count	trip_distance	RatecodeID store	e_and_fwd_flag	\
24870114	6	3.34	1	N	
35634249	1	1.80	1	N	
106203690	1	1.00	1	N	
38942136	1	3.70	1	N	
30841670	1	4.37	1	N	

	PULocati	onID	${\tt DOLocationID}$		<pre>payment_type</pre>		fare_amount	extra	\
24870114		100		231		1	13.0	0.0	
35634249		186		43		1	16.0	0.0	
106203690		262		236		1	6.5	0.0	
38942136		188		97		1	20.5	0.0	
30841670		4	112			2	16.5	0.5	
	$\mathtt{mta_tax}$	tip_	amount	tolls_	amount	improv	ement_surchar	ge \	
24870114	0.5		2.76		0.0		0	.3	
35634249	0.5 4.00		4.00	0.0		0.3			
106203690	0.5		1.45		0.0		0.3		
38942136	0.5		6.39		0.0		0.3		
30841670	0.5		0.00		0.0		0.3		
total_amount									
24870114	16.56								
35634249	20.80								
106203690	8.75								
38942136	2	7.69							
30841670	1	7.80							

You are interested in the relationship between payment type and the fare amount the customer pays. One approach is to look at the average fare amount for each payment type.

[8]: taxi_data.describe()

[8]:		VendorID	passenger_cou	nt trip_dista	nce Ratecod	leID \	
	count	22699.000000	22699.0000	00 22699.000	000 22699.000	0000	
	mean	1.556236	1.6423	19 2.913	313 1.043	394	
	std	0.496838	1.2852	31 3.653	171 0.708	391	
	min	1.000000	0.0000	0.000	000 1.000	0000	
	25%	1.000000	1.0000	0.990	000 1.000	0000	
	50%	2.000000	1.0000	00 1.610	000 1.000	0000	
	75%	2.000000	2.0000	3.060	000 1.000	000	
	max	2.000000	6.0000	33.960	99.000	000	
		${\tt PULocationID}$	${\tt DOLocationID}$	<pre>payment_type</pre>	fare_amount	extra	\
	count	22699.000000	22699.000000	22699.000000	22699.000000	22699.000000	
	mean	162.412353	161.527997	1.336887	13.026629	0.333275	
	std	66.633373	70.139691	0.496211	13.243791	0.463097	
	min	1.000000	1.000000	1.000000	-120.000000	-1.000000	
	25%	114.000000	112.000000	1.000000	6.500000	0.000000	
	50%	162.000000	162.000000	1.000000	9.500000	0.000000	
	75%	233.000000	233.000000	2.000000	14.500000	0.500000	
	max	265.000000	265.000000	4.000000	999.990000	4.500000	

```
improvement_surcharge
             mta_tax
                        tip_amount
                                     tolls_amount
                      22699.000000
                                     22699.000000
                                                              22699.000000
count
       22699.000000
mean
            0.497445
                           1.835781
                                          0.312542
                                                                   0.299551
            0.039465
                           2.800626
                                          1.399212
                                                                   0.015673
std
min
           -0.500000
                           0.000000
                                          0.000000
                                                                  -0.300000
25%
                                          0.00000
           0.500000
                           0.000000
                                                                   0.300000
50%
            0.500000
                                          0.000000
                                                                   0.300000
                           1.350000
75%
            0.500000
                           2.450000
                                          0.000000
                                                                   0.300000
            0.500000
                         200.000000
                                         19.100000
                                                                   0.300000
max
       total_amount
       22699.000000
count
mean
           16.310502
std
           16.097295
        -120.300000
min
25%
           8.750000
50%
           11.800000
75%
           17.800000
        1200.290000
max
```

```
[3]: taxi_data.groupby('payment_type')['fare_amount'].mean()
```

```
[3]: payment_type
```

- 1 13.429748
- 2 12.213546
- 3 12.186116
- 4 9.913043

Name: fare_amount, dtype: float64

Based on the averages shown, it appears that customers who pay in credit card tend to pay a larger fare amount than customers who pay in cash. However, this difference might arise from random sampling, rather than being a true difference in fare amount. To assess whether the difference is statistically significant, you conduct a hypothesis test.

4.2.2 Task 3. Hypothesis testing

Before you conduct your hypothesis test, consider the following questions where applicable to complete your code response:

1. Recall the difference between the null hypothesis and the alternative hypotheses. Consider your hypotheses for this project as listed below.

 H_0 : There is no difference in the average fare amount between customers who use credit cards and customers who use cash.

 H_A : There is a difference in the average fare amount between customers who use credit cards and customers who use cash.

Your goal in this step is to conduct a two-sample t-test. Recall the steps for conducting a hypothesis test:

- 1. State the null hypothesis and the alternative hypothesis
- 2. Choose a signficance level
- 3. Find the p-value
- 4. Reject or fail to reject the null hypothesis

Note: For the purpose of this exercise, your hypothesis test is the main component of your A/B test.

You choose 5% as the significance level and proceed with a two-sample t-test.

```
[5]: #Creating dataframes for samples

cash_taxi = taxi_data[taxi_data['payment_type'] == 2]['fare_amount']

credit_taxi = taxi_data[taxi_data['payment_type'] == 1]['fare_amount']
```

```
[7]: #Sig Level

sig_level = 0.05
sig_level
```

[7]: 0.05

```
[7]: #Compute P-Value stats.ttest_ind(a=cash_taxi, b=credit_taxi, equal_var=False)
```

[7]: Ttest_indResult(statistic=-6.866800855655372, pvalue=6.797387473030518e-12)

From the pvalue I have found, given the e-12 I'm going to reject the null hypothesis.

4.3 PACE: Execute

Consider the questions in your PACE Strategy Document to reflect on the Execute stage.

4.3.1 Task 4. Communicate insights with stakeholders

Ask yourself the following questions:

- 1. What business insight(s) can you draw from the result of your hypothesis test?
- 2. Consider why this A/B test project might not be realistic, and what assumptions had to be made for this educational project.

With the answers I have found, the insight is that we would move forward with a plan to have customers use credit more than cash.

There's a lot of factors as to why someone would pay with credit instead of cash. Variables we cannot account for fully which would make this A/B testing skewed. Even just credit being more convienent overall makes i difficult to get informed results.

Congratulations! You've completed this lab. However, you may not notice a green check mark next to this item on Coursera's platform. Please continue your progress regardless of the check mark. Just click on the "save" icon at the top of this notebook to ensure your work has been logged.