Final Project Document

Project Topic

Alibaba Digital Marketing



Team Details

Team 5

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Overview

Alibaba Group Holding Limited (also known as **Alibaba Group** and **Alibaba.com**) is a Chinese multinational technology company specializing in e-commerce, retail, Internet, and technology. On 19th September 2014, Alibaba's market value was US\$231 billion. It is one of the top 10 most valuable and is the 59th biggest public company in the world by the Global 2000 list.

As of 2020, Alibaba has the 6th highest global brand valuation. Alibaba is the world's largest retailer and e-commerce company, is on the list of largest Interest companies and artificial intelligence companies, is one of the biggest venture capital firms, and one of the biggest investment corporations in the world. Its online sales and profits surpassed all US retailers (including Walmart, Amazon, and eBay) combined since 2015.

Goals

- > Finding Recency, Frequency, and Monetary Value to find the company's best customers by using certain measures.
- > To find the customers lost by calculating the churn rate and the customer's lifetime value.
- > Build a recommendation system for suggesting products to the customers that they might also like.
- > To Crawl, Index, and Rank data using the Search engine for better customer experience.
- > Build a dashboard for business owners to better understand their sales and decide future marketing strategies.

> Create a web application for better user experience.

Dataset

- > We will be using datasets available on Kaggle along with data available through different open sources and websites.
- Kaggle Dataset Link: https://www.kaggle.com/AppleEcomerceInfo/ecommerce-inf ormation?select=products.txt
- > http://yongfeng.me/dataset/

Data Sample

```
■ carts_has_products.txt

    □ category.txt

                                 product_id
                                              product_name descriptions
■ option.txt
                                 1200 Macbook Pro (2017)
                                                               The ultimate pro notebook. MacBook Pro features faster processors ;upgraded memory;the Apple
                                 1300 Macbook Air (2015) MacBook Air lasts up to an incredible 12 hours between charges So from your morning coffee 1
■ order.txt
                               1400 Iphone X The iPhone X display is so immersive the device itself disappears into the experience.

1500 Iphone 7 Great connectivity of this device includes Bluetooth 4.2 version with A2DP
■ orders_has_products.txt
■ orders_paid_creditcard.txt
                                1600 Iphone 8 iPhone 8 introduces a glass design. The glass back enables easy wireless charging.

1700 Ipad Air 4th gen The iPad Air is unbelievably thin and light. And yet it is so much more powerful and capable
■ orders_placed_user.txt
product_sold_vendor.txt
                             1800 Ipad Mini 3th gen
                                                              3th gen Everything you love about iPad ♦ the beautiful screen and fast

■ shoppingcart.txt

                                2300 Monoprice Ultra Slim Series High Speed HDMI Cable
                                                                                             The Monoprice Ultra Slim Active High Speed HDMI Cable series
                                        Monoprice Ultra Slim Series High Speed HDMI Cable - 4K Monoprice Commercial Cable supports the following HDMI featu
                              2500 Avantree HT3189 Wireless Headphones Avantree HT3189 Wireless Headphones for TV Watching & PC Gaming with Bluetoc
user_has_creditcard.txt
vendor.txt
                                2600 COWIN E7 PRO Active Noise Cancelling Headphone Bluetooth Headphones with Microphone Hi-Fi Deep Bass Wireless Head
```

Personas

Who – Alibaba's technical teams, stakeholders, and sponsors for deciding their future marketing strategies.

What – Build a recommendation system for the company's customers and dashboards for stakeholders to analyze their sales and other insights.

When – It will be completed within 2 weeks of a given timeline.

Where – We will be working at our remote locations and then collaborate on our work.

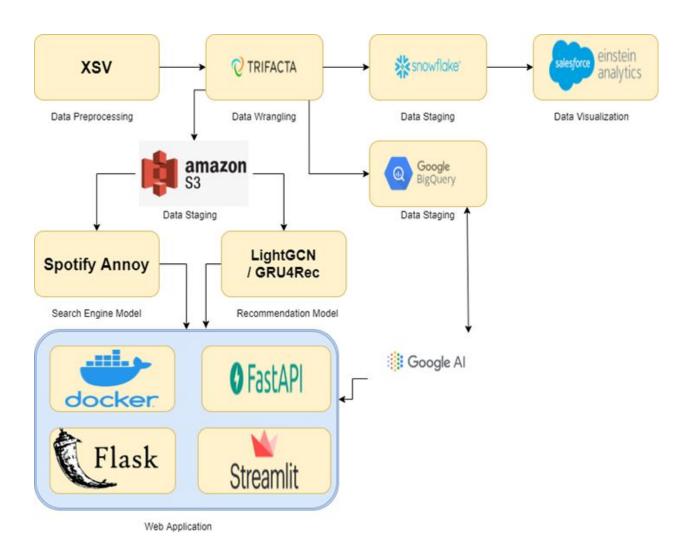
Why – To get data insights that will help the stakeholders decide their marketing strategies and to improve the customer experience by building a recommendation system that will recommend the customers product that may also like.

How – Using the tools and technologies learned in the course.

- > XSV
- > Python
- ➤ Trifacta
- ➤ Salesforce Einstein Analytics
- > Streamlit
- ➤ Flask

> JMeter

Workflow



Use Cases

- ➤ Making user shopping experience more efficient.
- ➤ Know the product trend to make future decisions based on it to increase profitability.
- ➤ Help stakeholders to make important decisions and changes.

Web scraping

Web Scraping (also termed Screen Scraping, Web Data Extraction, Web Harvesting, etc.) is a technique employed to extract large amounts of data from websites whereby the data is extracted and saved to a local file in your computer or to a database in table (spreadsheet) format.

Code Sample:

```
#For web scrapping we will import urlopen and BeautifulSoup
from urllib.request import urlopen
from bs4 import BeautifulSoup

#Link of website which we will be scrapping
url = "https://www.alibaba.com/"
html = urlopen(url)

#Creating a Beautiful Soup object
soup = BeautifulSoup(html, 'lxml')
type(soup)

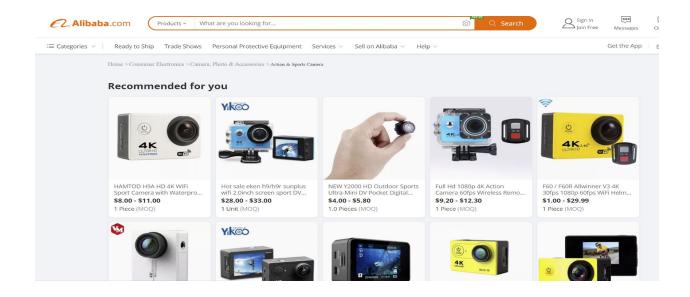
bs4.BeautifulSoup

#Cetting the title of the website which we are scrapping
title = soup.title
print(title)

#Printing the text on the webpage
text = soup.get_text()
#print(text)

soup.find_all('a')
```

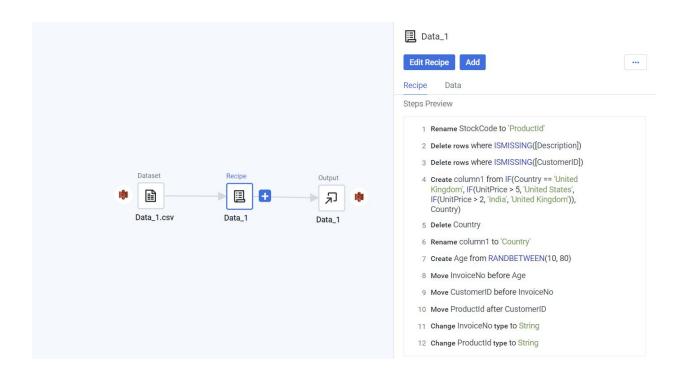
Website:



Trifacta (Data Wrangling)

Trifacta's data wrangling software allows you to prepare & visualize complex data in no time.

# Quantity	InvoiceDate	## UnitPrice	# CustomerID
6	12/1/2010-8:26	2.55	17850
6	12/1/2010 8:26	3.39	17850
8	12/1/2010 8:26	2.75	17850
6	12/1/2010 8:26	3.39	17850
6	12/1/2010 8:26	3.39	17850
2	12/1/2010 8:26	7.65	17850
6	12/1/2010 8:26	4.25	17850
6	12/1/2010 8:28	1.85	17850
6	12/1/2010 8:28	1.85	17850
32	12/1/2010 8:34	1.69	13047
5	12/1/2010 8:34	2.1	13047
6	12/1/2010 8:34	2.1	13047
3	12/1/2010 8:34	3.75	13047
6	12/1/2010 8:34	1.65	13047
6	12/1/2010 8:34	4.25	13047
3	12/1/2010 8:34	4.95	13047
2	12/1/2010 8:34	9.95	13047



RFM and Customer Segmentation

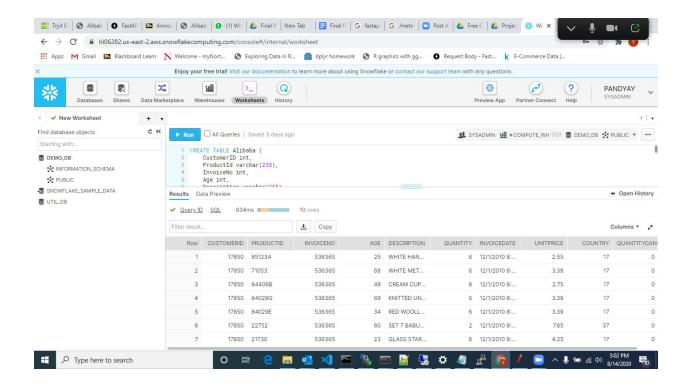
RFM stands for three dimensions:

- Recency How recently did the customer purchase?
- Frequency How often do they purchase?
- Monetary Value How much do they spend?

[71]:		Description	Quantity	InvoiceDate	UnitPrice	Country	QuantityCanceled	TotalPrice	min_recency	max_recency	frequency	monetary_value	RFMScore
	25	WHITE HANGING HEART T- LIGHT HOLDER	6	2010-12-01 08:26:00	2.55	17	0	15.30	372.0	373.0	34	5327.79	411
	68	WHITE METAL LANTERN	6	2010-12-01 08:26:00	3.39	17	0	20.34	372.0	373.0	34	5327.79	411
	49	CREAM CUPID HEARTS COAT HANGER	8	2010-12-01 08:26:00	2.75	17	0	22.00	372.0	373.0	34	5327.79	411
	69	KNITTED UNION FLAG HOT WATER BOTTLE	6	2010-12-01 08:26:00	3.39	17	0	20.34	372.0	373.0	34	5327.79	411
	34	RED WOOLLY HOTTIE WHITE HEART.	6	2010-12-01 08:26:00	3.39	17	0	20.34	372.0	373.0	34	5327.79	411
	60	SET 7 BABUSHKA NESTING BOXES	2	2010-12-01 08:26:00	7.65	37	0	15.30	372.0	373.0	34	5327.79	411
	23	GLASS STAR FROSTED T-LIGHT HOLDER	6	2010-12-01 08:26:00	4.25	17	0	25.50	372.0	373.0	34	5327.79	411

SnowFlake

Snowflake is a cloud-based Data Warehouse solution provided as a Saas (Software-as-a-Service) with full support for ANSI SQL. It also has a unique architecture that enables users to just create tables and start querying data with very little administration or DBA activities needed.

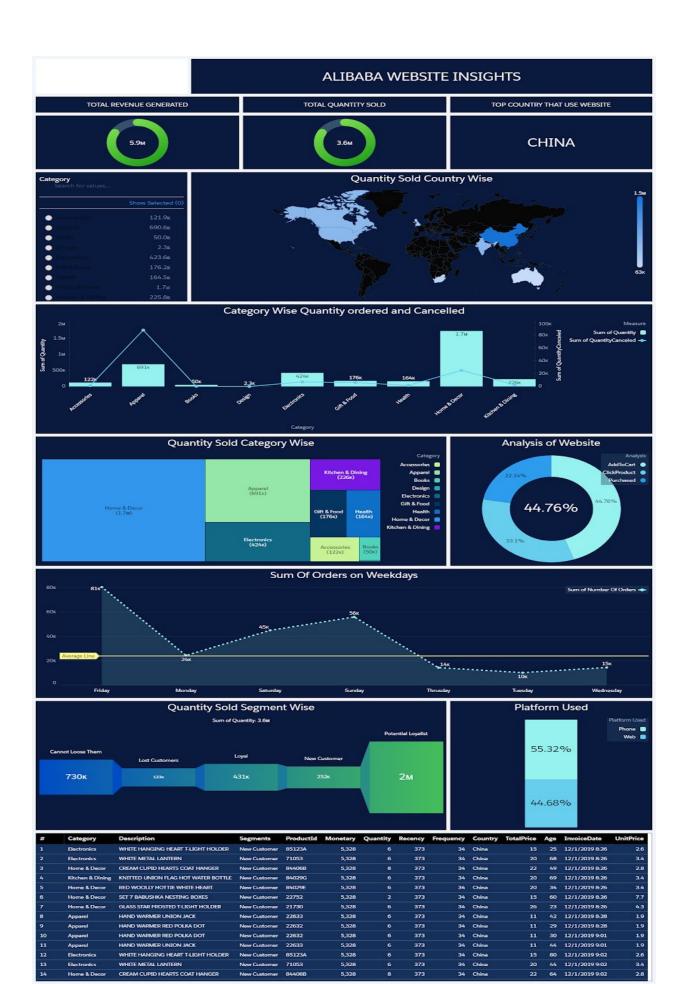


Salesforce Einstein Analytics Dashboard

Salesforce.com, inc. is an American cloud-based software company headquartered in San Francisco, California. It provides customer relationship management (CRM) service and also sells a complementary suite of enterprise applications focused on customer service, marketing automation, analytics, and application development.

- Total Revenue generated, Total Quantity sold, Top country that use the website
- Which Country has purchased most items
- Category wise items ordered and Cancelled
- How many items after placing in cart where purchased
- Number of orders based on weekdays
- Which has platform(Phone and Web) used for Most

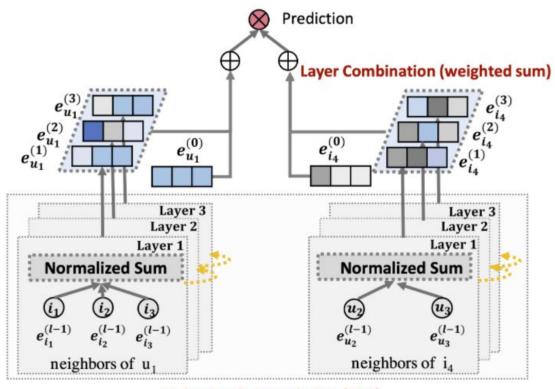
- Quantity sold based on segments (Loyal, potential Loyalist, Cannot lose them, new Customer and Lost customers)
- Tablewise segmentation



Recommendation Model

LightGCN

LightGCN is a simplified design of GCN to make it more concise and appropriate for a recommendation.



Light Graph Convolution (LGC)

In LightGCN, we adopt the simple weighted sum aggregator and abandon the use of feature transformation and nonlinear activation.

```
In [12]: ► topk_scores = model.recommend_k_items(test, top_k=TOP_K, remove_seen=True)
             topk_scores.head()
   Out[12]:
                userID itemID prediction
              0 12349 23245 8.174675
              1 12349 22839 7.465651
              2 12349 22423 7.303855
                12349 23284 6.950679
              4 12349 22507 6.786175
In [13]: ► eval map = map at k(test, topk scores, k=TOP K)
             eval_ndcg = ndcg_at_k(test, topk_scores, k=TOP_K)
             eval_precision = precision_at_k(test, topk_scores, k=TOP_K)
             eval_recall = recall_at_k(test, topk_scores, k=TOP_K)
             print("MAP:\t%f" % eval map,
                   "NDCG: \t%f" % eval ndcg,
                   "Precision@K:\t%f" % eval_precision,
                   "Recall@K:\t%f" % eval recall, sep='\n')
             MAP:
                   0.041018
             NDCG: 0.124642
             Precision@K: 0.088660
Recall@K: 0.090790
```

Surprise Singular Value Decomposition (SVD)

SVD introduces two new scalar variables: the user biases bu and item biases bi. The user biases are supposed to capture the tendency of some users to rate items higher (or lower) than the average. The same goes for items: some items are usually rated higher than some others.

```
In [41]: ► svd = surprise.SVD(random state=0, n factors=200, n epochs=30, verbose=True)
             with Timer() as train time:
                svd.fit(train set)
             print("Took {} seconds for training.".format(train_time.interval))
             Processing epoch 0
             Processing epoch 1
             Processing epoch 2
             Processing epoch 3
             Processing epoch 4
             Processing epoch 5
             Processing epoch 6
             Processing epoch 7
             Processing epoch 8
             Processing epoch 9
             Processing epoch 10
             Processing epoch 11
             Processing epoch 12
             Processing epoch 13
             Processing epoch 14
          print("RMSE:\t\t%f" % eval_rmse,
                "MAE:\t\t%f" % eval mae,
                "rsquared:\t%f" % eval_rsquared,
                "exp var:\t%f" % eval_exp_var, sep='\n')
          print('----')
          print("MAP:\t%f" % eval_map,
                "NDCG: \t%f" % eval_ndcg,
                "Precision@K:\t%f" % eval_precision,
                "Recall@K:\t%f" % eval recall, sep='\n')
          RMSE:
                         2.451093
          MAE:
                        2.000027
          rsquared:
                        -1.992368
          exp var:
                        -0.000013
          ----
          MAP:
                  0.002902
          NDCG: 0.016522
          Precision@K: 0.013356
          Recall@K:
                        0.008077
```

Search Engine

Search By an Artistic Style

One typical business case is an eCommerce website that allows to search of a poster based on the example uploaded by the user. A user would usually expect to get results that are similar in terms of artistic style. A real search system would typically mix style similarity with other similarity scores such as image subject or category (landscape, still life, etc.) that can also be obtained using deep neural networks.

```
[] def search_by_style(reference_image, max_results=10):
    v0 = image_style_embeddings[reference_image]
    distances = {}
    for k,v in image_style_embeddings.items():
        d = sc.spatial.distance.cosine(v0, v)
        distances[k] = d

    sorted_neighbors = sorted(distances.items(), key=lambda x: x[1], reverse=False)

    f, ax = plt.subplots(1, max_results, figsize=(16, 8))
    for i, img in enumerate(sorted_neighbors[:max_results]):
        ax[i].imshow(images[img[0]])
        ax[i].set_axis_off()

    plt.show()

[] search_by_style('23200.jpg')

[]
```

Streamlit Application

Streamlit is an open-source app framework for Machine Learning and Data Science teams. Create beautiful data apps in hours, not weeks. All in pure Python.

https://finalproject-streamlit-app.herokuapp.com/

SEARCH FOR SIMILAR PRODUCTS



Choose a product

1528.jpg

Product Selected



Similar Products



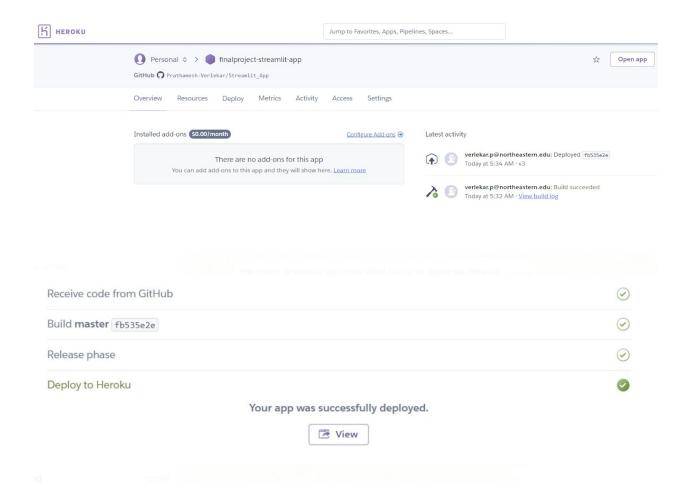






Heroku Deployment

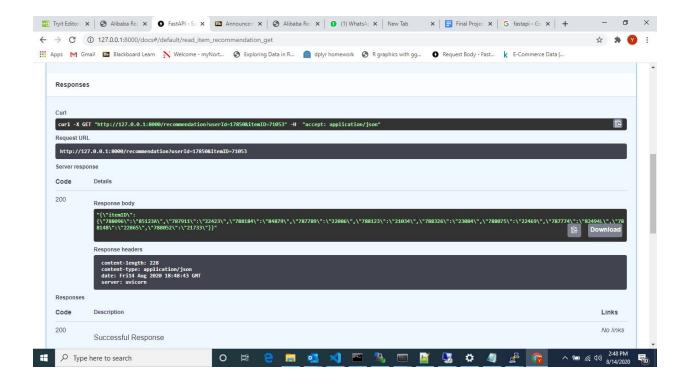
Heroku is a cloud platform as a service supporting several programming languages. One of the first cloud platforms, Heroku has been in development since June 2007, when it supported only the Ruby programming language, but now supports Java, Node.js, Scala, Clojure, Python, PHP, and Go.



FastAPI

FastAPI is a modern, fast (high-performance), a web framework for building APIs with Python 3.6+ based on standard Python type hints. The key features are: ... One of the

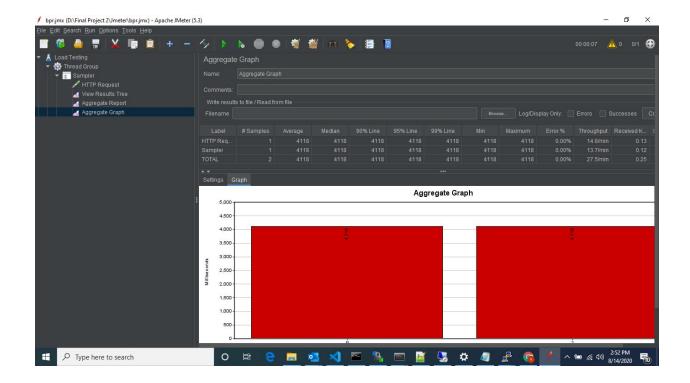
fastest Python frameworks available. Fast to code: Increase the speed to develop features by about 200% to 300%.



JMeter

Apache **JMeter** is a **testing** tool used for analyzing and measuring the performance of different software services and products. It is a pure Java open source software used for **testing** Web Application or FTP application. It is used to execute performance **testing**, load **testing**, and functional **testing** of web applications.

We used Jmeter to test our three recommendation models and based on the throughput rate we were able to decide that LightGCN was much better and faster as compared to other models so we used LightGCN for building our recommendation web application.



Flash Application Deployed on AWS

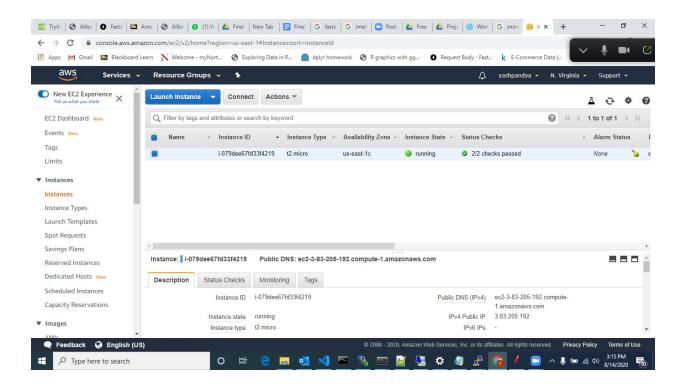
http://ec2-3-83-205-192.compute-1.amazonaws.com:5000/home.html

Flask is a micro web framework written in Python. It is classified as a microframework because it does not require particular tools or libraries. It has no database abstraction layer, form validation, or any other components where pre-existing third-party libraries provide common functions.

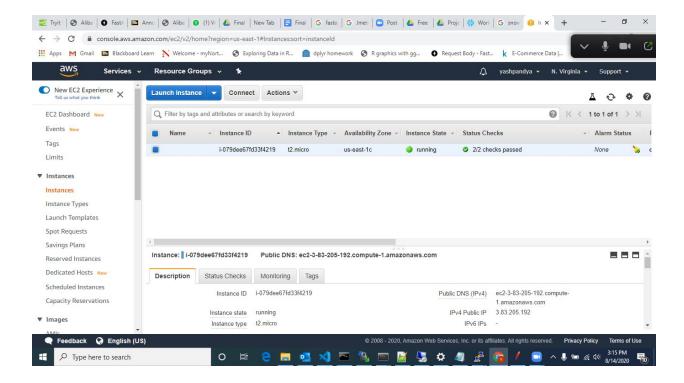
AWS provides on-demand access to scalable web and **application** servers, storage, databases, content delivery, cache, search, and other **application services** that make it easier to build and run apps that deliver a great customer experience.



1) AWS Instance created on AWS account

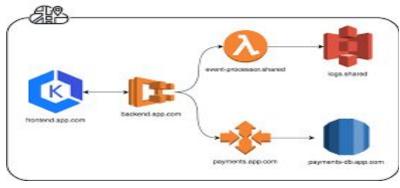


2) File Transferred using WinSCP



3) Used putty for deployment

```
### Occasion | 1.00 | 1.00 | 1.00 | 1.00 | 1.00 | 1.00 | 1.00 | 1.00 | 1.00 | 1.00 | 1.00 | 1.00 | 1.00 | 1.00 | 1.00 | 1.00 | 1.00 | 1.00 | 1.00 | 1.00 | 1.00 | 1.00 | 1.00 | 1.00 | 1.00 | 1.00 | 1.00 | 1.00 | 1.00 | 1.00 | 1.00 | 1.00 | 1.00 | 1.00 | 1.00 | 1.00 | 1.00 | 1.00 | 1.00 | 1.00 | 1.00 | 1.00 | 1.00 | 1.00 | 1.00 | 1.00 | 1.00 | 1.00 | 1.00 | 1.00 | 1.00 | 1.00 | 1.00 | 1.00 | 1.00 | 1.00 | 1.00 | 1.00 | 1.00 | 1.00 | 1.00 | 1.00 | 1.00 | 1.00 | 1.00 | 1.00 | 1.00 | 1.00 | 1.00 | 1.00 | 1.00 | 1.00 | 1.00 | 1.00 | 1.00 | 1.00 | 1.00 | 1.00 | 1.00 | 1.00 | 1.00 | 1.00 | 1.00 | 1.00 | 1.00 | 1.00 | 1.00 | 1.00 | 1.00 | 1.00 | 1.00 | 1.00 | 1.00 | 1.00 | 1.00 | 1.00 | 1.00 | 1.00 | 1.00 | 1.00 | 1.00 | 1.00 | 1.00 | 1.00 | 1.00 | 1.00 | 1.00 | 1.00 | 1.00 | 1.00 | 1.00 | 1.00 | 1.00 | 1.00 | 1.00 | 1.00 | 1.00 | 1.00 | 1.00 | 1.00 | 1.00 | 1.00 | 1.00 | 1.00 | 1.00 | 1.00 | 1.00 | 1.00 | 1.00 | 1.00 | 1.00 | 1.00 | 1.00 | 1.00 | 1.00 | 1.00 | 1.00 | 1.00 | 1.00 | 1.00 | 1.00 | 1.00 | 1.00 | 1.00 | 1.00 | 1.00 | 1.00 | 1.00 | 1.00 | 1.00 | 1.00 | 1.00 | 1.00 | 1.00 | 1.00 | 1.00 | 1.00 | 1.00 | 1.00 | 1.00 | 1.00 | 1.00 | 1.00 | 1.00 | 1.00 | 1.00 | 1.00 | 1.00 | 1.00 | 1.00 | 1.00 | 1.00 | 1.00 | 1.00 | 1.00 | 1.00 | 1.00 | 1.00 | 1.00 | 1.00 | 1.00 | 1.00 | 1.00 | 1.00 | 1.00 | 1.00 | 1.00 | 1.00 | 1.00 | 1.00 | 1.00 | 1.00 | 1.00 | 1.00 | 1.00 | 1.00 | 1.00 | 1.00 | 1.00 | 1.00 | 1.00 | 1.00 | 1.00 | 1.00 | 1.00 | 1.00 | 1.00 | 1.00 | 1.00 | 1.00 | 1.00 | 1.00 | 1.00 | 1.00 | 1.00 | 1.00 | 1.00 | 1.00 | 1.00 | 1.00 | 1.00 | 1.00 | 1.00 | 1.00 | 1.00 | 1.00 | 1.00 | 1.00 | 1.00 | 1.00 | 1.00 | 1.00 | 1.00 | 1.00 | 1.00 | 1.00 | 1.00 | 1.00 | 1.00 | 1.00 | 1.00 | 1.00 | 1.00 | 1.00 | 1.00 | 1.00 | 1.00 | 1.00 | 1.00 | 1.00 | 1.00 | 1.00 | 1.00 | 1.00 | 1.00 | 1.00 | 1.00 | 1.00 | 1.00 | 1.00 | 1.00 | 1.00 | 1.00 | 1.00 | 1.00 | 1.00 | 1.00 | 1.00 | 1.00 | 1.00 | 1.00 | 1.00 | 1.00 | 1.00 | 1.00 | 1.00 | 1.00 | 1.00 | 1.00 | 1.00 | 1.00 | 1.00 | 1.00 | 1.00 | 1.00 | 1.00 | 1.00 | 1.00 | 1.00 | 1.00 | 1
```





Conclusion

We were able to implement various tools we have learned in this course to help support Alibaba's Website analyze their sales and increase their profit.

References

- 1. https://github.com/microsoft/recommenders
- 2. https://fastapi.tiangolo.com/tutorial/body/
- 3. https://github.com/microsoft/recommenders/blob/master/examples/00_quick_start/seque https://github.com/microsoft/recommenders/blob/master/examples/00_quick_start/seque https://github.com/microsoft/recommenders/blob/master/examples/00_quick_start/seque https://github.com/microsoft/recommenders/blob/master/examples/00_quick_start/seque https://github.com/microsoft/recommenders/blob/master/examples/00_quick_start/seque
- 4. https://github.com/microsoft/recommenders/blob/master/examples/02_model_collaborative_filtering/lightgcn_deep_dive.ipynb
- 5. https://github.com/microsoft/recommenders/tree/master/tests
- 6. https://github.com/davidefiocco/streamlit-fastapi-model-serving