

# Friend up Your Cash App Game

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## **About Us**



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# THE WAY FORWARD

## **A**genda

- Building a ML Pipeline with Prefect
- Hands-On: Setup Prefect & GCP
- Introduction to BigQuery
- Hands-On: Upload data to BigQuery via Prefect
- Data Exploration Techniques
- Hands-On: Data Exploration Challenge
- Feature Encoding for ML
- Hands-On: Model Exploration & Embeddings



## **Disclaimer: Imaginary Data Alert!**

The data you're about to see is purely a product of our collective imagination. It's the stuff of dreams, the figment of our data wizards' creativity.







## Github Repo:

https://tinyurl.com/friend-up-your-cash-app-game

#### Collab Notebook:





## **Building an ML Pipeline**

Understanding Data Visualizing Insights Engineering features Algorithm Selection
Training and Validation
Hyperparameter Tuning

Deployment
Input Processing
Output Generation

#### **EXPLORATION**



#### **TRAINING**



#### **INFERENCE**







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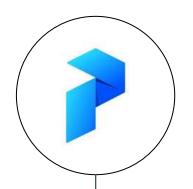






## **Prefect**

Workflow orchestration tool

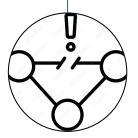




**Streamlined** development



**Scalability** 



Fault tolerance



Versioning



Monitoring





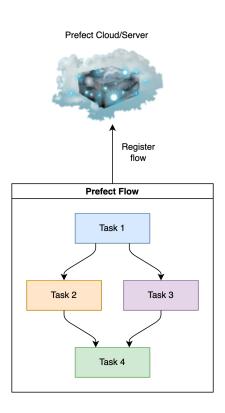
## **Prefect Architecture**

#### Task:

- Represents a single unit of work
- Can be a python function or a callable

#### Flow:

- Collection of tasks in a Directed Acyclic Graph (DAG)
- Represents your ML pipeline
- Monitors workflow runs









## **Setup Prefect**

**Notebook Code:** 



## **Setup GCP**

**Notebook Code:** 



## **Data Infrastructure and Exploration**

Setup GCP Infrastructure Upload Data into BigQuery Explore Data

#### **Notebook Code:**

## **BigQuery**

- A serverless, fully-managed data warehouse by Google Cloud.
- Designed for speed and scalability, analyzing massive datasets effortlessly.

#### Dataset:

- A container that holds tables, views, and other dataset-specific metadata,
- Provides a structured way to organize and manage your data within the platform.

#### BigQuery Table:

- A structured representation of data in BigQuery.
- Organized in rows and columns, with defined schema.
- Supports SQL-like queries and joins for data exploration.







## **Service Accounts**

- A Service Account is a Google Cloud identity
- Used for authenticating applications and services
- Allows controlled access to Google Cloud resources







## What is the Purpose of the Key?

- Confidential piece of information used to securely generate digital signatures and authenticate API requests.
- It ensures data integrity and secure communication between applications and GCP services.









## Create Table and upload data using Prefect

#### **Notebook Code:**

## **Data Exploration**

- Understand patterns, trends, relationships
- Assess data quality (missing, outliers)
- Select relevant features
- Validate assumptions
- Support informed decisions







#### Data Exploration Challenge <a href="https://tinyurl.com/dataexplorationchallenge">https://tinyurl.com/dataexplorationchallenge</a>

#### Basic Stats:

- How many rows are in the dataset?
- How many distinct rows are there?.

#### Example Cash App Usage:

- Find the number of users who used Cash App for less than 1 year.
- Find the number of users who used Cash App for more than 8 years.

#### Transaction Amount:

Calculate the 99th percentile of the transaction amount.

#### Most Interacted Users:

Determine the count of mutual interactions among the most interacted users

#### Feature Correlation:

Discover if any features (columns) are correlated with each other.

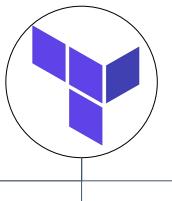




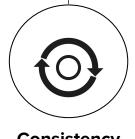


## Infrastructure as Code (laaC)

Managing and Provisioning Infrastructure using Code











Consistency

Reliability







## **Explore Model**



#### **Notebook Code:**

## **Encoding Our features**

#### **Categorical Encoding**

| user_occupation | Encoded value |
|-----------------|---------------|
| Accountant      | 0             |
| Lawyer          | 1             |
| Engineer        | 2             |
|                 |               |
| Doctor          | N             |

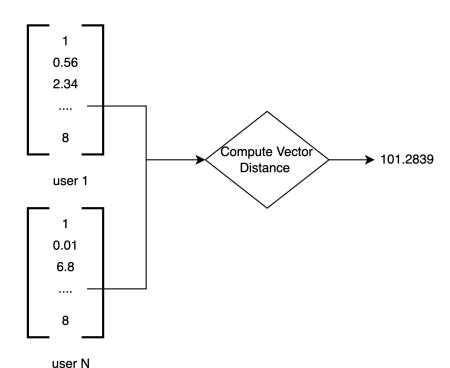
#### **Binary Encoding**

| cash_boost_used | Encoded value |
|-----------------|---------------|
| Yes             | 1             |
| No              | 0             |





## Intro to Vector Similarity



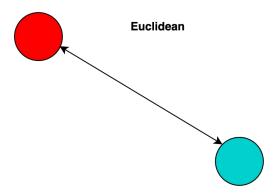




## **Vector Distance Metrics**

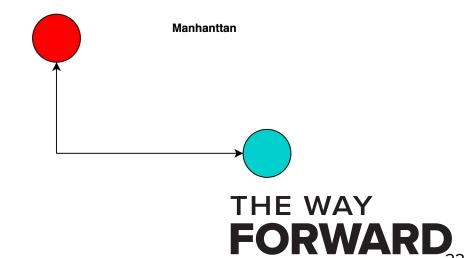
#### **Euclidean Distance**

is the straight-line distance between two points in a space. It's commonly used when the data points have continuous numeric attributes.



#### Manhattan Distance

measures the distance between two points by summing the absolute differences along each dimension.





## Let's compute these distances!

Compute the each of these vector distances between user 1 and the rest of our users in the cash app dataset!

Let's sort the results and return the top 5 most "similar" users according to each metric

#### **Notebook Code:**

https://tinyurl.com/friend-up-your-cash-game-nb

#### **Solution:**

https://tinyurl.com/modelling-solution





#### This doesn't work well ...

#### Why to use embeddings instead:

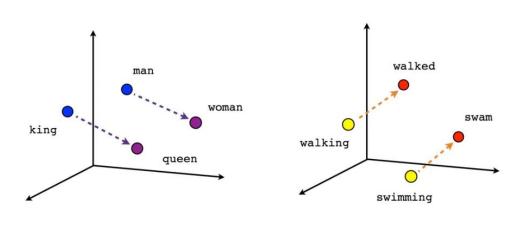
- Our encoded vectors do not adequately represent the data
  - The range of values for each columns is different, for example columns deal with account\_balance skew the data
- Embeddings capture semantic relationships and context, which can enhance the quality of similarity measures.

Let's use a neural network to create embeddings for each of our users!





## Intro to Embeddings



Spain

Italy Madrid

Germany Rome

Berlin

Turkey Ankara

Russia Moscow

Canada Ottawa

Japan Tokyo

Vietnam Hanoi

China Beijing

Male-Female

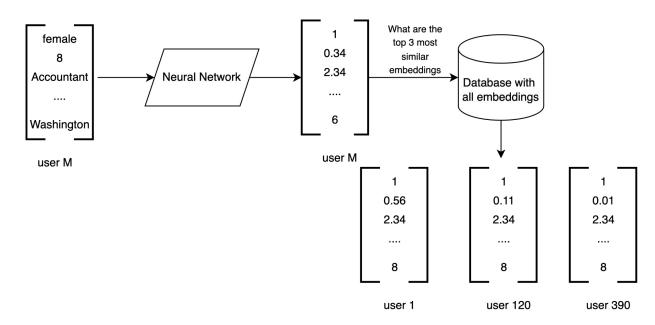
Verb tense

Country-Capital





## Intro to Ranking using Machine Learning







## **Enhance Performance**

- Using Prefect to run your predictions daily
- Using a vector feature store to support storing and computing vector similarities
- Tracking data drift and model metrics using Whylabs









## **Questions?**

# GRACE HOPPER CELEBRATION FORWARD



Feedback survey

## **Thank You**





