


DERIVANDO DADOS NO PROJETO KHIPO

Importando tabelas


```
1 import pandas as pd

1 dfProd=pd.read_csv("product.csv",sep=";")
2 dfProd.head(5)
```




	ProductID	ProviderID	CategoryID	Description
0	99POP	3	1	99POP
1	1	1	5	Taxi Comum
2	2	1	6	Executivo
3	46cec5c4d23e57bcb	4	5	Easy Taxi Corp (~15%)
4	5fc141256dc70a394d0ce4c5c1444dfc	4	5	Taxi

```
1 dfRide=pd.read_csv("ride_v2.csv",sep=";")
2 dfRide.head(1)
```



	RideID	UserID	Schedule	Create	RideStatusID	CompanyID	ProviderID	RideProviderID	price	Updated	C
0	1685755	e15b8cc3-5a67-4630-b89f-ee69f302b582	2025-02-10 14:31:10.8858446	2025-02-10 14:31:10.9084221	1	2	NaN	NaN	0.0	2025-02-10 14:31:10.9084233	

```
1 dfRideAdd=pd.read_csv("rideaddress_v1.csv",sep=";",low_memory=False)
2 dfRideAdd.head(1)
```



	RideAddressID	Address	Street	Number	Neighborhood	City	State	Lat	Lng	RideAddressTypeID	RideID
		Rua									

```
1 dfRideEst=pd.read_csv("rideestimative_v3.csv",sep=";",low_memory=False)
2 dfRideEst.head(1)
```



	RideEstimativeID	RideID	ProductID	WaitingTime	Price	FareID	Selected	RideReasonSelected	EstimativeID	Fee
	8640046	1482000	Flash	0	80.0	c6aaac64-5f89-4fc4-8b66-	0		NaN	0.0

Tratando dados nulos e visualizando tamanho das bases

```
1 print(len(dfProd))
2 print(len(dfRide))
3 print(len(dfRideAdd))
4 print(len(dfRideEst))
5 print(len(dfProd.dropna()))
6 print(len(dfRide.dropna()))
7 print(len(dfRideAdd.dropna()))
8 print(len(dfRideEst.dropna()))
```

↻ 237  
500000  
1000000  
2000000  
237  
14781  
121971  
132236

Criando coluna com dados da semana

```
1 import datetime
2 data=pd.to_datetime(dfRide.iloc[0]["Schedule"])
3 print(data)
4 print(data.day_name())
5 print(data.weekday())
```

↻ 2025-02-10 14:31:10.885844600  
Monday  
0

Criando a coluna dia para receber os dados (dia da semana(segunda= 0))

```
1 dia=[]
2 for i in range(len(dfRide)):
3     data=pd.to_datetime(dfRide.iloc[i]["Schedule"])
4     dia.append(data.weekday())
5
6 dfRide["Dia"]=dia
7 dfRide.head(5)
```

↻

	RideID	UserID	Schedule	Create	RideStatusID	CompanyID	ProviderID	RideProviderID	price	Updated
0	1685755	e15b8cc3-5a67-4630-b89f-ee69f302b582	2025-02-10 14:31:10.8858446	2025-02-10 14:31:10.9084221	1	2	NaN	NaN	0.00	2025-02-10 14:31:10.9084233
1	1685754	5c3fb011-0aea-429a-8305-b88953b77df1	2025-02-10 14:26:35.3411403	2025-02-10 14:26:35.4169873	2	230	5.0	NaN	30.45	2025-02-10 14:28:02.4656963
2	1685753	d7e2f4dc-337f-45f5-b762-67b72b077abc	2025-02-10 14:23:45.2540905	2025-02-10 14:24:32.7058722	2	52	3.0	NaN	11.40	2025-02-10 14:24:46.5037165
3	1685752	2125ed9c-89b8-4df6-9be6-53195397a269	2025-02-10 14:23:12.9838635	2025-02-10 14:23:12.9975475	8	230	36.0	1589157	45.79	2025-02-10 14:30:30.6031123
4	1685751	72cbebf5-5d70-49ab-ab23-8e3b57c7e399	2025-02-10 14:19:30.5937678	2025-02-10 14:19:30.6117184	2	2	3.0	NaN	17.28	2025-02-10 14:24:45.9711764

DIMINUINDO O TAMANHO DA BASE PARA COMEÇAR A EXPLORAR E DERIVAR

```
1 dfRide["RideID"].median()
```

↻ 1434271.5

```
1 len(dfRide.query("RideID<=1300000"))
```

↻ 116782

```
1 dfRide=dfRide.query("RideID<=1300000")
2 dfRideAdd=dfRideAdd.query("RideID<=1300000")
3 dfRideEst=dfRideEst.query("RideID<=1300000")
4 print(len(dfRide))
```

```
5 print(len(dfRideAdd))
6 print(len(dfRideEst))
```

```
116782
233564
905383
```

RENOMIANDO COLUNAS DATAFRAME dfRideAdd TEM A COLUNA RideAddressTypeID QUE INDICA SE O ENDEREÇO É ORIGEM(1) OU DESTINO(2) PODEMOS RENOMEAR PARA OrigDest

```
1 dfRideAdd = dfRideAdd.rename(columns={'RideAddressTypeID': 'OrigDest'})
2 dfRideAdd.head(1)
```

RideAddressID	Address	Street	Number	Neighborhood	City	State	Lat	Lng	OrigDest	RideID
	Rua João									

SELECIONANDO AS COLUNAS QUE QUEREMOS TRABALHAR (dfRideAdd podemos querer trabalhar com RideID, OrigDest, Lat e Lng)

```
1 dfRideAdd=dfRideAdd[["RideID", "OrigDest", "Lat", "Lng"]]
2 dfRideAdd.head(1)
```

	RideID	OrigDest	Lat	Lng
0	1183200	1	-26.329754299999998	-48.840427999999996

Colocando tudo em uma linha Latitude e Longitude Origem e Destino e a distancia Para isso vamos CRIAR dfRideAdd2 (Destino) e tirar o destino de dfRideAdd

```
1 dfRideAdd2=dfRideAdd.query("OrigDest==2")
2 dfRideAdd2.head(3)
```

	RideID	OrigDest	Lat	Lng
1	1183200	2	-26.2554657	-48.6434197
3	1183201	2	-27.4371486	-48.398243099999999
5	1183202	2	-19.936899	-43.9401603

```
1 dfRideAdd=dfRideAdd.query("OrigDest!=2")
2 dfRideAdd.head(3)
```


	RideID	OrigDest	Lat	Lng
0	1183200	1	-26.329754299999998	-48.840427999999996
2	1183201	1	-27.4919788	-48.528287999999996
4	1183202	1	-19.8495799	-44.019915999999995

Renomeando nome novamente as colunas para fazer o merge



```
1 dfRideAdd=dfRideAdd.rename(columns={"Lat":"Lat1", "Lng":"Lng1"})
2 dfRideAdd2=dfRideAdd2.rename(columns={"Lat":"Lat2", "Lng":"Lng2"})
```

Fazendo o join

```
1 dfJoinAdd = pd.merge(dfRideAdd, dfRideAdd2, left_on='RideID', right_on='RideID', how='inner')
2 dfJoinAdd.head(5)
```



	RideID	OrigDest_x	Lat1	Lng1	OrigDest_y	Lat2	Lng2
0	1183200	1	-26.329754299999998	-48.840427999999996	2	-26.2554657	-48.6434197
1	1183201	1	-27.4919788	-48.528287999999996	2	-27.4371486	-48.398243099999999
2	1183202	1	-19.8495799	-44.019915999999995	2	-19.936899	-43.9401603
3	1183203	1	-23.9624233	-46.254657599999999	2	-23.8373074	-46.1321725
4	1183204	1	-10.9198019	-37.077441799999995	2	-10.9071288	-37.0877194



Lógica de distância

```
1 !pip install geopy


Requirement already satisfied: geopy in /usr/local/lib/python3.11/dist-packages (2.4.1)
Requirement already satisfied: geographiclib<3,>=1.52 in /usr/local/lib/python3.11/dist-packages (from geopy) (2.0)

1 from geopy.distance import geodesic
2
3 # Pontos: (latitude, longitude)
4 ponto1 = (-27.4919788, -48.528287999999996)
5 ponto2 = (-27.4371486, -48.398243099999999)
6
7 # Distância em km
8 distancia_km = geodesic(ponto1, ponto2).kilometers
9 print(f"Distância: {distancia_km:.2f} km")



Distância: 14.22 km
```

Adicionando a distância em toda a tabela

```
1 Dist=[]
2
3 for i in range(len(dfJoinAdd)):
4     lat1=float(dfJoinAdd.iloc[i]["Lat1"].replace(",","."))
5     lng1=float(dfJoinAdd.iloc[i]["Lng1"].replace(",","."))
6     lat2=float(dfJoinAdd.iloc[i]["Lat2"].replace(",","."))
7     lng2=float(dfJoinAdd.iloc[i]["Lng2"].replace(",","."))
8     ponto1 = (lat1, lng1)
9     ponto2 = (lat2, lng2)
10    distkm = geodesic(ponto1, ponto2).kilometers
11    Dist.append(distkm)
12
13 dfJoinAdd["Distancia"]=Dist
14
15 dfJoinAdd.head(5)
```




	RideID	OrigDest_x	Lat1	Lng1	OrigDest_y	Lat2	Lng2	Distancia
0	1183200	1	-26.329754299999998	-48.840427999999996	2	-26.2554657	-48.6434197	21.327087
1	1183201	1	-27.4919788	-48.528287999999996	2	-27.4371486	-48.398243099999999	14.217724
2	1183202	1	-19.8495799	-44.019915999999995	2	-19.936899	-43.9401603	12.774728
3	1183203	1	-23.9624233	-46.254657599999999	2	-23.8373074	-46.1321725	18.643943
4	1183204	1	-10.9198019	-37.077441799999995	2	-10.9071288	-37.0877194	1.796511



Separando o dia da semana vinculando pelo RideID Já usando o melhor modo de treinar a ML.

```
1 dfRide3 = dfRide[["RideID","Dia"]]
2 print(dfRide3)
```




	RideID	Dia
383218	1300000	3
383219	1299999	3
383220	1299998	3
383221	1299997	3
383222	1299996	3
...	...	...
499995	1183204	1

```
499996 1183203 1
499997 1183202 1
499998 1183201 1
499999 1183200 1
```

[116782 rows x 2 columns]

```
1 dfJoinAdd2 = pd.merge(dfJoinAdd, dfRide3, left_on = 'RideID', right_on = 'RideID', how='inner')
2 dfJoinAdd2.head(5)
```



	RideID	OrigDest_x	Lat1	Lng1	OrigDest_y	Lat2	Lng2	Distancia	Dia
0	1183200	1	-26.329754299999998	-48.840427999999996	2	-26.2554657	-48.6434197	21.327087	1
1	1183201	1	-27.4919788	-48.528287999999996	2	-27.4371486	-48.398243099999999	14.217724	1
2	1183202	1	-19.8495799	-44.019915999999995	2	-19.936899	-43.9401603	12.774728	1
3	1183203	1	-23.9624233	-46.254657599999999	2	-23.8373074	-46.1321725	18.643943	1
4	1183204	1	-10.9198019	-37.077441799999995	2	-10.9071288	-37.0877194	1.796511	1

Selecionando os produtos para treinar a ML e imprimindo a quantidade de corridas para treinamento.


```
1 dfprodutos = dfRideEst.query("ProductID in ['UberX', '99POP', 'Flash', 'Confort', 'pop99','poupa99', 'Bag', 'UberFlash', '', '', '7da40a"]
2 print(len(dfprodutos))
```



421534

Separando o preço por RideID


```
1 dftprodutos = dfprodutos[["RideID","Price"]]
2 print(dftprodutos)
```



	RideID	Price
0	1183200	89.00
1	1183200	89.00
3	1183200	170.21
4	1183200	170.21
7	1183201	31.50
...	...	...
905373	1299999	51.50
905376	1300000	20.33
905377	1300000	23.62
905380	1300000	20.00
905382	1300000	20.00

[421534 rows x 2 columns]

```
1 dftprodutos1 = dfRideEst[["RideID","ProductID"]]
2 print(dftprodutos1)
```



	RideID	ProductID
0	1183200	Flash
1	1183200	UberX
2	1183200	Comfort
3	1183200	poupa99
4	1183200	pop99
...	...	...
905378	1300000	turbo-taxi
905379	1300000	regular-taxi
905380	1300000	Flash
905381	1300000	Comfort
905382	1300000	UberX

[905383 rows x 2 columns]

```
1 dfJoinAdd3 = pd.merge(dfJoinAdd2, dftprodutos, left_on = 'RideID', right_on = 'RideID', how='inner')
2 dfJoinAdd3.head(5)
```

	RideID	OrigDest_x	Lat1	Lng1	OrigDest_y	Lat2	Lng2	Distancia	Dia	Price
0	1183200	1	-26.329754299999998	-48.840427999999996	2	-26.2554657	-48.6434197	21.327087	1	89.00
1	1183200	1	-26.329754299999998	-48.840427999999996	2	-26.2554657	-48.6434197	21.327087	1	89.00
2	1183200	1	-26.329754299999998	-48.840427999999996	2	-26.2554657	-48.6434197	21.327087	1	170.21
3	1183200	1	-26.329754299999998	-48.840427999999996	2	-26.2554657	-48.6434197	21.327087	1	170.21
4	1183201	1	-27.4919788	-48.528287999999996	2	-27.4371486	-48.398243099999999	14.217724	1	31.50

```

1 dfRide['schedule'] = pd.to_datetime(dfRide['Schedule'])
2 dfRide['schedule_time'] = dfRide['schedule'].dt.time
3 dfRide.head(1)

```

	RideID	UserID	Schedule	Create	RideStatusID	CompanyID	ProviderID	RideProviderID	price	Update
383218	1300000	acbb3e9f-62a8-4f01-b04f-f58058728fc4	2021-09-30 16:01:11.5153413	2021-09-30 16:01:11.5488417	2	40	NaN	NaN	20.0	2021-09-30 16:01:23.458564

Separando a hora da coluna schedule da dfRideId

```

1 dfRideT=dfRide[["RideID","schedule_time"]]
2 dfRideT.head(1)

```

	RideID	schedule_time
383218	1300000	16:01:11.515341

Fazendo merge da schedulo(hora de inicio)

```

1 dfJoinAdd3 = pd.merge(dfJoinAdd3, dfRideT, left_on = 'RideID', right_on = 'RideID', how='inner')
2 dfJoinAdd3.head(1)

```

	RideID	OrigDest_x	Lat1	Lng1	OrigDest_y	Lat2	Lng2	Distancia	Dia	Price	schedule_tim
0	1183200	1	-26.329754299999998	-48.840427999999996	2	-26.2554657	-48.6434197	21.327087	1	89.0	10:09:45.62817

```

1 dfJoinAdd3 = pd.merge(dfJoinAdd3, dftprodutos1, left_on = 'RideID', right_on = 'RideID', how='inner')
2 dfJoinAdd3.head(1)

```

	RideID	OrigDest_x	Lat1	Lng1	OrigDest_y	Lat2	Lng2	Distancia	Dia	Price	schedule_tim
0	1183200	1	-26.329754299999998	-48.840427999999996	2	-26.2554657	-48.6434197	21.327087	1	89.0	10:09:45.62817

## ✓ Criação do banco de dados

Extrair a base pronta para ML

```
1 dfJoinAdd3.to_csv('dfJoinAdd3.csv', index=False)
```

```

1 dfJoinAdd3=pd.read_csv("dfJoinAdd3.csv",sep=",")
2 dfJoinAdd3.head(2)

```

```
1 import sqlite3
```

```

1 conn = sqlite3.connect("base_corridas")
2 cursor = conn.cursor()

```

## ✓ Criação da tabela no banco de Dados (não executar)

```

1 cursor.execute('''
2 DROP TABLE Corridas
3 ''')
4
5 conn.commit()

1 cursor.execute('''
2 CREATE TABLE Corridas (
3     RideID INT,
4     OrigDest_x INT,
5     Lat1 REAL,
6     Lng1 REAL,
7     OrigDest_y INT,
8     Lat2 REAL,
9     Lng2 REAL,
10    Distancia REAL,
11    Dia INT,
12    Price DECIMAL(10, 2),
13    schedule_time TIME
14 )
15 ''')
16
17 conn.commit()

```

#### Populando a tabela com os dados

```

1 strSQL = '''
2 INSERT INTO Corridas (RideID, OrigDest_x, Lat1, Lng1, OrigDest_y, Lat2, Lng2, Distancia, Dia, Price, schedule_time, ProductID)
3 VALUES (?, ?, ?, ?, ?, ?, ?, ?, ?, ?, ?)'''
4
5
6 dfJoinAdd3.to_sql('Corridas', conn, if_exists='append', index=False)

1 dfJoinAdd3 = pd.read_sql_query("SELECT * FROM Corridas", conn)
2 dfJoinAdd3.head(2)

```

	RideID	OrigDest_x	Lat1	Lng1	OrigDest_y	Lat2	Lng2	Distancia	Dia	Price	schedule_tim
0	1183200	1	-26.329754299999998	-48.840427999999996	2	-26.2554657	-48.6434197	21.327087	1	89.0	10:09:45.62817
1	1183200	1	-26.329754299999998	-48.840427999999996	2	-26.2554657	-48.6434197	21.327087	1	89.0	10:09:45.62817

#### Transformando schedule em horas

```

1 from datetime import datetime
2 dfJoinAdd3 = pd.read_sql_query("SELECT * FROM Corridas", conn)
3 X = dfJoinAdd3[["OrigDest_x", "Lat1", "Lng1", "OrigDest_y", "Lat2", "Lng2", "Distancia", "Dia", "schedule_time", "ProductID"]]
4 y = dfJoinAdd3["Price"]
5 time_str = X.iloc[0]['schedule_time']
6 # Converter para objeto time
7 time_obj = datetime.strptime(time_str, "%H:%M:%S.%f").time()

```

Para testar inclua o BD(base corridas) e Data frame (dfJoinAdd3.cv)

#### Treinamento da Machine Learning

```

1 from sklearn.model_selection import train_test_split
2 from sklearn.ensemble import RandomForestRegressor
3 from sklearn.metrics import mean_absolute_error
4 from sklearn.metrics import mean_squared_error
5 from sklearn.metrics import r2_score
6 from datetime import datetime
7 import pandas as pd
8
9
10 def strtotime(time_str):
11     time_obj = datetime.strptime(time_str, "%H:%M:%S.%f").time()
12     return time_obj

```

```

13
14 dfJoinAdd3 = pd.read_sql_query("SELECT * FROM Corridas", conn)
15
16 dfJoinAdd3 = dfJoinAdd3.sample(frac=0.5) #reduz em 50% da base para teste
17
18 X = dfJoinAdd3[["OrigDest_x","Lat1","Lng1","OrigDest_y","Lat2","Lng2","Distancia","Dia","schedule_time"]]
19 y = dfJoinAdd3["Price"]
20
21 SEED = 20
22
23 treino_x, teste_x, treino_y, teste_y = train_test_split(X, y, test_size=0.3, random_state=SEED) # Linha adicionada
24 #coluna limpa:
25
26 # Convertendo colunas de latitude e longitude para o formato correto
27 for col in ["Lat1", "Lng1", "Lat2", "Lng2"]:
28     treino_x[col] = treino_x[col].astype(str).str.replace(',', '.').astype(float)
29     teste_x[col] = teste_x[col].astype(str).str.replace(',', '.').astype(float)
30
31 # Convertendo 'schedule_time' para segundos desde a meia-noite
32 treino_x['schedule_time'] = treino_x['schedule_time'].apply(lambda x: strptime(x).hour * 3600 + strptime(x).minute * 60 + strptime(x).second)
33 teste_x['schedule_time'] = teste_x['schedule_time'].apply(lambda x: strptime(x).hour * 3600 + strptime(x).minute * 60 + strptime(x).second)
34
35 #treino_x = pd.get_dummies(treino_x, columns=['ProductID'], prefix=['ProductID'])
36 #teste_x = pd.get_dummies(teste_x, columns=['ProductID'], prefix=['ProductID'])
37
38 # Alinhar as colunas após o get_dummies
39 treino_x, teste_x = treino_x.align(teste_x, join='left', axis=1, fill_value=0)
40
41 # Agora você pode treinar o modelo:
42 model = RandomForestRegressor(n_estimators=100, random_state=SEED)
43 model.fit(treino_x, treino_y)
44
45 predicao_y = model.predict(teste_x)
46 r2 = r2_score(teste_y, predicao_y)
47 mse = mean_squared_error(teste_y, predicao_y)
48 mae = mean_absolute_error(teste_y, predicao_y)
49
50 print(f"Erro médio absoluto: {mae}")
51 print(f"Erro quadrático médio: {mse}")

```

Erro médio absoluto: 4.989349073690697  
 Erro quadrático médio: 107.83222191258031

```

1 from sklearn.ensemble import GradientBoostingRegressor
2
3 # 2 - Definindo o modelo:
4 modelo = GradientBoostingRegressor()
5 modelo.fit(treino_x, treino_y)

```

▼ GradientBoostingRegressor ⓘ ?

GradientBoostingRegressor()

```

1 from sklearn.metrics import r2_score
2 from sklearn.metrics import mean_squared_error
3 from sklearn.metrics import mean_absolute_error
4
5 # 2 - Aplicação do Teste:
6 #predicao_y = modelo.predict(teste_x)
7 predicao_y = modelo.predict(teste_x)
8
9 # 3 - Avaliação do Teste (com todos os modelos de Metrics):
10 r2 = r2_score(teste_y, predicao_y)
11
12 mse = mean_squared_error(teste_y, predicao_y)
13 mae = mean_absolute_error(teste_y, predicao_y)
14
15 # 4 - Mostrar os resultados:
16
17 print(f"R2 (Acuracia): {r2*100:.2f}%")
18 print(f"MSE : {mse:.2f}")
19 print(f"MAE : {mae:.2f}")

```

R2 (Acuracia): 89.73%  
 MSE : 192.00  
 MAE : 7.44



## ✓ Logica Final pedindo endereço

```
1 !pip install --upgrade WazeRouteCalculator
```

```
Collecting WazeRouteCalculator
  Downloading WazeRouteCalculator-0.15-py3-none-any.whl.metadata (448 bytes)
Requirement already satisfied: requests in /usr/local/lib/python3.11/dist-packages (from WazeRouteCalculator) (2.32.3)
Requirement already satisfied: charset-normalizer<4,>=2 in /usr/local/lib/python3.11/dist-packages (from requests->WazeRouteCalculator) (3.4.1)
Requirement already satisfied: idna<4,>=2.5 in /usr/local/lib/python3.11/dist-packages (from requests->WazeRouteCalculator) (3.10)
Requirement already satisfied: urllib3<3,>=1.21.1 in /usr/local/lib/python3.11/dist-packages (from requests->WazeRouteCalculator) (2.4.0)
Requirement already satisfied: certifi>=2017.4.17 in /usr/local/lib/python3.11/dist-packages (from requests->WazeRouteCalculator) (2025.11.11)
Downloading WazeRouteCalculator-0.15-py3-none-any.whl (16 kB)
Installing collected packages: WazeRouteCalculator
Successfully installed WazeRouteCalculator-0.15
```

```
1 from WazeRouteCalculator import WazeRouteCalculator, WRCError
2 from geopy.geocoders import Nominatim
3 from datetime import datetime
4 import pandas as pd
5 import warnings
6 import unicodedata
7
8 warnings.filterwarnings("ignore")
9
10 # Inicializa geolocalizador
11 geolocator = Nominatim(user_agent="predictor-app")
12
13 def limpar_endereco(endereco):
14     nfkd = unicodedata.normalize('NFKD', endereco)
15     return u"".join([c for c in nfkd if not unicodedata.combining(c)])
16
17 def calcular_distancia_waze(origem, destino, region='EU'):
18     try:
19         lat1, lng1 = endereco_para_coordenadas(origem)
20         lat2, lng2 = endereco_para_coordenadas(destino)
21
22         origem_coord = f"{lat1},{lng1}"
23         destino_coord = f"{lat2},{lng2}"
24
25         rota = WazeRouteCalculator(origem_coord, destino_coord, region)
26         tempo_min, distancia_km = rota.calc_route_info(real_time=False) # ou real_time=True, se quiser tráfego atual
27
28         return distancia_km, tempo_min
29     except WRCError as e:
30         raise ValueError(f"Erro ao calcular rota Waze: {e}")
31
32
33 def endereco_para_coordenadas(endereco):
34     endereco_limpo = limpar_endereco(endereco)
35     location = geolocator.geocode(endereco_limpo, timeout=15)
36     if location:
37         return location.latitude, location.longitude
38     else:
39         raise ValueError(f"Não foi possível geocodificar o endereço: {endereco}")
40
41 def gerar_features(end_origem, end_destino, horario=None):
42     lat1, lng1 = endereco_para_coordenadas(end_origem)
43     lat2, lng2 = endereco_para_coordenadas(end_destino)
44     distancia, tempo = calcular_distancia_waze(end_origem, end_destino)
45
46     now = horario or datetime.now()
47     schedule_time = now.hour * 3600 + now.minute * 60 + now.second
48     dia = now.weekday()
49
50     orig_dest_x = abs(hash(end_origem)) % (10 ** 8)
51     orig_dest_y = abs(hash(end_destino)) % (10 ** 8)
52
53     df = pd.DataFrame([
54         "OrigDest_x": orig_dest_x,
55         "Lat1": lat1,
56         "Lng1": lng1,
57         "OrigDest_y": orig_dest_y,
58         "Lat2": lat2,
59         "Lng2": lng2,
```

```

60     "Distancia": distancia,
61     "Tempo": tempo,
62     "Dia": dia,
63     "schedule_time": schedule_time
64 })
65
66 return df
67
68 def prever_preco(end_origem, end_destino):
69     dados = gerar_features(end_origem, end_destino)
70     dados_para_prever = dados.drop(columns=["Tempo"])
71     preco_estimado = model.predict(dados_para_prever)[0]
72
73     dados["Preco_Previsto"] = preco_estimado
74     dados["Endereco_Origem"] = end_origem
75     dados["Endereco_Destino"] = end_destino
76     return dados
77
78 # =====
79
80 if __name__ == "__main__":
81     print("=== Previsão de Preço Uber ===")
82     origem = input("Digite o endereço de ORIGEM: ").strip()
83     destino = input("Digite o endereço de DESTINO: ").strip()
84
85     try:
86         resultado = prever_preco(origem, destino)
87         preco = resultado["Preco_Previsto"].values[0]
88         distancia = resultado["Distancia"].values[0]
89         tempo = resultado["Tempo"].values[0]
90
91         preco_confort = preco * 1.25
92         preco_black = preco_confort * 1.24
93
94         print(f"\n🚗 Preço estimado da corrida (Ubex): R$ {preco:.2f}")
95         print(f"\n🚗 Preço estimado (Uber Confort): R$ {preco_confort:.2f}")
96         print(f"\n🚗 Preço estimado (Uber Black): R$ {preco_black:.2f}")
97         print(f"\n📍 Distância estimada: {distancia:.2f} km")
98         print(f"\n🕒 Tempo estimado: {tempo:.0f} minutos")
99
100     except Exception as e:
101         print(f"\n❌ Erro ao calcular o preço: {e}")
102

```

→ === Previsão de Preço Uber ===  
 Digite o endereço de ORIGEM: Rua Itajai 125 - Sao Paulo  
 Digite o endereço de DESTINO: Av Paulista 2000 - Sao Paulo

🚗 Preço estimado da corrida (Ubex): R\$ 36.29

🚗 Preço estimado (Uber Confort): R\$ 45.37

🚗 Preço estimado (Uber Black): R\$ 56.25

📍 Distância estimada: 7.99 km

🕒 Tempo estimado: 17 minutos