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import numpy as np
import pandas as pd
import plotly.express as px
import matplotlib.pyplot as plt
import seaborn as sns
from sklearn.preprocessing import OneHotEncoder, StandardScaler
from sklearn.model_selection import train_test_split
from sklearn import metrics
from sklearn.pipeline import make_pipeline
from sklearn.ensemble import RandomForestRegressor
import warnings
warnings.filterwarnings('ignore')
# Importing Raw Files
train_raw = pd.read_csv('train.csv')
test_raw = pd.read_csv('test.csv')
meal = pd.read_csv('meal_info.csv')
centerinfo = pd.read_csv('fulfilment_center_info.csv')
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# Analyzing Data
print("The Shape of Demand dataset :", train_raw.shape)
print("The Shape of Fulfillment Center Information dataset :", centerinfo.shape)
print("The Shape of Meal information dataset :", meal.shape)
print("The Shape of Test dataset :", test raw.shape)
train_raw.head()
centerinfo.head()
meal.head()
test raw.head()
# Check for missing values
train_raw.isnull().sum().sum()
test_raw.isnull().sum().sum()
# Analysis report
print("The company has", centerinfo["center_id"].nunique(), " warehouse ", "spreed
into ",
      centerinfo["city code"].nunique(), "City and ",
centerinfo["region_code"].nunique(), "Regions")
print("The products of the company are ", meal["meal_id"].nunique(), "unique meals
, divided into ",
      meal["category"].nunique(), "category and ", meal["cuisine"].nunique(),
"cuisine")
# Merge meal, center-info data with train and test data
train = pd.merge(train_raw, meal, on="meal_id", how="left")
train = pd.merge(train, centerinfo, on="center_id", how="left")
print("Shape of train data : ", train.shape)
train.head()
# Merge test data with meal and center info
test = pd.merge(test_raw, meal, on="meal_id", how="outer")
test = pd.merge(test, centerinfo, on="center id", how="outer")
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print("Shape of test data : ", test.shape)
test.head()
# Typecasting to assign appropriate data type to variables
col_names = ['center_id', 'meal_id', 'category', 'cuisine', 'city_code',
'region_code', 'center_type']
train[col_names] = train[col_names].astype('category')
test[col_names] = test[col_names].astype('category')
print("Train Datatype\n", train.dtypes)
print("Test Datatype\n", test.dtypes)
# Orders by centers
center_orders = train.groupby("center_id", as_index=False).sum()
center_orders = center_orders[["center_id",
"num_orders"]].sort_values(by="num_orders", ascending=False).head(10)
fig = px.bar(x=center_orders["center_id"].astype("str"),
y=center_orders["num_orders"], title="Top 10 Centers by Order",
             labels={"x": "center_id", "y": "num_orders"})
fig.show()
# Pie chart on food category
fig = px.pie(values=train["category"].value_counts(),
names=train["category"].unique(),
             title="Most popular food category")
fig.show()
# Orders by Cuisine types
cuisine_orders = train.groupby(["cuisine"], as_index=False).sum()
cuisine orders = cuisine orders[["cuisine",
"num orders"]].sort values(by="num orders", ascending=False)
fig = px.bar(cuisine_orders, x="cuisine", y="num_orders", title="orders by
cuisine")
fig.show()
# Impact of check-out price on order
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train_sample = train.sample(frac=0.2)
fig = px.scatter(train_sample, x="checkout_price", y="num_orders", title="number of
order change with checkout price")
fig.show()
sns.boxplot(train["checkout_price"])
# Orders weekly trend
week_orders = train.groupby(["week"], as_index=False).sum()
week_orders = week_orders[["week", "num_orders"]]
fig = px.line(week_orders, x="week", y="num_orders", markers=True, title="Order
weekly trend")
fig.show()
# Deriving discount percent and discount y/n
train['discount percent'] = ((train['base_price'] - train['checkout_price']) /
train['base_price']) * 100
# Discount Y/N
train['discount y/n'] = [1 if x > 0 else 0 for x in (train['base_price'] - 1 train['discount y/n'] = [1 if x > 0 else 0 for x in (train['base_price'] - 1 train['discount y/n'] = [1 if x > 0 else 0 for x in (train['base_price'] - 1 train['base_price'] - 1 train['base_p
train['checkout_price'])]
# Creating same feature in test dataset
test['discount percent'] = ((test['base_price'] - test['checkout_price']) /
test['base_price']) * 100
test['discount y/n'] = [1 if x > 0 else 0 for x in (test['base_price'] -
test['checkout_price'])]
train.head(2)
# Check for correlation between numeric features
plt.figure(figsize=(13, 13))
sns.heatmap(train.corr(), linewidths=.1, cmap='Reds', annot=True)
plt.title('Correlation Matrix')
plt.show()
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# Define One hot encoding function
def one_hot_encode(features_to_encode, dataset):
    encoder = OneHotEncoder(sparse=False)
    encoder.fit(dataset[features_to_encode])
    encoded cols = pd.DataFrame(encoder.transform(dataset[features to encode]),
columns=encoder.get_feature_names())
    dataset = dataset.drop(columns=features_to_encode)
    for cols in encoded_cols.columns:
        dataset[cols] = encoded_cols[cols]
    return dataset
# get list of categorical variables in data set
ls = train.select_dtypes(include='category').columns.values.tolist()
# Run one-hot encoding on all categorical variables
features to encode = 1s
data = one_hot_encode(features_to_encode, train)
data = data.reset_index(drop=True)
# Train-Validation Data Split
y = data[["num_orders"]]
X = data.drop(["num_orders", "id", "base_price", "discount y/n"], axis=1)
X = X.replace((np.inf, -np.inf, np.nan), 0) # replace nan and infinity values with
# 20% of train data is used for validation
X_train, X_val, y_train, y_val = train_test_split(X, y, test_size=0.20,
random_state=100)
# Prepare test data post applying onehot encoding
OH_test = one_hot_encode(features_to_encode, test)
test_final = OH_test.drop(["id", "base_price", "discount y/n"], axis=1)
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# Create pipeline for scaling and modeling
RF_pipe = make_pipeline(StandardScaler(), RandomForestRegressor(n_estimators=100,
max_depth=7))
# Build Model
RF_pipe.fit(X_train, y_train)
# Predict Value
RF_train_y_pred = RF_pipe.predict(X_val)
# Model Evaluation-
print('R Square:', RF_pipe.score(X_val, y_val))
print('RMSLE:', 100 * np.sqrt(metrics.mean_squared_log_error(y_val,
RF_train_y_pred)))
# Applying algorithm to predict orders
test_y_pred = RF_pipe.predict(test_final)
Result = pd.DataFrame(test_y_pred)
print(Result.values)
Result = pd.DataFrame(test_y_pred)
Submission = pd.DataFrame(columns=['id', 'num_orders'])
Submission['id'] = test['id']
Submission['num_orders'] = Result.values
Submission.to_csv('My submission.csv', index=False)
print(Submission.shape)
print(Submission.head())
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