## Product Demand Prediction with Machine Learning

## **Product Demand Prediction using Python**

Let's start by importing the necessary Python libraries and the dataset we need for the task of product demand prediction:

```
# Import necessary libraries
import pandas as pd
import numpy as np
import matplotlib.pyplot as plt
from sklearn.model selection import train test split
from sklearn.ensemble import RandomForestRegressor
from sklearn.metrics import mean squared error
import statsmodels.api as sm
# Load your dataset (assuming it has columns like 'date', 'quantity sold',
'price', 'promotion', etc.)
data = pd.read csv('product sales.csv')
# Data preprocessing and feature engineering
# Extract relevant date features
data['date'] = pd.to_datetime(data['date'])
data['year'] = data['date'].dt.year
data['month'] = data['date'].dt.month
data['day of week'] = data['date'].dt.dayofweek
```

```
# Split data into features (X) and target (y)
X = data[['year', 'month', 'day of week', 'price', 'promotion']]
y = data['quantity_sold']
# Split data into training and testing sets
X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.2,
random_state=42)
# Initialize and train a Random Forest regressor
rf model = RandomForestRegressor(n estimators=100, random state=42)
rf model.fit(X train, y train)
# Make predictions on the test set
y_pred_rf = rf_model.predict(X_test)
# Evaluate the Random Forest model
mse_rf = mean_squared_error(y_test, y_pred_rf)
print(f"Random Forest Mean Squared Error: {mse rf}")
# Optionally, you can visualize feature importances from the Random Forest
model
feature_importances = rf_model.feature_importances_
features = X.columns
plt.bar(features, feature importances)
plt.xlabel('Features')
plt.ylabel('Importance')
plt.title('Feature Importances')
```

```
plt.show()
# If you want to explore time series analysis, you can use the statsmodels
library
# Fit a time series model (e.g., SARIMA) to the data
model = sm.tsa.SARIMAX(y_train, order=(1, 1, 1), seasonal_order=(1, 1, 1, 12))
results = model.fit()
# Make predictions using the time series model
y pred ts = results.predict(start=len(y train), end=len(y train) + len(y test) -
1, dynamic=False)
# Evaluate the time series model
mse_ts = mean_squared_error(y_test, y_pred_ts)
print(f"Time Series Mean Squared Error: {mse_ts}")
# Use the models for demand forecasting as needed
# You can input future features to predict future demand
# Close any open figures from the Random Forest feature importances plot
plt.close()
```