

Urban Data Analysis

Week 12: Time Series Analysis & Forecasting

Today's Agenda



- INTRODUCTION
TO TIME SERIES
DATA



- KEY
COMPONENTS OF
TIME SERIES



- FORECASTING
WITH
STATSMODELS



- MODEL
EVALUATION
METRICS



- URBAN MOBILITY
FORECASTING
CASE STUDY

What is Time Series Analysis?



TIME SERIES DATA: DATA POINTS
COLLECTED OR RECORDED AT
SPECIFIC TIME INTERVALS



EXAMPLES: SUBWAY RIDERSHIP,
DAILY AIR QUALITY

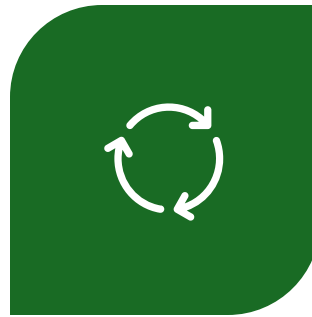


GOAL: UNDERSTAND PATTERNS
AND FORECAST FUTURE VALUES

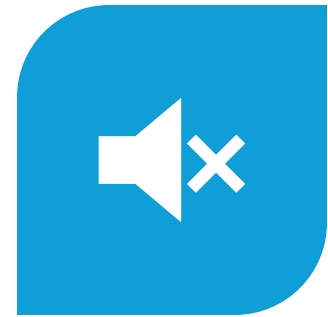
Time Series Components



1. TREND – LONG-TERM MOVEMENT



2. SEASONALITY – REPEATING CYCLES



3. NOISE – RANDOM FLUCTUATIONS

Urban Use Case Example



EXAMPLE:



CAN WE FORECAST NEXT
WEEK'S SUBWAY RIDERSHIP
USING PAST DAILY DATA?



USE GTFS OR MTA
TURNSTILE DATA

Step 1: Import and Plot



`import pandas as pd`



`import matplotlib.pyplot as plt`



```
ridership =  
pd.read_csv("daily_subway_ridership.csv"  
, parse_dates=['date'], index_col='date')
```



```
ridership['entries'].plot(figsize=(12,5),  
title='Daily Subway Entries')
```

Step 2: Decompose Time Series

```
from statsmodels.tsa.seasonal import  
seasonal_decompose
```

```
result =  
seasonal_decompose(ridership['entries'],  
model='additive', period=7)
```

```
result.plot()
```

Step 3: Forecast with ARIMA

```
from statsmodels.tsa.arima.model  
import ARIMA
```

```
model = ARIMA(ridership['entries'],  
order=(1,1,1))
```

```
fit = model.fit()
```

```
forecast = fit.forecast(steps=7)
```


Step 4: Plot Forecast



```
plt.plot(ridership.index[-30:],  
ridership['entries'][-30:], label='Actual')
```



```
plt.plot(pd.date_range(start=ridership.index[-  
1], periods=8, freq='D')[1:], forecast,  
label='Forecast')
```

Step 5: Evaluation Metrics

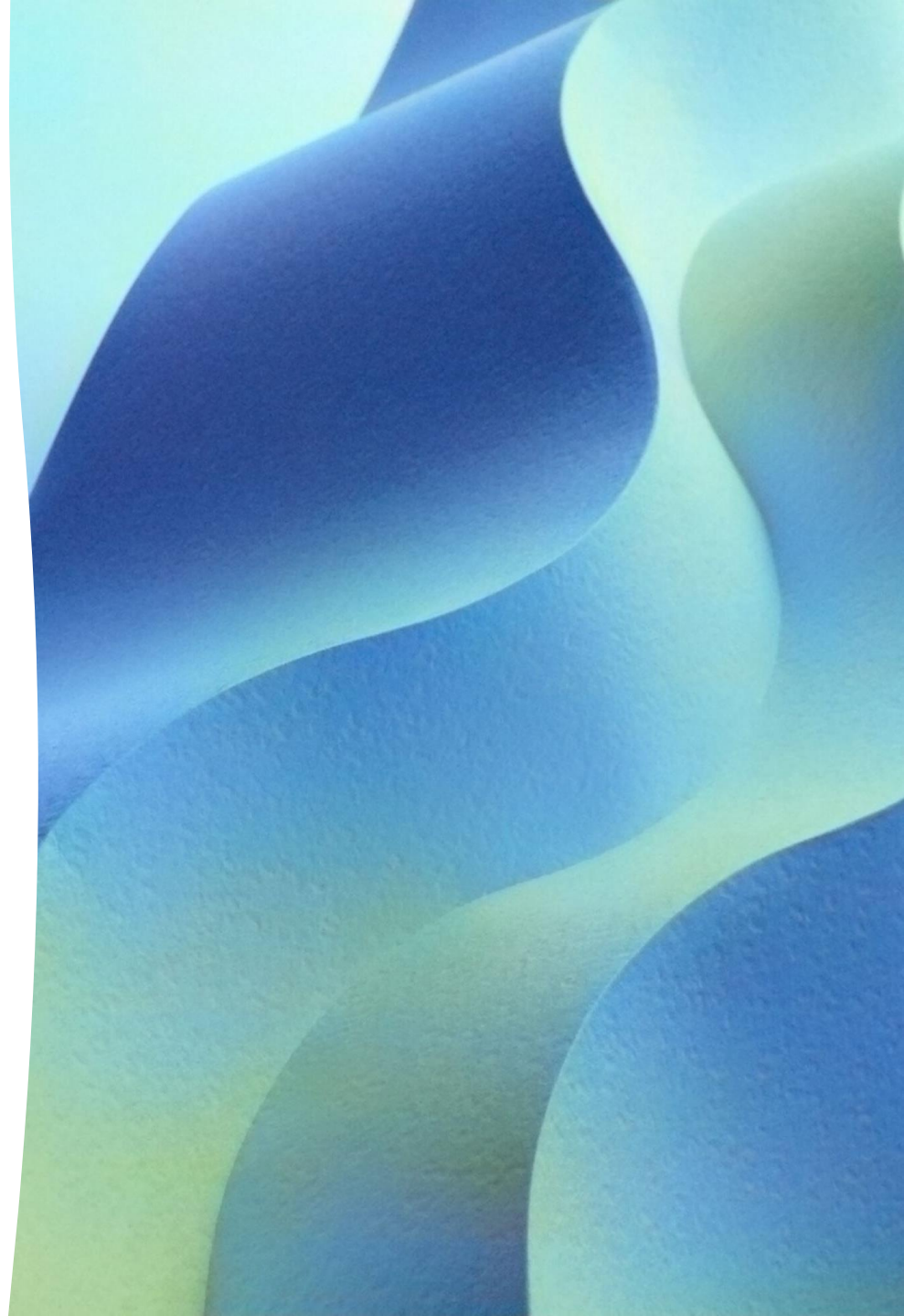
MAE, RMSE, MAPE

```
from sklearn.metrics import  
mean_absolute_error, mean_squared_error
```

```
import numpy as np
```

```
mae = mean_absolute_error(actual,  
forecast)
```

```
rmse = np.sqrt(mean_squared_error(actual,  
forecast))
```

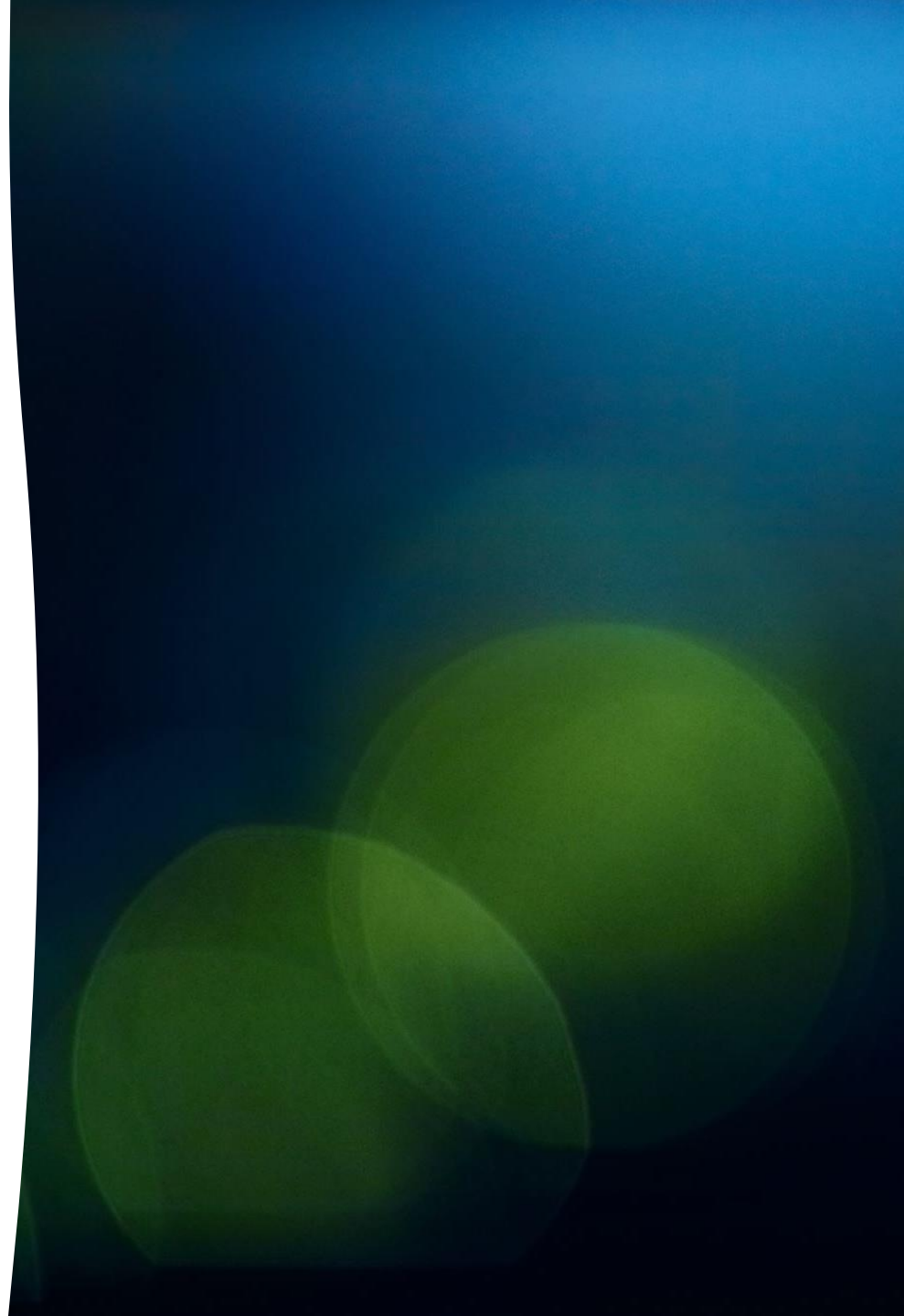


Challenges in Forecasting

- Data gaps
(weekends/holidays)

- Anomalies (weather, construction)

- High variability in demand



Wrap-Up



- TIME SERIES ANALYSIS
IS CRITICAL IN URBAN
PLANNING



- HELPS FORECAST
DEMAND AND SUPPORT
DECISIONS



- EVALUATE ACCURACY
CAREFULLY

Discussion Questions

- What time frequency is best: daily, weekly, monthly?

- How do shocks (e.g., COVID) affect results?

- How can forecasts be used by agencies?

Thank You!

