# **Time Series Note**

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python python python plotly d3.js plotly

## 1

```
import plotly.graph_objects as go
import numpy as np
```

## 1.1 Time Series v.s. Cross Sectional

(time series) (cross sectional)

Table 1.1: Time Series v.s. Cross Sectional

Time Series	Cross Sectional
2000/01/01 2020/01/01	2000/01/01 50

## 1.2

2000/01/01 2020/01/01

## 1.3

- $\quad \quad \boldsymbol{U}_t, t \in \{1, \cdots, T\}$   $\quad \quad \quad \boldsymbol{U}_t$
- $U \rightarrow$
- $t \rightarrow$

 $U_t$ 

```
U_t = T_t + S_t + C_t + R_t \label{eq:ut}
```

```
 \begin{array}{ll} \bullet & T_t: & \text{(Trend component)} \\ \bullet & S_t: & \text{(Seasonal component)} \\ \bullet & C_t: & \text{(Cyclical component)} \\ \bullet & R_t: & \text{(Random component)} \\ \end{array}
```

#### 1.3.1

• Downward Trend: Television Newspaper

• No Trend: Radio

• Upward Trand: Internet

```
title = 'Main Source for News'
labels = ['Television', 'Newspaper', 'Internet', 'Radio']
colors = ['rgb(67,67,67)', 'rgb(115,115,115)', 'rgb(49,130,189)', 'rgb(189,189,189)']
mode_size = [8, 8, 12, 8]
line size = [2, 2, 4, 2]
x_data = np.vstack((np.arange(2001, 2014),)*4)
y_data = np.array([
    [74, 82, 80, 74, 73, 72, 74, 70, 70, 66, 66, 69],
    [45, 42, 50, 46, 36, 36, 34, 35, 32, 31, 31, 28],
    [13, 14, 20, 24, 20, 24, 24, 40, 35, 41, 43, 50],
    [18, 21, 18, 21, 16, 14, 13, 18, 17, 16, 19, 23],
])
fig = go.Figure()
for i in range (0, 4):
    fig.add_trace(go.Scatter(x=x_data[i], y=y_data[i], mode='lines',
        name=labels[i],
```

```
line=dict(color=colors[i], width=line_size[i]),
        connectgaps=True,
    ))
    # endpoints
    fig.add_trace(go.Scatter(
        x=[x_data[i][0], x_data[i][-1]],
        y=[y_data[i][0], y_data[i][-1]],
        mode='markers',
        marker=dict(color=colors[i], size=mode_size[i])
    ))
fig.update_layout(
    xaxis=dict(
        showline=True,
        showgrid=False,
        showticklabels=True,
        linecolor='rgb(204, 204, 204)',
        linewidth=2,
        ticks='outside',
        tickfont=dict(
            family='Arial',
            size=12,
            color='rgb(82, 82, 82)',
        ),
    ),
    yaxis=dict(
        showgrid=False,
        zeroline=False,
        showline=False,
        showticklabels=False,
    ),
    autosize=False,
    margin=dict(
        autoexpand=False,
        1=100,
        r=20,
        t=110,
    ),
    showlegend=False,
    plot_bgcolor='white'
```

```
)
annotations = []
# Adding labels
for y_trace, label, color in zip(y_data, labels, colors):
    # labeling the left side of the plot
    annotations.append(dict(xref='paper', x=0.05, y=y_trace[0],
                                  xanchor='right', yanchor='middle',
                                  text=label + ' {}%'.format(y_trace[0]),
                                  font=dict(family='Arial',
                                             size=16),
                                  showarrow=False))
    # labeling the right_side of the plot
    annotations.append(dict(xref='paper', x=0.95, y=y_trace[11],
                                  xanchor='left', yanchor='middle',
                                  text='{}%'.format(y_trace[11]),
                                  font=dict(family='Arial',
                                             size=16),
                                  showarrow=False))
# Title
annotations.append(dict(xref='paper', yref='paper', x=0.0, y=1.05,
                              xanchor='left', yanchor='bottom',
                              text='Main Source for News',
                              font=dict(family='Arial',
                                        size=30,
                                        color='rgb(37,37,37)'),
                              showarrow=False))
# Source
annotations.append(dict(xref='paper', yref='paper', x=0.5, y=-0.1,
                              xanchor='center', yanchor='top',
                              text='Source: PewResearch Center & ' +
                                    'Storytelling with data',
                              font=dict(family='Arial',
                                         size=12,
                                         color='rgb(150,150,150)'),
                              showarrow=False))
fig.update_layout(annotations=annotations)
fig.show()
```

```
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```

#### 1.3.2

```
title = 'Seasonal Effect'

x_data = np.linspace(-10, 10, 100)

y_data = np.array(
    np.sin(x_data) + np.random.normal(size=x_data.shape) * 0.25
)

fig = go.Figure()

fig.add_trace(go.Scatter(x=x_data, y=y_data, mode='lines+markers', name="sin",
    line=dict(color="blue", width=1),
    connectgaps=True,
))
fig.update_layout(title=title)
fig.show()
```

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```
title = 'Cyclical Effect'

x_data = np.linspace(-10, 10, 100)

y_data = np.array(
    np.sin(x_data)*np.exp(x_data/10) + np.random.normal(size=x_data.shape) * 0.25
)

fig = go.Figure()
```

#### 1.3.3

```
sin

title = 'Seasonal Effect'

x_data = np.linspace(-10, 10, 100)

y_data = np.array(
    np.sin(x_data) + np.random.normal(size=x_data.shape) * 1
)

fig = go.Figure()

fig.add_trace(go.Scatter(x=x_data, y=y_data, mode='lines+markers', name="sin",
    line=dict(color="blue", width=1),
    connectgaps=True,
))
fig.update_layout(title=title)
fig.show()
```

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#### 1.4

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### 1.5 Stationary

#### 1.5.1 Weak stationary

```
\{Z_t\}
          (second-order or covariance stationary)
  • \mu(t) = C : \mu
  \bullet \quad \gamma(t,t-k) = \gamma(0,k):
                                      (lag)
                                        \sim \mathcal{N}(0, \sigma^2 I)
                                            X_t = X_{t-1} + \epsilon
   \begin{array}{ll} \bullet & \mathbb{E}(X_t) = \mathbb{E}(X_{t-1}) \\ \bullet & \mathrm{Var}(X_t) = t\sigma^2 \end{array} 
 title = 'Seasonal Effect'
 T = 1000
 walks = []
 loc = 0
 for i in range(T):
       loc += + np.random.normal(0,1)
       walks.append(loc)
 walks = np.array(walks)
 fig = go.Figure()
 fig.add_trace(go.Scatter(x=list(range(T)), y=walks, mode='lines',
       name="sin",
       line=dict(color="blue", width=1),
       connectgaps=True,
  fig.update_layout(title=title)
 fig.show()
```

$$D_t = X_t - X_{t-1} =$$

- $\mathbb{E}(D_t) = 0$
- $Var(X_t) = \sigma^2$

```
title = 'Seasonal Effect'

D = walks[1:] - walks[:-1]

fig = go.Figure()

fig.add_trace(go.Scatter(x=list(range(T-1)), y=D, mode='lines', name="sin", line=dict(color="blue", width=1), connectgaps=True,
))
fig.update_layout(title=title)
fig.show()
```

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#### 1.5.2 Strict stationary

$$\begin{aligned} \{Z_t\} \\ Z_{t_1}, Z_{t_2}, \cdots, Z_{t_n} & Z_{t_1-k}, Z_{t_2-k}, \cdots, Z_{t_n-k} \end{aligned}$$

Note:

•

•

### 1.5.3 Sample AutoCorrelation Function (ACF)

? \*\*\* \*\* ACF

$$\rho_{\ell} = \frac{\mathrm{cov}(r_t, r_{t-\ell})}{\mathrm{var}(r_t)}$$

- $\begin{array}{l} \bullet \quad \gamma_k = \mathrm{Cov}(r_t, r_{t-k}) = E[(r_t \mu)(r_{t-k} \mu)]: \\ \bullet \quad \rho_0 = 1: \end{array}$
- $\rho_k = \rho_{-k}$ :

Lag- $\ell$   $\mu$  ( )

$$\hat{\rho}_{\ell} = \frac{\Sigma_{t=1}^{T-\ell} (r_t - \bar{r}) (r_{t+\ell} - \bar{r})}{\Sigma_{t=1}^T (r_t - \bar{r})^2}$$

- $\bullet$   $\bar{r}$ :
- T:

python numpy.correlate numpy.correlate(a, v, mode) v filter v a ( ) numpy.correlate

$$c_k = \sum_n a_{n+k} * \bar{v}_n$$

- $\bar{v}_n$  complex conjugation
- mode:

- valid: v a a - same: v - full: v

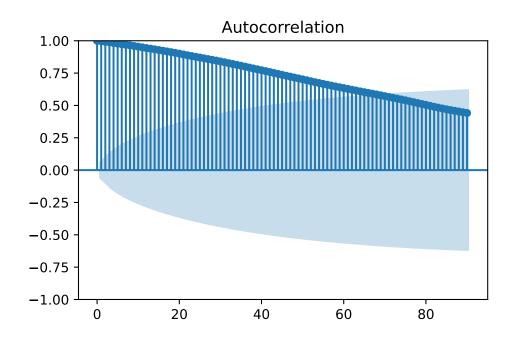
def acf(series, lag=None):

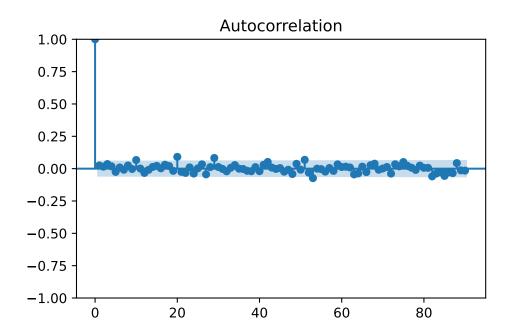
Calculate the autocorrelation function (ACF) for a given time series.

Parameters:

```
series (array-like): The time series data.
        lag (int, optional): The maximum lag for which to calculate the ACF.
            If None (default), the ACF is calculated for all possible lags.
    Returns:
        acf_values (array): Autocorrelation values for the given time series.
            If `lag` is provided, returns a single value representing
            autocorrelation at that lag. Otherwise, returns an array of
            autocorrelation values for all lags.
    11 11 11
    series = np.asarray(series)
    n = len(series)
    mean = np.mean(series)
    centered_data = series - mean
    # Calculate autocovariance function
    acov = np.correlate(centered_data, centered_data, mode='full')
    acf_values = acov / sum(centered_data ** 2)
    if lag is not None:
        return acf_values[(n-1):(n-1+lag)]
    else:
        return acf_values[(n-1):]
       ACF
rw_acf = acf(walks, lag=90)
title = 'ACF Random Walks'
fig = go.Figure()
fig.add_trace(go.Scatter(x=list(range(T)), y=rw_acf, mode='markers',
    name="sin",
    line=dict(color="blue", width=1),
    connectgaps=True,
))
fig.update_layout(title=title)
fig.show()
```

```
d_acf = acf(D, lag=90)
  title = 'ACF Random Walks'
  fig = go.Figure()
  fig.add_trace(go.Scatter(x=list(range(T-1)), y=d_acf, mode='markers',
      name="sin",
      line=dict(color="blue", width=1),
      connectgaps=True,
  ))
  fig.update_layout(title=title)
  fig.show()
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      python statsmodels.graphics.tsaplots.plot_acf
  from statsmodels.graphics.tsaplots import plot_acf
  import matplotlib.pyplot as plt
  import numpy as np
  import pandas as pd
  plot_acf(walks, lags=90)
  plt.show()
  plot_acf(D, lags=90)
  plt.show()
```





## References