

Matplotlib Cheatsheet

Essential operations for data visualization and plotting

This cheatsheet provides a quick reference to fundamental Matplotlib operations, syntax, and advanced features, ideal for both beginners and experienced data scientists for creating effective data visualizations.

Basic Plotting Create fundamental plots and charts	Plot Customization Style and format your visualizations	Multiple Plots Create subplots and complex layouts
Advanced Charts Specialized visualization types	Export & Save Output plots in various formats	

Basic Plotting & Chart Types

Line Plot: `plt.plot()`

Create line charts for continuous data visualization.

```
import matplotlib.pyplot as plt
import numpy as np

# Basic line plot
x = [1, 2, 3, 4, 5]
y = [2, 4, 6, 8, 10]
plt.plot(x, y)
plt.show()

# Multiple lines
plt.plot(x, y, label='Line 1')
plt.plot(x, [1, 3, 5, 7, 9], label='Line 2')
plt.legend()

# Line styles and colors
plt.plot(x, y, 'r--', linewidth=2, marker='o')
```

Scatter Plot: `plt.scatter()`

Display relationships between two variables.

```
# Basic scatter plot
plt.scatter(x, y)

# With different colors and sizes
colors = [1, 2, 3, 4, 5]
sizes = [20, 50, 100, 200, 500]
plt.scatter(x, y, c=colors, s=sizes, alpha=0.6)
plt.colorbar() # Add color bar
```

Bar Chart: `plt.bar()` / `plt.barh()`

Create vertical or horizontal bar charts.

```
# Vertical bars
categories = ['A', 'B', 'C', 'D']
values = [20, 35, 30, 25]
plt.bar(categories, values)

# Horizontal bars
plt.barh(categories, values)

# Grouped bars
x = np.arange(len(categories))
plt.bar(x - 0.2, values, 0.4, label='Group 1')
plt.bar(x + 0.2, [15, 25, 35, 20], 0.4, label='Group 2')
```

Plot Customization & Styling

Enhance your plots with labels, titles, colors, and formatting.

01	02	03
Labels & Titles: `plt.xlabel()` / `plt.title()` Add descriptive text to your plots for clarity and context.	Colors & Styles: `color` / `linestyle` / `marker` Customize the visual appearance of plot elements.	Legends & Annotations: `plt.legend()` / `plt.annotate()` Add legends and annotations to explain plot elements.

Axes & Layout Control

Axis Limits: `plt.xlim()` / `plt.ylim()`

Control the range of values displayed on each axis.

```
# Set axis limits
plt.xlim(0, 10)
plt.ylim(-5, 15)

# Auto-adjust limits with margin
plt.margins(x=0.1, y=0.1)

# Invert axis
plt.gca().invert_yaxis() # Invert y-axis
```

Ticks & Labels: `plt.xticks()` / `plt.yticks()`

Customize axis tick marks and their labels.

```
# Custom tick positions
plt.xticks([0, 2, 4, 6, 8, 10])
plt.yticks(np.arange(0, 101, 10))

# Custom tick labels
plt.xticks([0, 1, 2, 3], ['Jan', 'Feb', 'Mar', 'Apr'])

# Rotate tick labels
plt.xticks(rotation=45)

# Remove ticks
plt.xticks([])
plt.yticks([])
```

Aspect Ratio: `plt.axis()`

Control the aspect ratio and axis appearance.

```
# Equal aspect ratio
plt.axis('equal')

# Square plot
plt.axis('square')

# Turn off axis
plt.axis('off')

# Custom aspect ratio
plt.gca().set_aspect('equal', adjustable='box')
```

Subplots & Multiple Plots

Basic Subplots: `plt.subplot()` / `plt.subplots()`

Create multiple plots in a single figure.

```
# Create 2x2 subplot grid
fig, axes = plt.subplots(2, 2, figsize=(10, 8))

# Plot in each subplot
axes[0, 0].plot(x, y)
axes[0, 1].scatter(x, y)
axes[1, 0].bar(x, y)
axes[1, 1].hist(y, bins=10)

# Alternative syntax
plt.subplot(2, 2, 1) # 2 rows, 2 cols, 1st subplot
plt.plot(x, y)
plt.subplot(2, 2, 2) # 2nd subplot
plt.scatter(x, y)
```

Shared Axes: `sharex` / `sharey`

Link axes across subplots for consistent scaling.

```
# Share x-axis across subplots
fig, axes = plt.subplots(2, 1, sharex=True)
axes[0].plot(x, y1)
axes[1].plot(x, y2)

# Share both axes
fig, axes = plt.subplots(2, 2, sharex=True, sharey=True)
```

Advanced Visualization Types

Heatmaps: `plt.imshow()` / `plt.pcolormesh()`

Visualize 2D data as color-coded matrices.

```
# Basic heatmap
data = np.random.randn(10, 10)
plt.imshow(data, cmap='viridis')
plt.colorbar()

# Pcolormesh for irregular grids
x = np.linspace(0, 10, 11)
y = np.linspace(0, 5, 6)
X, Y = np.meshgrid(x, y)
Z = np.sin(X) * np.cos(Y)
plt.pcolormesh(X, Y, Z, shading='auto')
plt.colorbar()
```

Contour Plots: `plt.contour()` / `plt.contourf()`

Show level curves and filled contour regions.

```
# Contour lines
x = np.linspace(-3, 3, 100)
y = np.linspace(-3, 3, 100)
X, Y = np.meshgrid(x, y)
Z = X**2 + Y**2
plt.contour(X, Y, Z, levels=10)
plt.clabel(plt.contour(X, Y, Z), inline=True, fontsize=8)

# Filled contours
plt.contourf(X, Y, Z, levels=20, cmap='RdBu')
plt.colorbar()
```

3D Plots: `mplot3d`

Create three-dimensional visualizations.

```
from mpl_toolkits.mplot3d import Axes3D
fig = plt.figure()
ax = fig.add_subplot(111, projection='3d')

# 3D scatter
ax.scatter(x, y, z)

# 3D surface plot
ax.plot_surface(X, Y, Z, cmap='viridis')

# 3D line plot
ax.plot(x, y, z)
```

Interactive & Animation Features

Create dynamic and interactive visualizations.

Interactive Backend: `%matplotlib widget`

Enable interactive plots in Jupyter notebooks.

```
# In Jupyter notebook
%matplotlib widget

# Or for basic interactivity
%matplotlib notebook

# Interactive zoom, pan, and hover
```

Event Handling: Mouse & Keyboard

Respond to user interactions with plots.

```
def onclick(event):
    if event.inaxes:
        print(f'Clicked at x={event.xdata}, y={event.ydata}')

fig, ax = plt.subplots()
ax.plot(x, y)
fig.canvas.mpl_connect('button_press_event', onclick)
plt.show()
```

Saving & Exporting Plots

Save your visualizations in various formats for different purposes.

Save Figure: `plt.savefig()` Export plots to image files with various options. <pre># Basic save plt.savefig('my_plot.png') # High-quality save plt.savefig('plot.png', dpi=300, bbox_inches='tight') # Different formats plt.savefig('plot.pdf') # PDF plt.savefig('plot.svg') # SVG (vector) plt.savefig('plot.eps') # EPS # Transparent background plt.savefig('plot.png', transparent=True)</pre>	Figure Quality: DPI & Size Control resolution and dimensions of saved plots. <pre># High DPI for publications plt.savefig('plot.png', dpi=600) # Custom size (width, height in inches) plt.figure(figsize=(12, 8)) plt.savefig('plot.png', figsize=(12, 8)) # Crop whitespace plt.savefig('plot.png', bbox_inches='tight', pad_inches=0.1)</pre>	Batch Export & Memory Management Handle multiple plots and memory efficiently. <pre># Close figures to free memory plt.close() # Close current figure plt.close('all') # Close all figures # Context manager for automatic cleanup with plt.figure() as fig: plt.plot(x, y) plt.savefig('plot.png') # Batch save multiple plots for i, data in enumerate(datasets): plt.figure() plt.plot(data) plt.savefig(f'plot_{i}.png') plt.close()</pre>
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Configuration & Best Practices

Optimize Matplotlib settings and follow visualization best practices.

RC Parameters: `plt.rcParams`

Set default styling and behavior for all plots.

```
# Common rc parameters
plt.rcParams['figure.figsize'] = (10, 6)
plt.rcParams['font.size'] = 12
plt.rcParams['lines.linewidth'] = 2
plt.rcParams['axes.grid'] = True

# Save and restore settings
original_params = plt.rcParams.copy()
# ... make changes ...
plt.rcParams.update(original_params) # Restore
```

Color Management: Colormaps & Palettes

Work effectively with colors and colormaps.

```
# List available colormaps
print(plt.colormaps())

# Use colormap for multiple lines
colors = plt.cm.viridis(np.linspace(0, 1, len(datasets)))
for i, (data, color) in enumerate(zip(datasets, colors)):
    plt.plot(data, color=color, label=f'Dataset {i+1}')

# Custom colormap
from matplotlib.colors import
LinearSegmentedColormap
custom_cmap =
LinearSegmentedColormap.from_list('custom', ['red',
'yellow', 'blue'])
```

Integration with Data Libraries

Pandas Integration: Direct Plotting

Use Matplotlib through Pandas DataFrame methods.

```
import pandas as pd

# Dataframe plotting (uses matplotlib backend)
df.plot(kind='line', x='date', y='value')
df.plot.scatter(x='x_col', y='y_col')
df.plot.hist(bins=30)

df.plot.box()

# Access underlying matplotlib objects
ax = df.plot(kind='line')
ax.set_title('Custom Title')
plt.show()
```

NumPy Integration: Array Visualization

Efficiently plot NumPy arrays and mathematical functions.

```
# 2D array visualization
arr = np.random.randn(10, 10)
plt.imshow(arr, cmap='hot', interpolation='nearest')

# Mathematical functions
x = np.linspace(0, 4*np.pi, 1000)
y = np.sin(x) * np.exp(-x/10)
plt.plot(x, y)

# Statistical distributions
data = np.random.normal(0, 1, 10000)
plt.hist(data, bins=50, density=True, alpha=0.7)
```

Installation & Environment Setup

Install and configure Matplotlib for various Python environments.

Pip: `pip install matplotlib` Standard Python package installer for Matplotlib. <pre># Install Matplotlib pip install matplotlib # Upgrade to latest version pip install matplotlib --upgrade # Install with additional backends pip install matplotlib[qts] # Show package information pip show matplotlib</pre>	Conda: `conda install matplotlib` Package manager for Anaconda/Miniconda environments. <pre># Install in current environment conda install matplotlib # Update matplotlib conda update matplotlib # Create environment with matplotlib conda create -n dataviz matplotlib numpy pandas # List matplotlib info conda list matplotlib</pre>	Backend Configuration Set up display backends for different environments. <pre># Check available backends import matplotlib print(matplotlib.get_backend()) # Set backend programmatically matplotlib.use('TkAgg') # For Tkinter matplotlib.use('Qt5Agg') # For PyQt5 # For headless servers matplotlib.use('Agg') # Import after setting backend import matplotlib.pyplot as plt</pre>
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Reference: This cheatsheet covers essential Matplotlib commands and modern practices for creating effective data visualizations in the data science workflows.