

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

Advanced Charts

Specialized visualization types

Plot Customization

Style and format your visualizations

Multiple Plots

Create subplots and complex layouts

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([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.barch()'

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.barch(categories, values)
# Grouped bars
x = np.arange(len(categories))
plt.barch(x - 0.2, values, 0.4, label='Group 1')
plt.barch(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.

```
# Basic labels and title
plt.plot(x, y)
plt.xlabel('X-axis Label')
plt.ylabel('Y-axis Label')
plt.title('Plot Title')
# Formatted titles with font properties
plt.title('My Plot', fontsize=16, fontweight='bold')
plt.xlabel('X Values', fontsize=12)
# Grid for better readability
plt.grid(True, alpha=0.3)
```

Colors & Styles: `color` / `linestyle` / `marker`

Customize the visual appearance of plot elements.

```
# Color options
plt.plot(x, y, color='red') # Named colors
plt.plot(x, y, color="#FF5733") # Hex colors
plt.plot(x, y, color=(0.1, 0.2, 0.5)) # RGB tuple
# Line styles
plt.plot(x, y, linestyle='--') # Dashed
plt.plot(x, y, linestyle=': ') # Dotted
plt.plot(x, y, linestyle='-.') # Dash-dot
# Markers
plt.plot(x, y, marker='o', markersize=8, markerfacecolor='red')
```

Legends & Annotations: `plt.legend()` / `plt.annotate()`

Add legends and annotations to explain plot elements.

```
# Basic legend
plt.plot(x1, y1, label='Dataset 1')
plt.plot(x2, y2, label='Dataset 2')
plt.legend()
# Customize legend position
plt.legend(loc='upper right', fontsize=10, frameon=False)
# Annotations
plt.annotate('Important Point', xy=(2, 4), xytext=(3, 6),
            arrowprops=dict(arrowstyle='>', color='red'))
```

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.subplots()` / `plt.subplot()`

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(y, x)
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)
```

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: `mpl3d`

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

Interactive & Animation Features

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.

```
# 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
# Transparent background
plt.savefig('plot.png',
transparent=True)
```

Figure Quality: DPI & Size

Control resolution and dimensions of saved plots.

```
# High DPI for publications
plt.savefig('plot.png', dpi=600)
# Custom size (width, height in inches)
plt.savefig('plot.png', figsize=(12, 8))
# Crop whitespace
plt.savefig('plot.png',
bbox_inches='tight',
pad_inches=0.1)
```

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')
```

Advanced Visualization Types

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

Visualize 2D data as color-coded matrices.

```
# Basic heatmap
data = np.random.rand(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(Y) * np.cos(X)
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: `mpl3d`

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

Installation & Environment Setup

Install and configure Matplotlib for various Python environments.

Pip: `pip install matplotlib`

Standard Python package installer for Matplotlib.

```
# Install Matplotlib
pip install matplotlib
# Upgrade to latest version
pip install matplotlib --upgrade
# Install with additional backends
pip install matplotlib[qt5]
# Show package information
pip show matplotlib
```

Conda: `conda install matplotlib`

Package manager for Anaconda/Miniconda environments.

```
# 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
```

Backend Configuration

Set up display backends for different environments.

```
# Check available backends
import matplotlib
print(matplotlib.get_backend())
# Use backend programmatically
matplotlib.use('TkAgg')
```

Performance Optimization

Improve plotting performance for large datasets.

```
# Use blitting for animations
ani = FuncAnimation(fig, animate, frames=200, blit=True)
# Rasterize complex plots
plt.rasterize(True)
# Reduce data points for large datasets
indices = np.arange(0, len(large_data), step=10)
plt.plot(large_data[indices])
```

Memory Usage: Efficient Plotting

Manage memory when creating many plots or large visualizations.

```
# Clear axes instead of creating new figures
fig, ax = plt.subplots()
for data in datasets:
    ax.clear() # Clear previous plot
    ax.plot(data)
# Use generators for large datasets
def data_generator():
    for i in range(1000):
        yield np.random.randn(100)
```

Seaborn Integration: Enhanced Styling

Combine Matplotlib with Seaborn for better default aesthetics.

```
import seaborn as sns
# Use seaborn styling with matplotlib
sns.set_style('whitegrid')
plt.plot(x, y)
plt.show()
```

Jupyter Integration: Inline Plotting

Optimize Matplotlib for Jupyter notebook environments.

```
# Magic commands for Jupyter
%matplotlib inline
# Static plots
%matplotlib widget # Interactive plots
# High-DPI displays
%config InlineBackend.figure_format = 'retina'
# Automatic figure sizing
%matplotlib inline
# Set backend for Jupyter
plt.rcParams['figure.dpi'] = 100
```

Reference

This cheatsheet covers essential Matplotlib commands and modern practices for creating effective data visualizations in Python data science workflows.