# Python Data Analysis Reference Guide

# by Manu Gupta

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# 1 Python Fundamentals for Data Analysis

## 1.1 Jupyter Notebook Essentials

## Keyboard Shortcuts

- Shift + Enter: Execute cell
- Esc + m: Change to Markdown cell
- Esc + y: Change to Code cell
- Ctrl + Enter: Run cell in place

### 1.2 Markdown Basics

```
# Heading 1
## Heading 2
**Bold Text**
*Italic Text*
[Link Text](URL)
```

## 1.3 Data Types

- Integer: -256, 15
- **Float:** -253.23, 1.253e-10
- String: "Hello", 'Goodbye'

- Boolean: True, False
- **List:** [value, ...]
- **Dictionary:** {key: value}

## 1.4 Basic Python Operations

## 1.4.1 Arithmetic Operators

Arithmetic Operations								
• x + y	(add)	• x / y	(divide)					
• x - y	(subtract)	• x % y	(modulus)					
• x * y	(multiply)	• x ** y	(power)					
Assignment shortcuts: $x \text{ op} = y$ Example: $x += 1$ increments $x$								

## 1.4.2 Comparison Operators

Comparison Operations			
• x < y	(less than)	• x >= y	(greater or equal)
• x <= y	(less or equal)	• x == y	(equal)
• x > y	(greater than)	• x ! = y	(not equal)

#### 1.4.3 Boolean Operators

```
Boolean Operations

• not x

(logical NOT)

• x and y

(logical AND)

• x or y

(logical OR)
```

#### 1.4.4 If-Else Statements

```
# Basic if statement
if expression:
    statements
elif expression:
    statements
else:
    statements

# Example
score = 85
if score >= 90:
    grade = 'A'
elif score >= 80:
    grade = 'B'
else:
    grade = 'C'
```

#### 1.4.5 List Operations

```
Common List Operations
# List creation and modification
lst = [1, 2, 3, 4, 5]
                 # Deletes ith item from 1st
del lst[i]
lst.append(e) # Appends e to lst
lst.insert(i, e) # Inserts e before ith item in lst
                 # Sorts 1st
lst.sort()
lst.reverse()
                # Reverses 1st
                # Removes and returns last element
lst.pop()
lst.index(x)
                # Returns index of first x
lst.count(x)
                # Counts occurrences of x
```

## 1.5 Control Structures

## 1.5.1 List Comprehension

```
squares = [x**2 for x in range(10)]
evens = [x for x in range(10) if x % 2 == 0]
```

#### 1.5.2 Loops

```
# For Loop
for i in range(start, end, step):
    print(i)
# While Loop
```

```
count = 0
while count < 5:
    print(count)
    count += 1</pre>
```

#### 1.6 Functions and Classes

```
# Function Definition
def calculate_mean(numbers):
    return sum(numbers) / len(numbers)

# Class Definition
class DataPoint:
    def __init__(self, value):
        self.value = value

    def transform(self):
        return self.value ** 2
```

## 2 Data Analysis Tools

## 2.1 NumPy Essentials

```
import numpy as np

# Array Creation
arr = np.array([1, 2, 3])
zeros = np.zeros((2, 3))
ones = np.ones((3, 2))
identity = np.eye(3)

# Array Operations
mean = np.mean(arr)
std = np.std(arr)
sum = np.sum(arr)
```

## 2.2 Pandas Fundamentals

```
import pandas as pd
# Reading Data
df = pd.read_csv("file.csv")
# Basic Operations
df.head() # First 5 rows
df.describe() # Statistical summary
df.info() # DataFrame information
# Data Cleaning
df.dropna(inplace=True) # Remove rows with missing values
df.fillna(0, inplace=True) # Fill missing values
#inplace=True, in order to modifying the already existing dataframe.
# Column Operations
df['new_column'] = df['col1'] + df['col2']
df.rename(columns={'old': 'new'}, inplace=True)
df.drop(['Column name'], axis=1, inplace=True) # Remove a column
pd.to_numeric(df['Temperature']) # Convert the column data type (e.g., string) to
   numeric
# Date Operations
df['date'] = pd.to_datetime(df['date']) # Convert the column to date time format
```

## 2.3 Data Visualization with Matplotlib

```
import matplotlib.pyplot as plt

# Line Plot
plt.plot(x, y, marker='o')
plt.xlabel('Time')
plt.ylabel('Value')
plt.title('Time Series Data')

# Bar Plot
plt.bar(categories, values)
plt.show()

# Multiple Plots
fig, (ax1, ax2) = plt.subplots(1, 2)
ax1.plot(x1, y1)
ax2.scatter(x2, y2)
```

### 2.4 Random Data Generation

```
import random

# Random Numbers
random.random() # [0, 1)
random.uniform(0, 10) # [0, 10]
random.gauss(10, 2) # Gaussian distribution
random.randint(1, 100) # Integer [1, 100]

# Random Selections
random.choice(['apple', 'banana', 'cherry'])
random.sample(range(100), 10)

# Set Seed
random.seed(42)
```

## 3 Data Analysis Best Practices

## Python Data Analysis Tips

- Use vectorized operations (NumPy arrays) instead of loops when possible
- Create functions for repetitive analysis tasks
- Keep original data unchanged, work with copies
- Use appropriate data types for statistical calculations
- Handle missing values appropriately for your analysis
- Document your analysis pipeline
- Use version control for your analysis scripts

## Statistical Analysis Tips

- Always check data distribution before applying statistical tests
- Use appropriate visualization for different types of data
- Check for outliers and their impact on analysis
- Validate assumptions of statistical models
- Document data cleaning and transformation steps
- Use appropriate measures of central tendency
- Consider the sample size when drawing conclusions

## 4 Useful Resources

- Python Documentation: https://docs.python.org
- Pandas Documentation: https://pandas.pydata.org/docs
- NumPy Documentation: https://numpy.org/doc
- Matplotlib Documentation: https://matplotlib.org