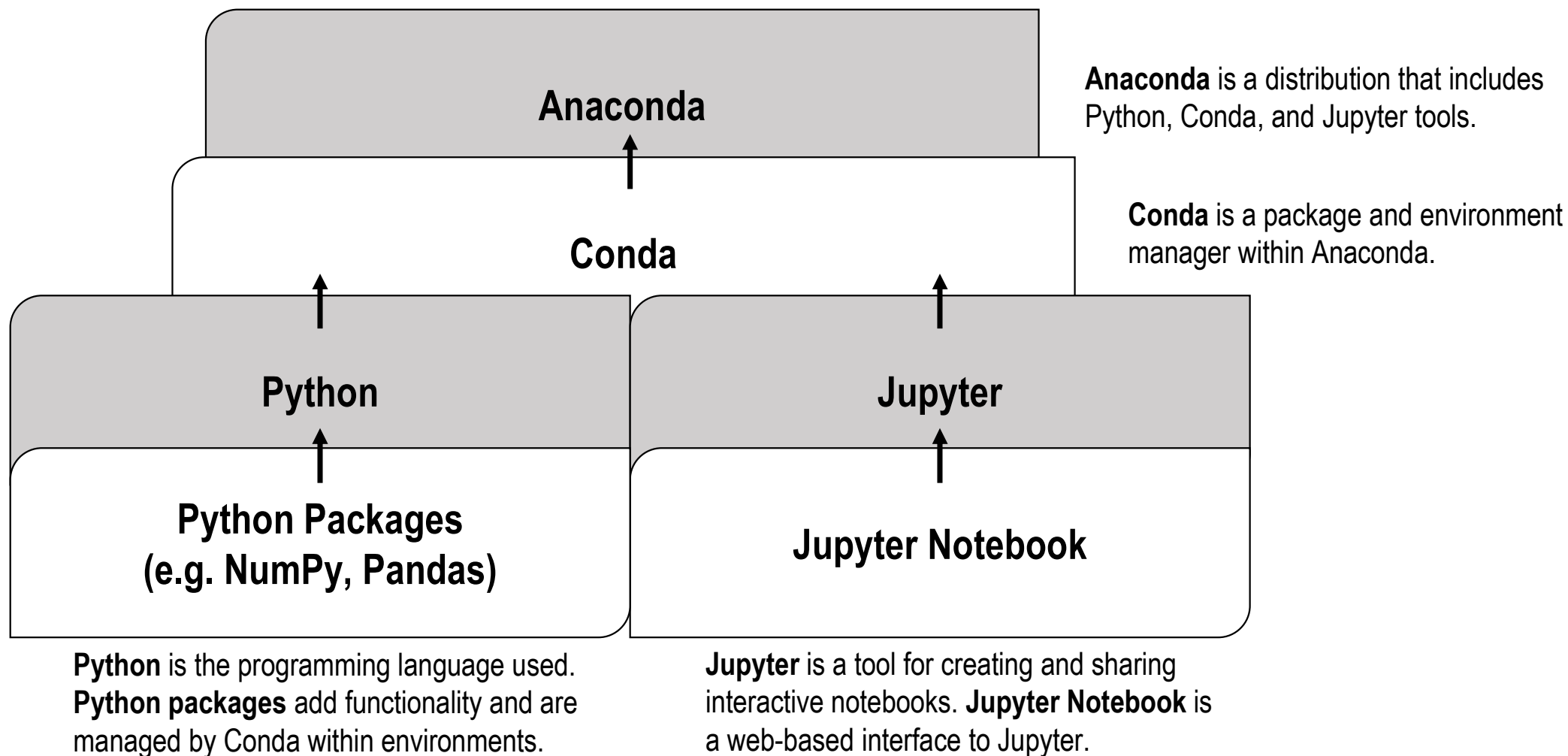
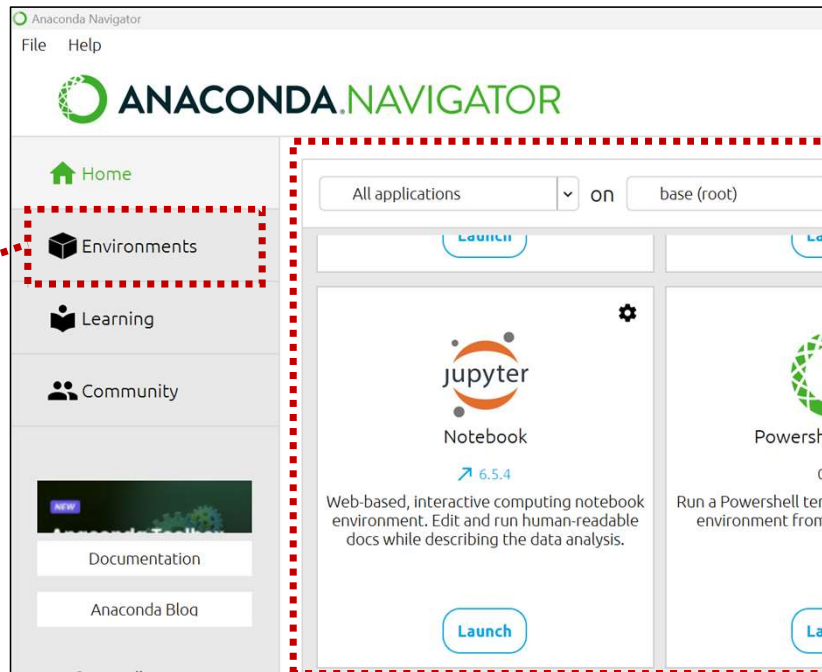


How Python, Conda, and Jupyter fit together?

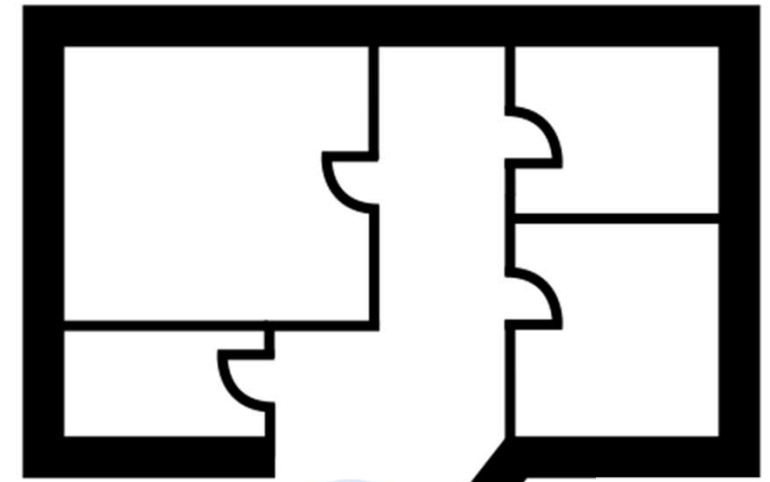


Explore Anaconda



**Installed
environments**

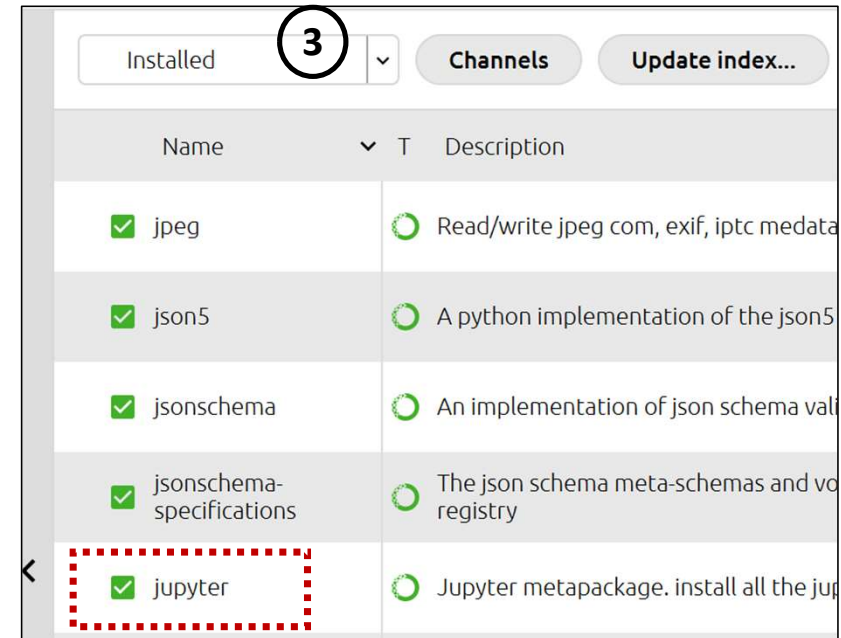
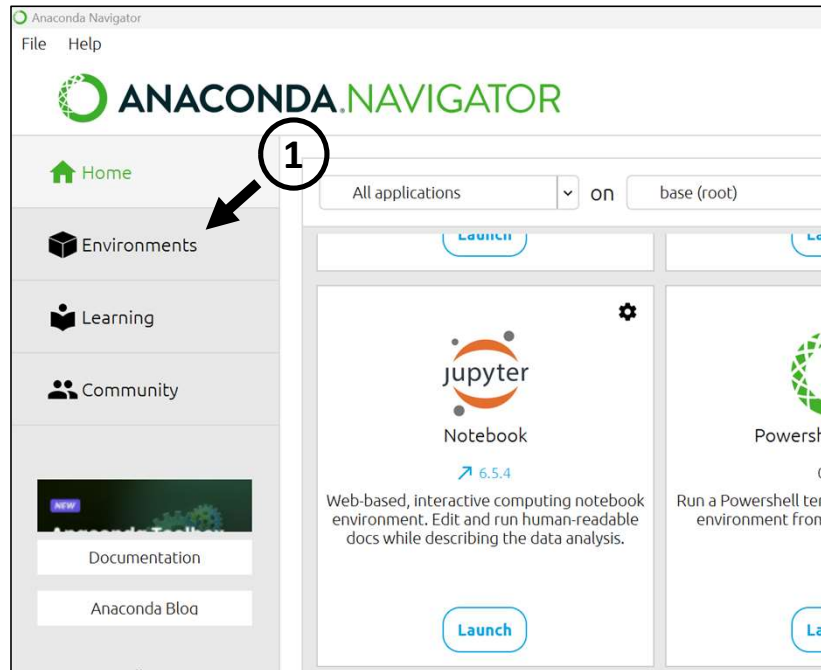
**Applications available
through Anaconda**



Created by Patrick Multani
from Noun Project

**Environments are like
separate rooms in a house**

Python packages



Jupyter is installed!

or

...

Python packages



```
Anaconda Prompt - conda in  X  +  v

1 (base) C:\Users\annika>conda activate Data_Visualization
2 (Data_Visualization) C:\Users\annika>conda install jupyter
Channels:
- defaults
Platform: win-64
Collecting package metadata (repodata.json): done
Solving environment: done

# All requested packages already installed.

(Data_Visualization) C:\Users\annika>conda list
# packages in environment at C:\Users\annika\anaconda3\envs\Data_Visualization:
#
# Name                                Version                                Build      Channel
anyio                                 4.2.0                                py311haa95532_0
argon2-cffi                           21.3.0                               pyhd3eb1b0_0
argon2-cffi-bindings                  21.2.0                               py311h2bbff1b_0
asttokens                             2.0.5                               pyhd3eb1b0_0
```

Launch Jupyter Notebook

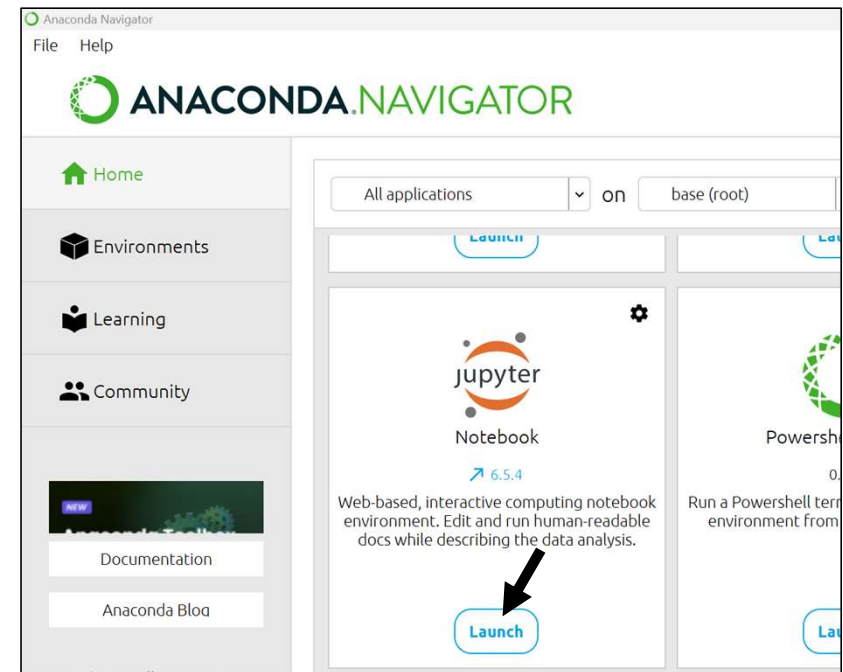
What is Jupyter Notebook?

- A web-based interactive computing platform
- Write and execute code, visualize data, and add narrative text all in one place
- One out of multiple options to execute Python code

```
Anaconda Prompt - jupyter n  X  +  v

(base) C:\Users\annika>conda activate Data_Visualization
(Data_Visualization) C:\Users\annika>jupyter notebook
```

or



Jupyter home screen



File explorer

Create new files & folders

Upload files and existing notebooks

File View Settings Help

Files Running

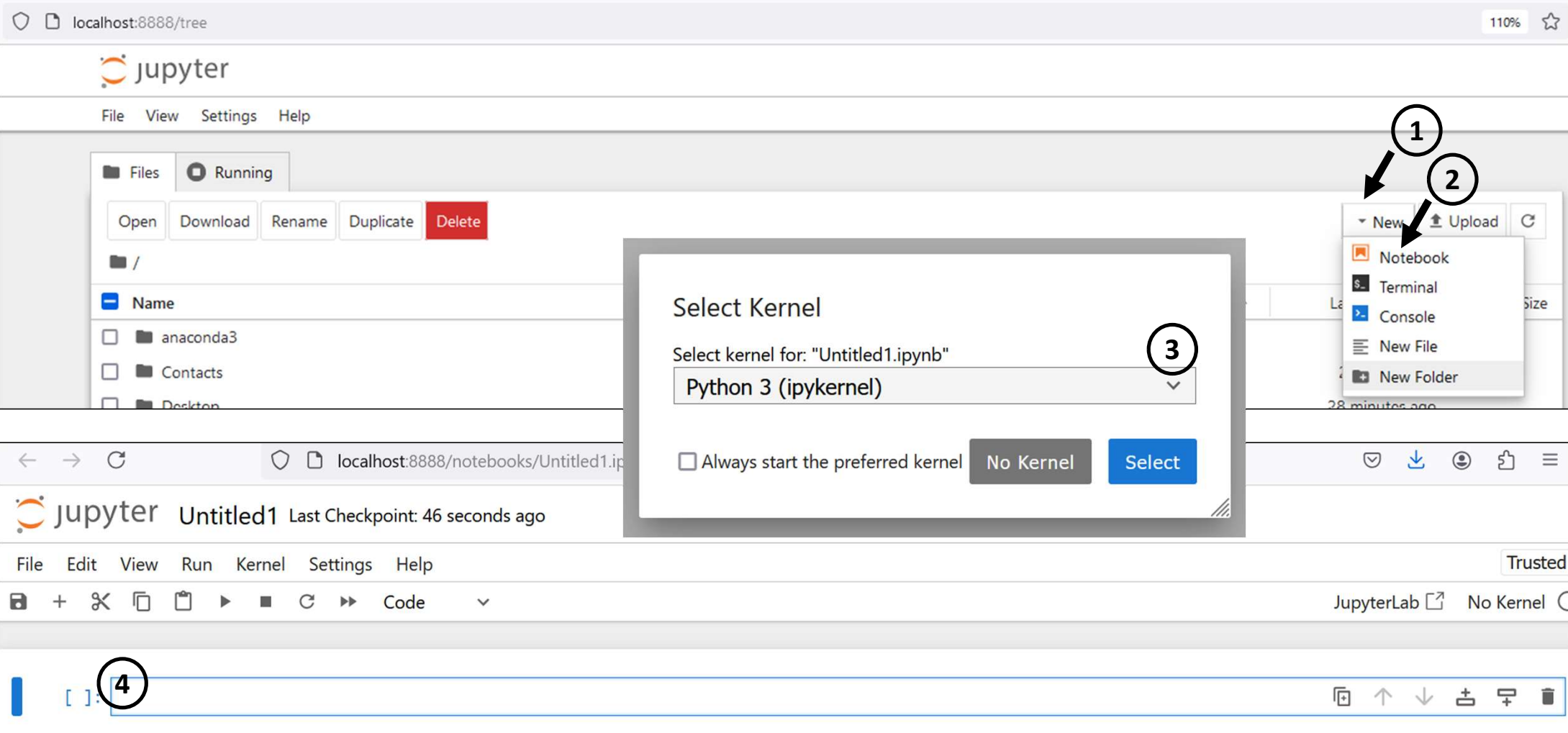
Select items to perform actions on them.

New Upload

/

<input type="checkbox"/> Name	Last Modified	File Size
<input type="checkbox"/> anaconda3	last month	
<input type="checkbox"/> Contacts	2 months ago	
<input type="checkbox"/> Desktop	24 minutes ago	
<input type="checkbox"/> Documents	last month	
<input type="checkbox"/> Downloads	20 hours ago	

Open a new Jupyter Notebook



The screenshot illustrates the steps to open a new Jupyter Notebook in JupyterLab. The interface is shown in two states: the initial file browser and the notebook editor.

Step 1: Click the 'New' button in the top right of the file browser.

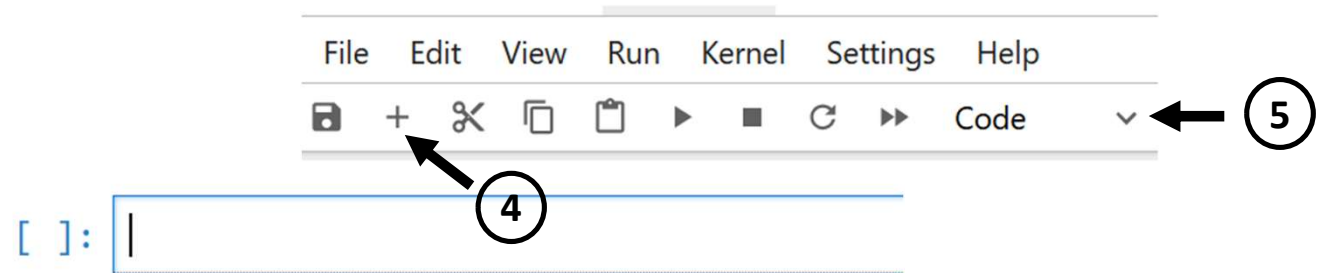
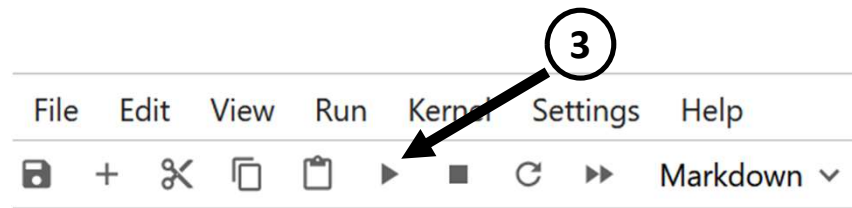
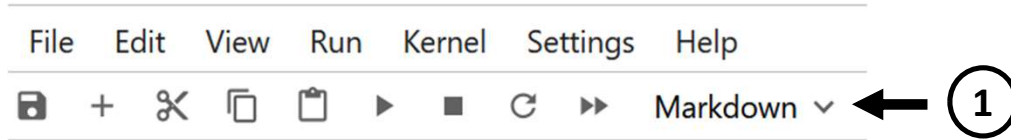
Step 2: Click 'Notebook' in the dropdown menu.

Step 3: In the 'Select Kernel' dialog, select 'Python 3 (ipykernel)' from the dropdown menu.

Step 4: Click the 'Select' button in the 'Select Kernel' dialog.

The final state shows the JupyterLab interface with the notebook 'Untitled1' open. The kernel is 'Python 3 (ipykernel)' and the status is 'No Kernel'.

Jupyter cell executions



What Python code looks like

1. Variables and data types

```
my_integer = 10
my_float = 3.14
my_string = "Hello, Python!"
print(my_integer)
print(my_float)
print(my_string)
```

4. Loops

```
print("Counting from 1 to 3:")
for i in range(1, 4):
    print(i)
```

5. Functions

```
def greet(name):
    return f"Hello, {name}!"

print(greet("Ana"))
print(greet("Alex"))
```

2. Basic operations

```
sum_result = my_integer + 5
product_result = my_float * 2
string_result = my_string + " Hello, world!"
print("Sum:", sum_result, "Product:", product_result, "Concatenation:", string_result)
```

```
# This is a comment!
1 + 1

2
```

3. Conditionals

```
if my_integer > 5:
    print("my_integer is greater than 5")
else:
    print("my_integer is less than or equal to 5")
```

6. Lists and list comprehensions

```
squares = [x * x for x in range(1, 6)]
print("Squares of numbers from 1 to 5:", squares)
```

What Python code looks like

1. Variables and data types

```
my_integer = 10
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my_string = "Hello, Python!"
print(my_integer)
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```

```
10
3.14
Hello, Python!
```

4. Loops

```
print("Counting from 1 to 3:")
for i in range(1, 4):
    print(i)
```

```
Counting from 1 to 3:
1
2
3
```

5. Functions

```
def greet(name):
    return f"Hello, {name}!"
```

```
print(greet("Ana"))
print(greet("Alex"))
```

```
Hello, Ana!
Hello, Alex!
```

2. Basic operations

```
Sum: 15 Product: 6.28 Concatenation: Hello, Python! Hello, world!
```

```
sum_result = my_integer + 5
product_result = my_float * 2
string_result = my_string + " Hello, world!"
print("Sum:", sum_result, "Product:", product_result, "Concatenation:", string_result)
```

```
# This is a comment!
1 + 1
```

```
2
```

3. Conditionals

```
if my_integer > 5:
    print("my_integer is greater than 5")
else:
    print("my_integer is less than or equal to 5")
```

```
my_integer is greater than 5
```

6. Lists and list comprehensions

```
squares = [x * x for x in range(1, 6)]
print("Squares of numbers from 1 to 5:", squares)
```

```
Squares of numbers from 1 to 5: [1, 4, 9, 16, 25]
```

Pandas and NumPy



Pandas and NumPy

Use-case: Iris dataset



Iris Versicolor



Iris Setosa



Iris Virginica

Figure from: <https://www.datacamp.com/tutorial/machine-learning-in-r>

Introduction to Pandas and DataFrames

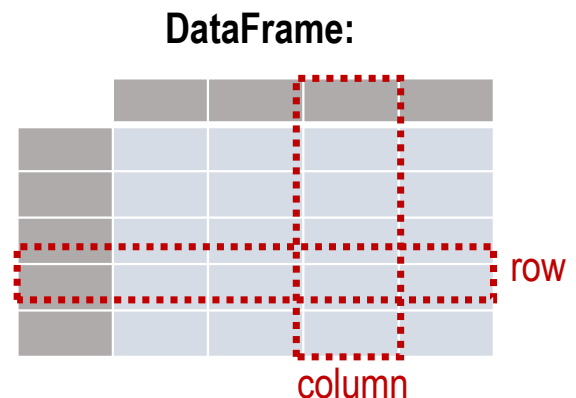
What is Pandas?

- The go-to Python library for handling and analyzing data in almost all data science tasks.
- Works with tabular data using **DataFrames**, similar to Excel spreadsheets or database tables.
- Simplifies data cleaning, transformation, and complex data analysis.
- Reading and writing data from CSV, Excel, and other commonly used formats in research



What is a DataFrame?

- 2-dimensional data structure
- Has labeled axes (rows and columns)
- Rows and columns are indexed for efficient data retrieval



Introduction to Pandas and DataFrames



1. Import packages

```
import numpy as np
import pandas as pd
```

2. Import the Iris dataset into a Pandas DataFrame

```
url = 'https://github.com/anolte-DSC/Python_for_Earth_Sciences/blob/main/Quickstart_Python_Jupyter/Datasets/Iris.csv'
data = pd.read_csv(url)
```

3. Explore the data

Plot the first 5 rows of the dataset.

```
data.head(5) # note: indexing starts at 0!
```

How many samples per species?

```
data['Species'].value_counts()
```

Plot basic statistics of the dataset.

```
data.describe()
```

Introduction to Pandas and DataFrames

4. Data manipulation

Check for missing values.

```
data.isnull().sum()
```

Create a new feature as a function of other features.

```
data['PetalAreaCm2'] = data['PetalLengthCm'] * data['PetalWidthCm']  
data.head(5)
```


Introduction to NumPy and arrays

What is NumPy?

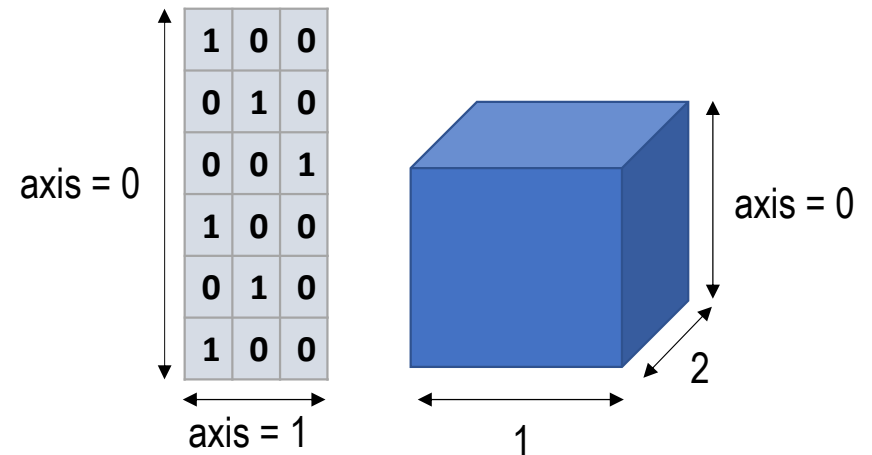
- The fundamental library for numerical computing.
- Support for large, multi-dimensional arrays and matrices.
- Offers a vast collection of mathematical functions to operate on these arrays.
- Enables efficient and fast computations, often used as a base for other scientific libraries like Pandas.



What is an array?

- A data structure for storing elements of the same type.
- Can be one-dimensional (like a list) or multi-dimensional (like a matrix).
- Supports vectorized operations, meaning you can perform operations on entire arrays without using loops.

Array with
shape=(6,3):



Introduction to NumPy and arrays

5. Data analysis with NumPy

Convert DataFrame to NumPy array.

```
# Factorize the species column to convert it into integers:
data_int = data.copy()
data_int['Species'], unique_species = pd.factorize(data['Species'])
species_mapping = dict(enumerate(unique_species)) # mapping of species to integers
print("Species to integer mapping:", species_mapping)
# Receive NumPy array from DataFrame
data_array = data_int.iloc[:, 1:].values
data_array
```

Similar calculations possible with NumPy & Pandas:

```
species = data_array[:, -2] # select the column with the species
unique_species, counts = np.unique(species, return_counts=True)
print("Species and their counts:", dict(zip(unique_species, counts)))
```

Introduction to NumPy and arrays

Compute mean and standard deviation for each feature.

```
mean_values = np.mean(data_array, axis=0)
std_deviation = np.std(data_array, axis=0)
mean_values = mean_values.round(2)
std_deviation = std_deviation.round(2)
print("Mean of each feature:", mean_values)
print("Standard deviation of each feature:", std_deviation)
```

Calculate Pearson correlation with continuous numerical variables.

```
correlation_matrix = np.corrcoef(data_array.T) # Transpose due to expectation for the input format: variables (features) as rows, observations as columns
print("Correlation matrix:\n", correlation_matrix)
```

Select data.

```
filtered_data = data_array[data_array[:, 2] > 2.0] # Example: Filter data where petal length (third column) is greater than 2.0 cm
print('Full data sample size:', data_array.size)
print('Filtered data sample size:', filtered_data.size)
```

Pandas DataFrames vs. NumPy arrays

Feature	Pandas DataFrames	Numpy Arrays
Purpose	Data manipulation and analysis	Numerical computations
Data Structure	Heterogeneous (different data types)	Homogeneous (same data types)
Operations	Filtering, grouping, merging, reshaping	Mathematical, linear algebra, simulations
Use cases	Data cleaning, exploratory analysis	Scientific computing, machine learning