



# Machine Learning: Applications and Practices

## Lecture 1

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University of Louisiana at Lafayette

# Welcome!

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- **Welcome all participants from four universities:**
  - University of Louisiana at Lafayette
  - Southern University
  - University of South Alabama
  - Western Kentucky University
  - Others

# Course Information

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- **Class Meeting Time:**

- Wednesday: 10: 00am to 11:15am (Lecture series)
- Friday: 10: 00am to 11:30am (Hands-on series)

- **Prerequisite:**

- Have a Windows OS laptop
- Know the basic of Python programming

- **Course Assistants:**

- Mr. Yihe Zhang
- Mr. Jiadong Lou

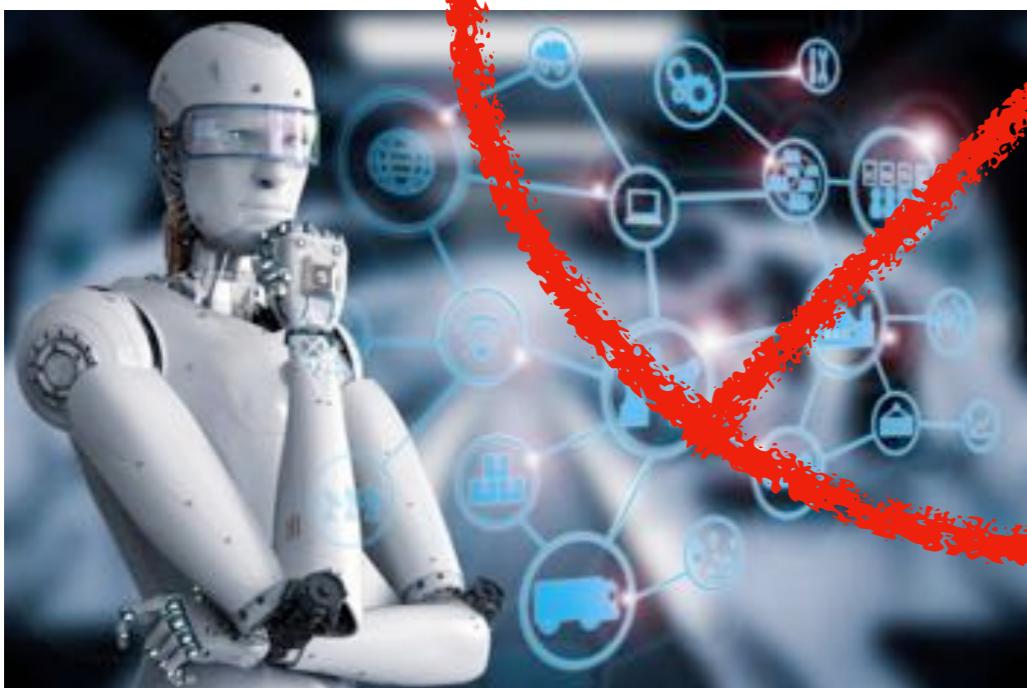
- **Course Website:**

- [https://people.cmix.louisiana.edu/yuan/2022\\_Summer\\_Tutorial\\_Courses.html](https://people.cmix.louisiana.edu/yuan/2022_Summer_Tutorial_Courses.html)
- *Please don't distribute/spread Twitter Dataset*

# What's Our Goals?



# We are not ambitious...



# Our Goals

This is just an entry level of Machine Learning course!

No credits, no grading!

1. Learning the fundamental knowledge

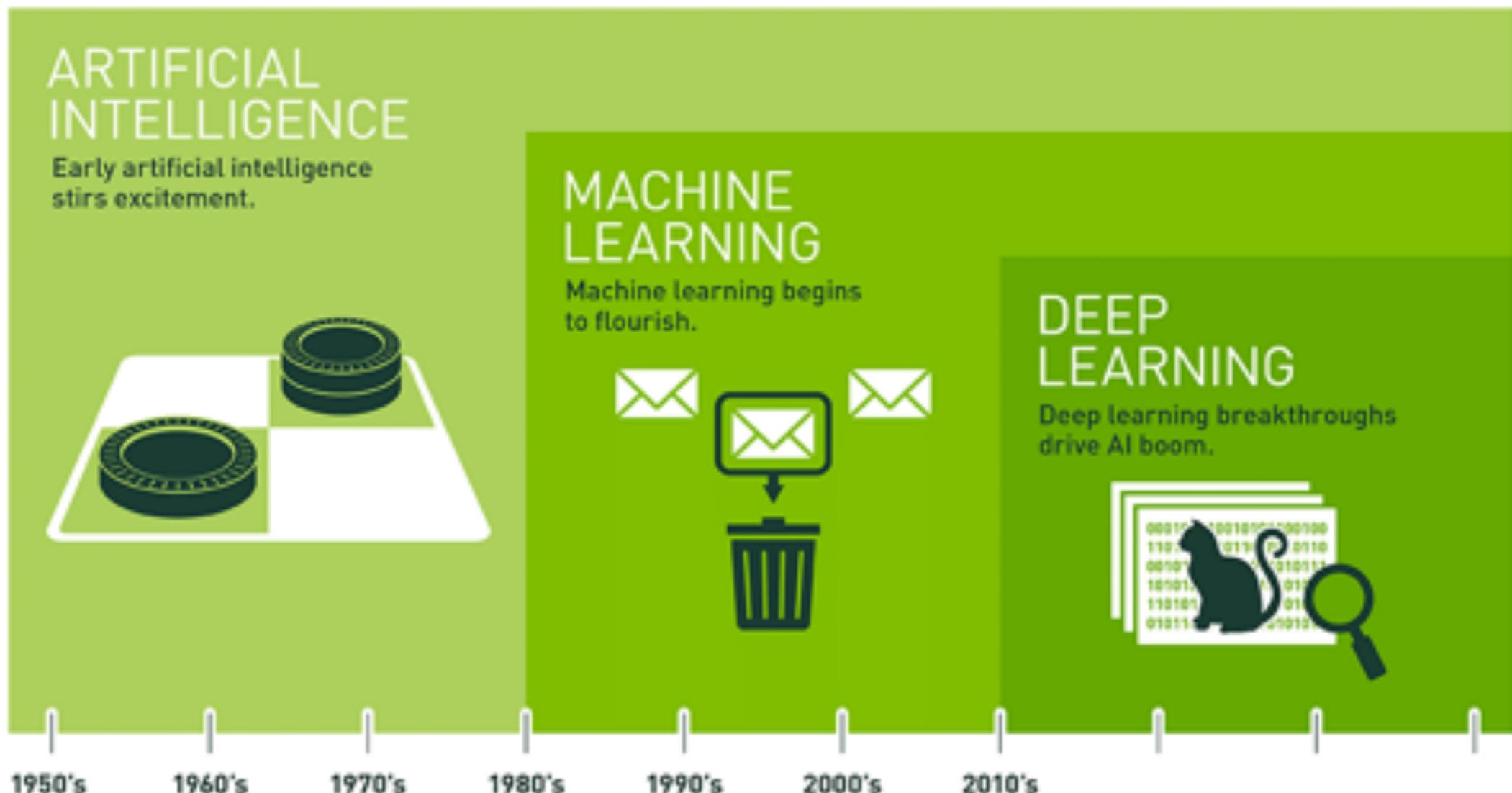
2. Coding practice for Python

3. Practicing on real-world data

# My Suggestions

Please attend each lecture and hands-on;  
Otherwise, you will be  
**lost!**

# AI History



Since an early flush of optimism in the 1950s, smaller subsets of artificial intelligence – first machine learning, then deep learning, a subset of machine learning – have created ever larger disruptions.

Source from: <https://blogs.nvidia.com/blog/2016/07/29/whats-difference-artificial-intelligence-machine-learning-deep-learning-ai/>

# AI and ML

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- **Artificial Intelligence (AI)**
  - Role of Statistics: Inference from a sample.
- **Machine Learning (ML)**
  - Arthur Samuel (1959): Machine Learning: Field of study that gives computers the ability to learn without being explicitly programmed.
  - Tom Mitchell (1998): Well-posed Learning Problem: A computer program is said to learn from experience with respect to **some task T** and some **performance measure P**, if its performance on T, as measured by P, improves with **experience E**.

# What is Machine Learning?

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- Study of *Algorithms* that *improve* their *performance* at some *task* with *experience*.
- **Role of Computers:**
  - Having efficient algorithms to solve the optimization problems to learn models
  - Learning Models for unknown and changing worlds
  - Representing and Evaluating the model for inference.

# What is Machine Learning?

Experience



This is a “cat”

Algorithms



Tasks



Not a “cat”



Not a “cat”



“cat”



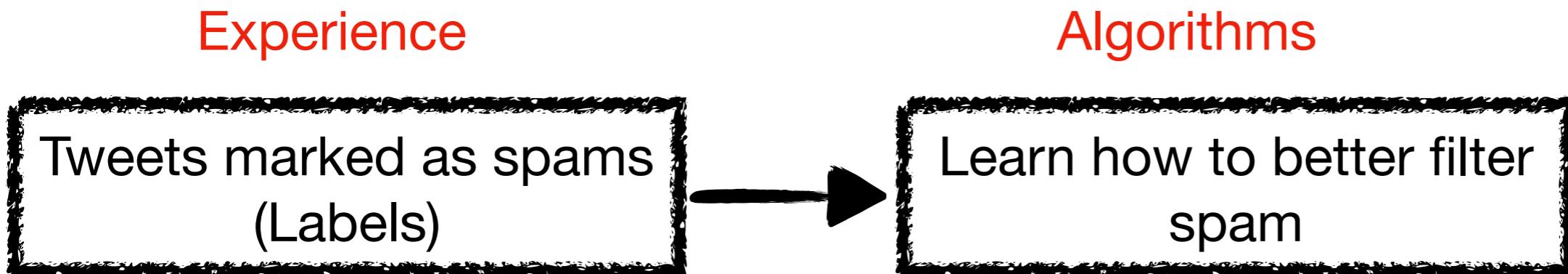
# Spam Classification Example

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- *Suppose Twitter server watches which tweets marked as spam message. Based on this information, he will learn how to better filter spam.*

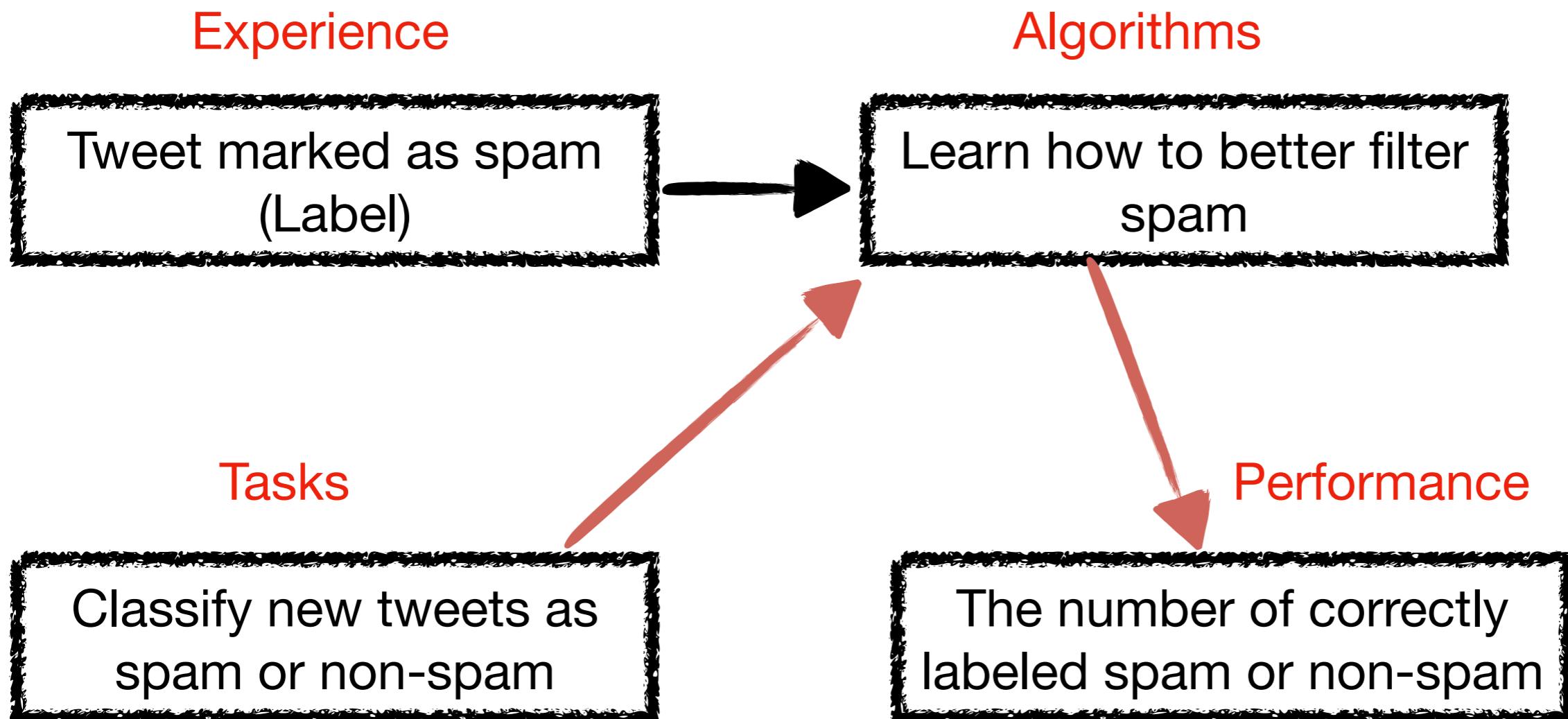
# Spam Classification Example

- Suppose Twitter server watches which tweets marked as spam message. Based on this information, he will learn how to better filter spam.



# Spam Classification Example

- Suppose a Twitter server watches which tweets are marked as spam messages. Based on this information, it will learn how to better filter spam.



# Weather Prediction Example

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- Suppose a Mesonet station monitors the weather conditions for the past several years, then based on this information, a computer program can learn and predict the weather conditions in next several days.

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Past several years'  
observation

Experience

Algorithms



# Weather Prediction Example

- Suppose a Mesonet station monitors the weather conditions for the past several years, then based on this information, a computer program can learn and predict the weather conditions in next several days.



Past several years' observation

Last one week's observation

Tasks



Next week



# Machine Learning ~ Looking for a Function

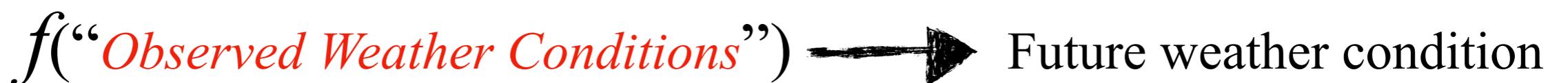
- Image recognition



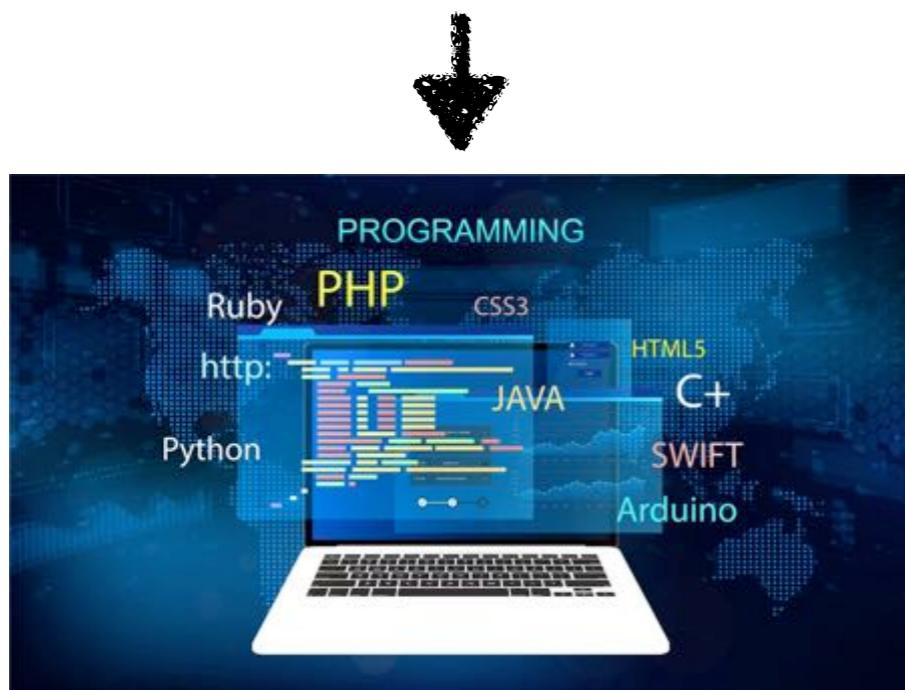
- Spam classification



- Weather prediction



# Machine Learning ~ Training Framework



Training  
Data

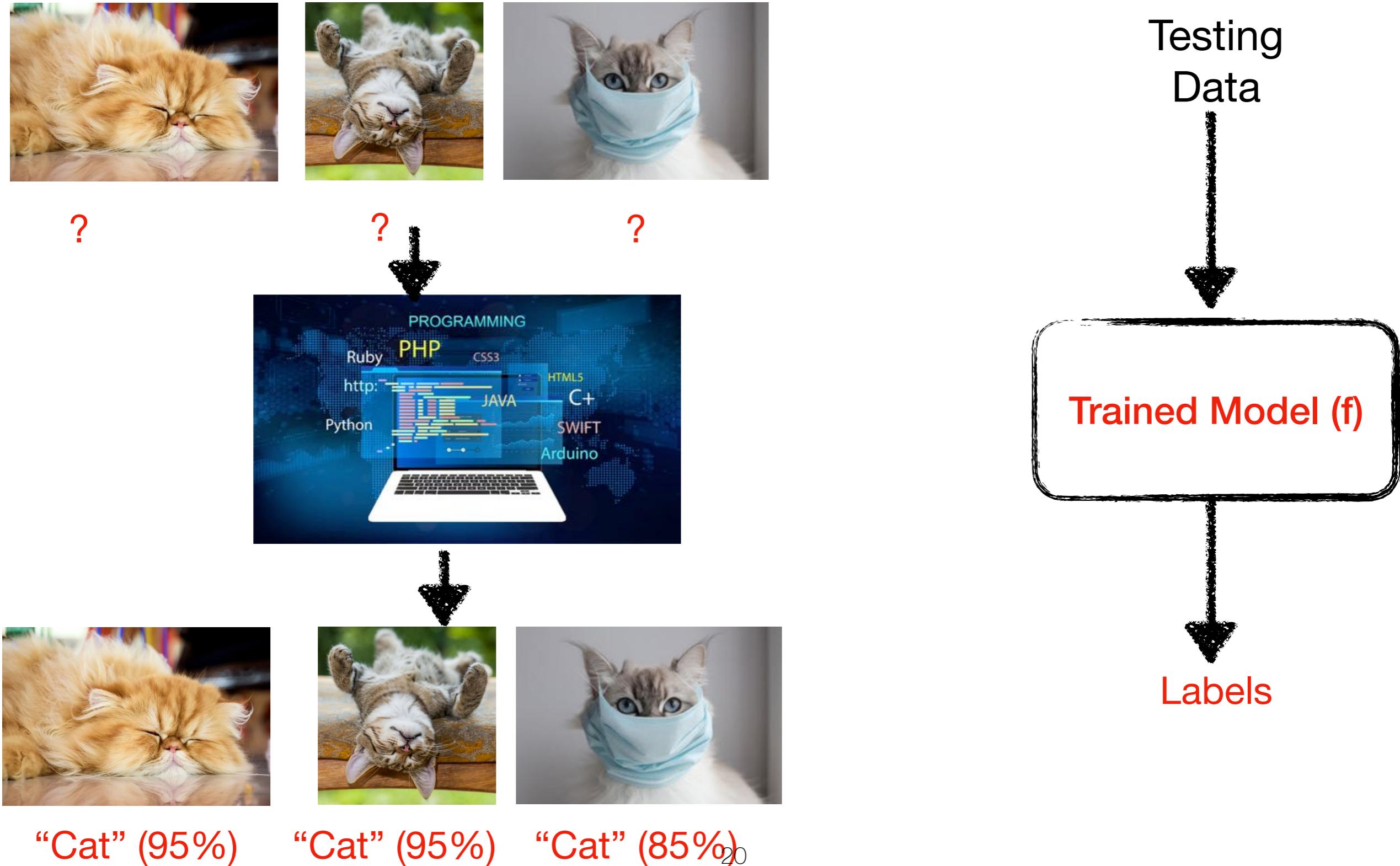
A set of functions  
(models)  $f_1, f_2, \dots$

Goodness of  
function  $f$

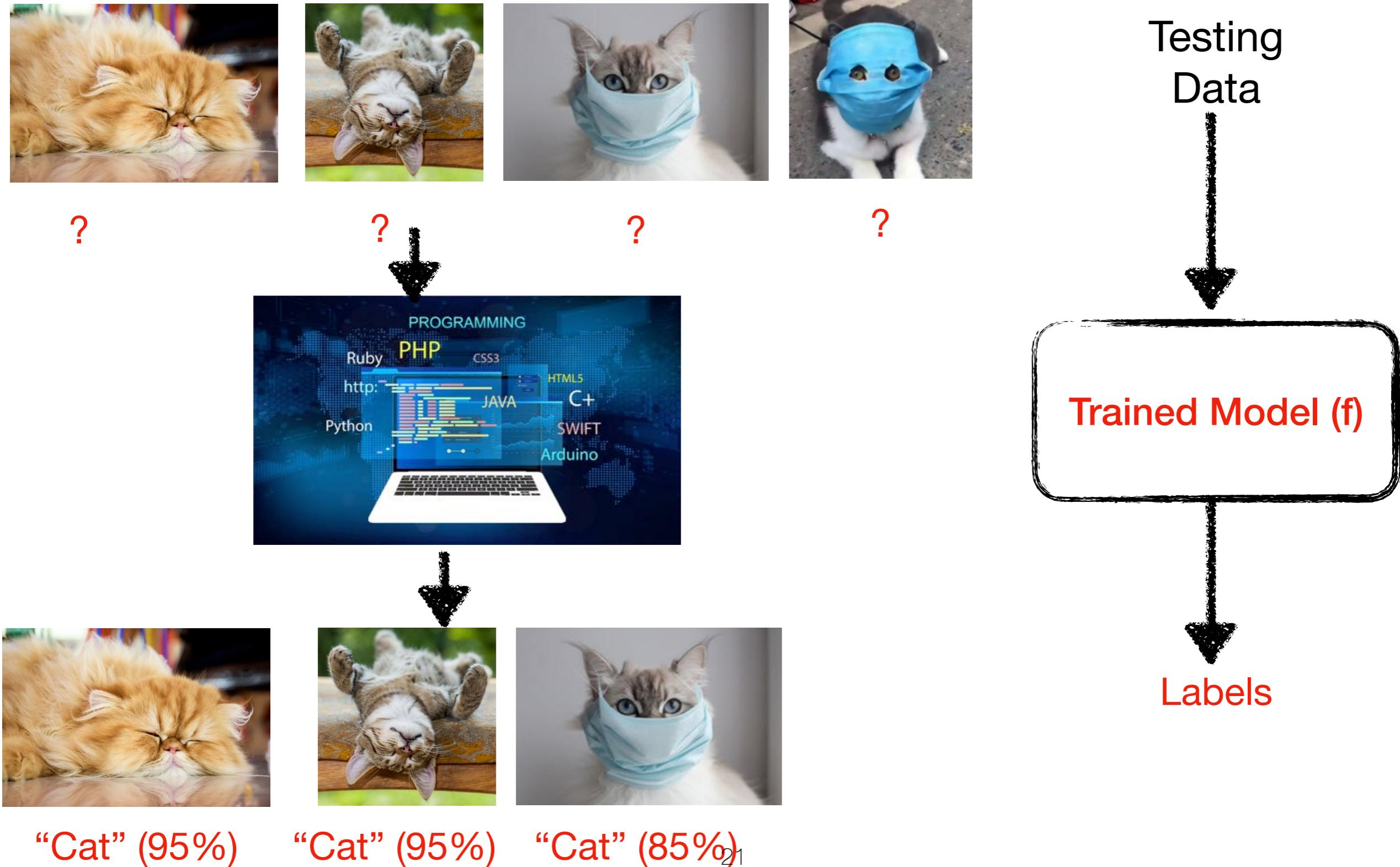
Pick the “best”  
function  $f^*$

Trained Model

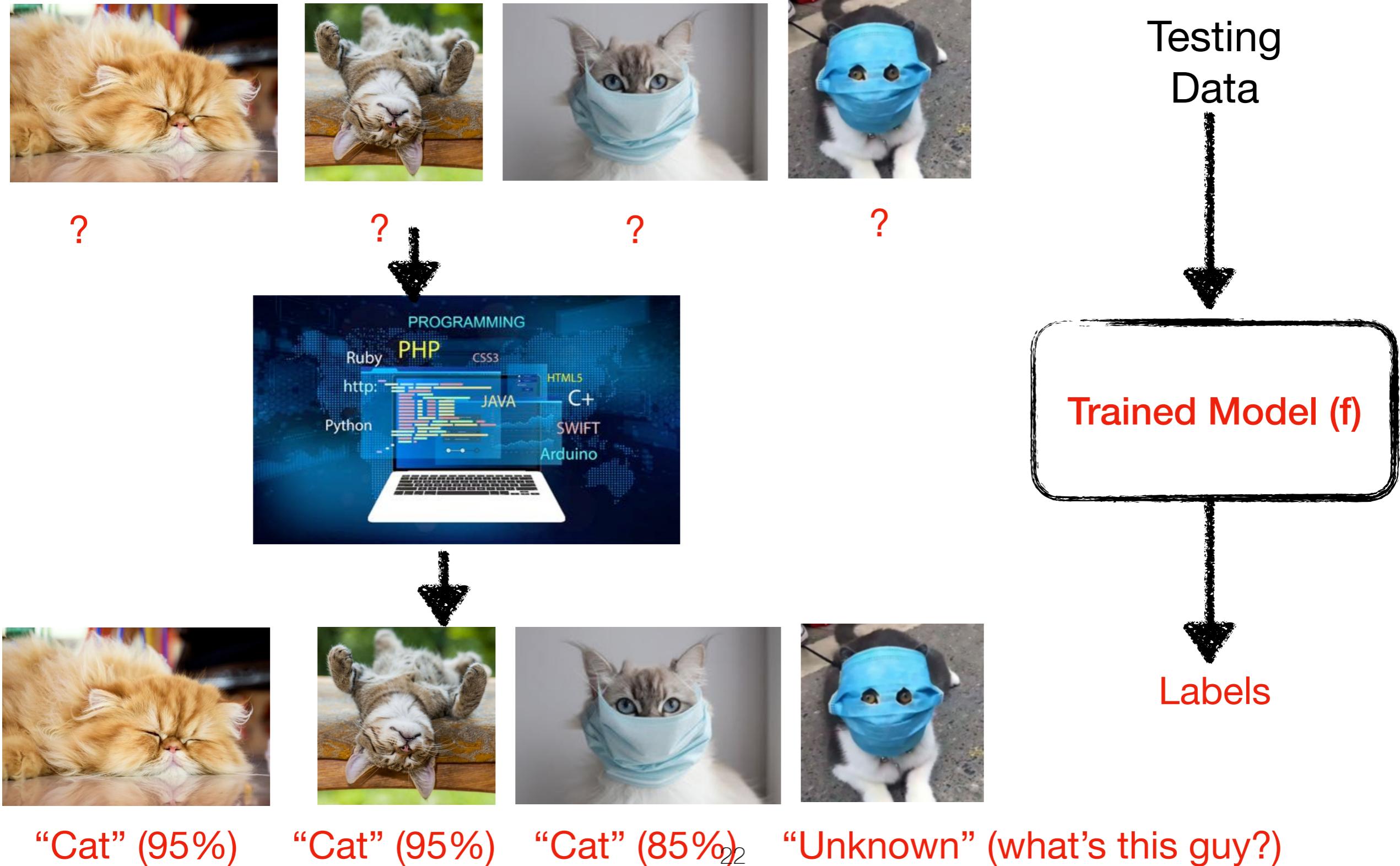
# Machine Learning ~ Testing Framework



# Machine Learning ~ Testing Framework

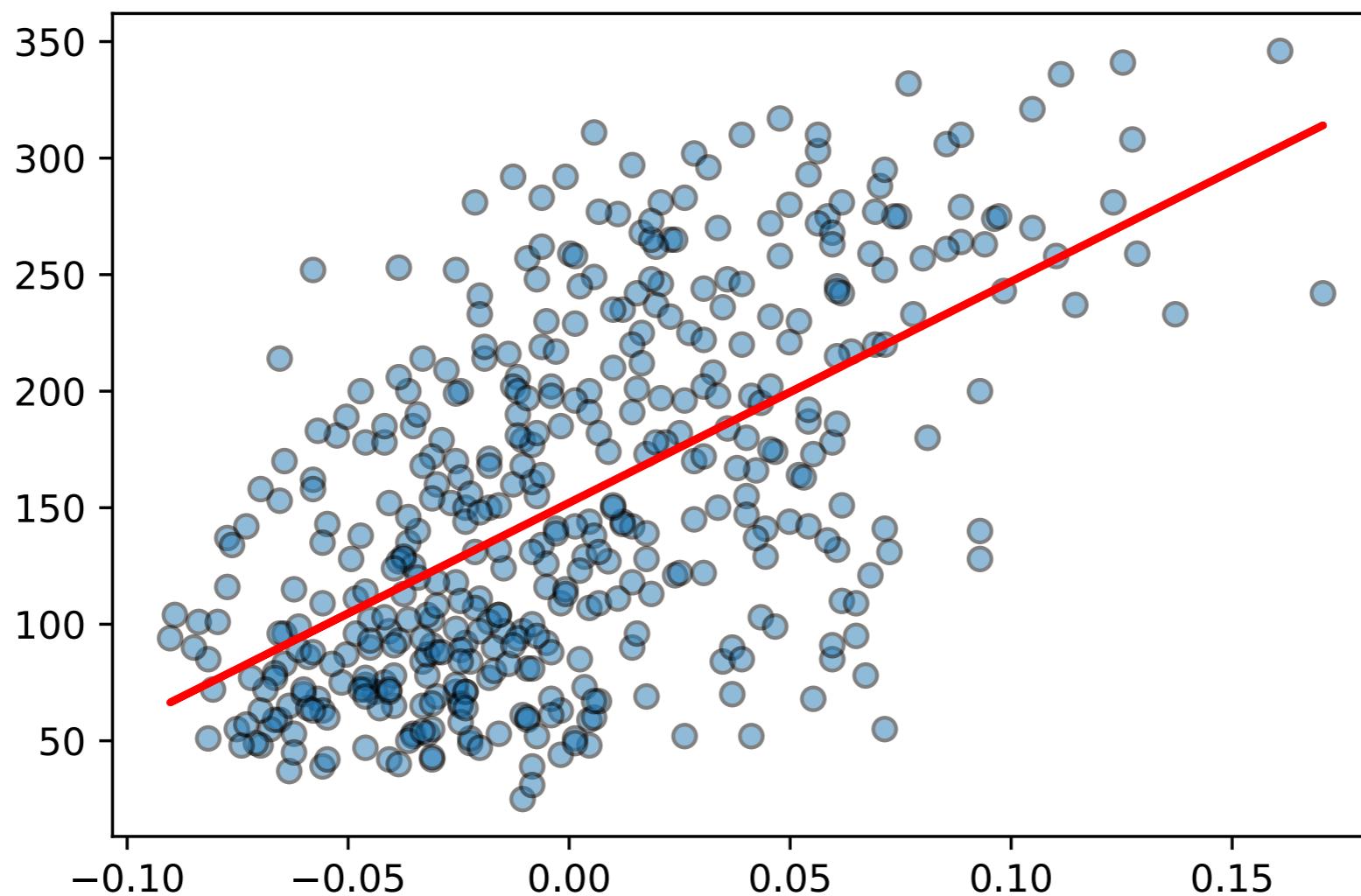


# Machine Learning ~ Testing Framework



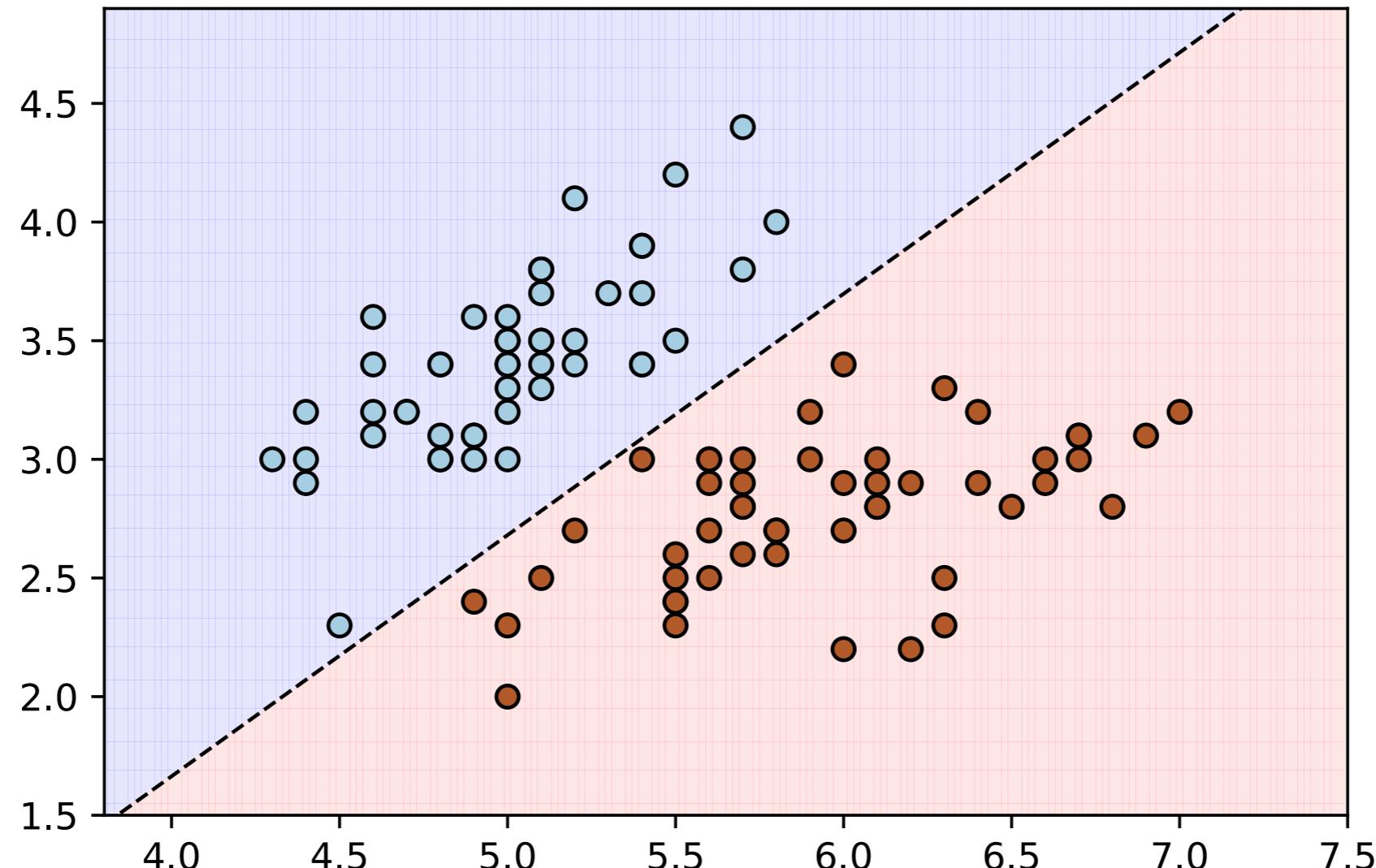
So far,  
you can see finding **a suitable function** is the  
core of machine learning

# Linear Regression



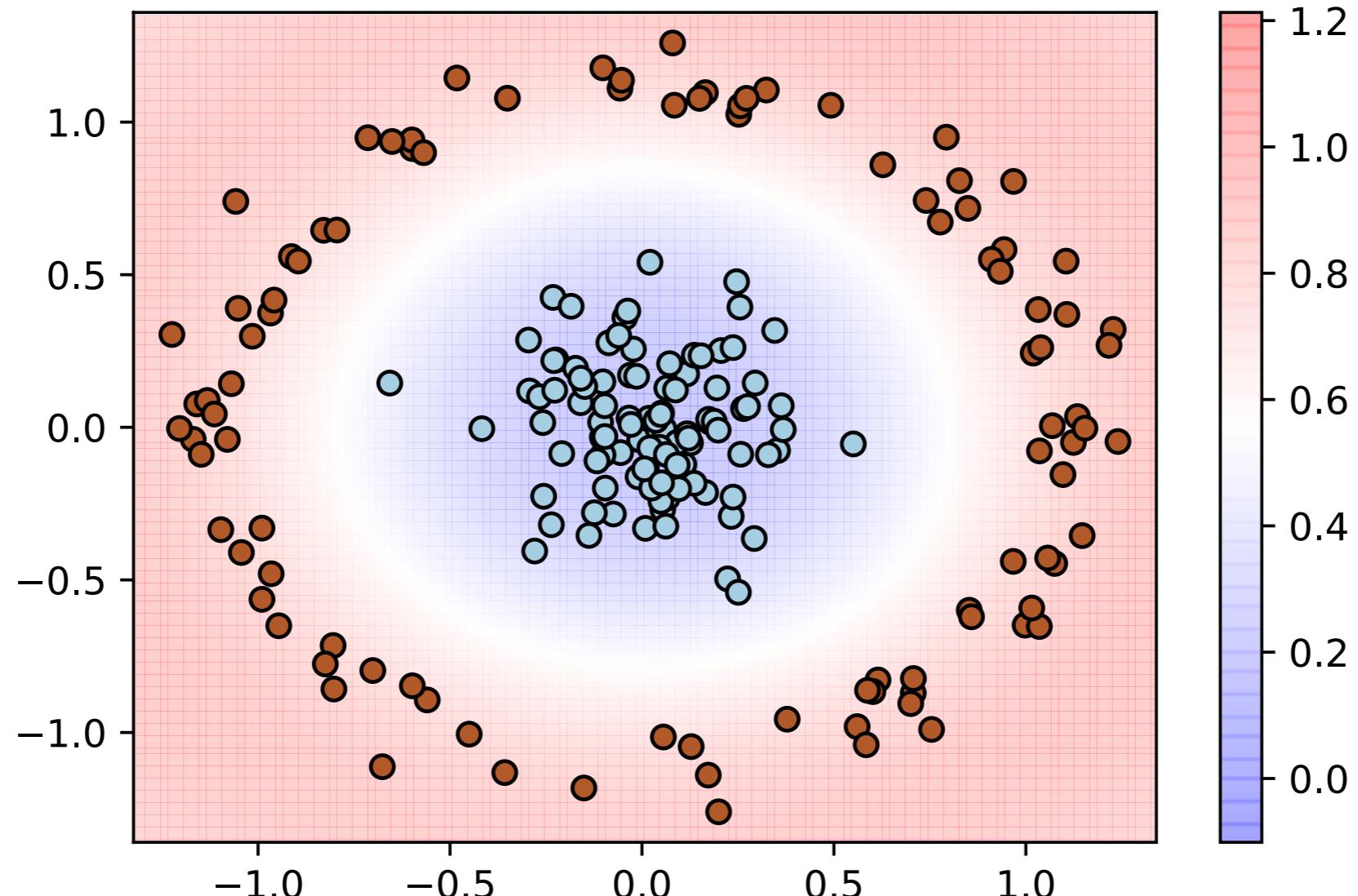
Finding a function that best fits the curve

# Logistic Regression



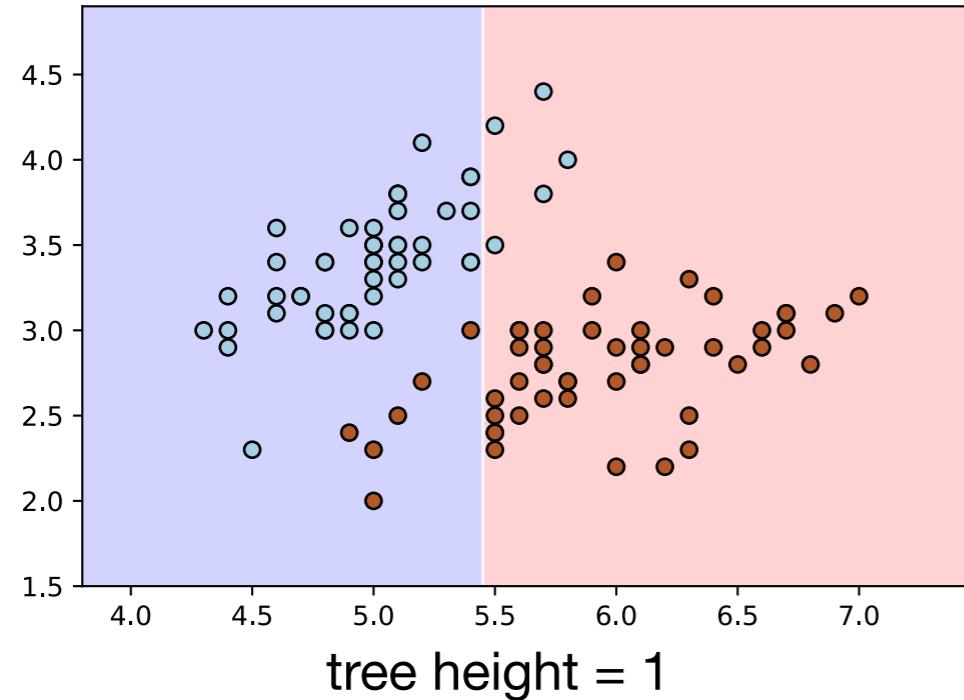
A function is used to define the boundary line

# Supported Vector Machine (SVM)



The boundary curves are non-linear.

# Decision Tree

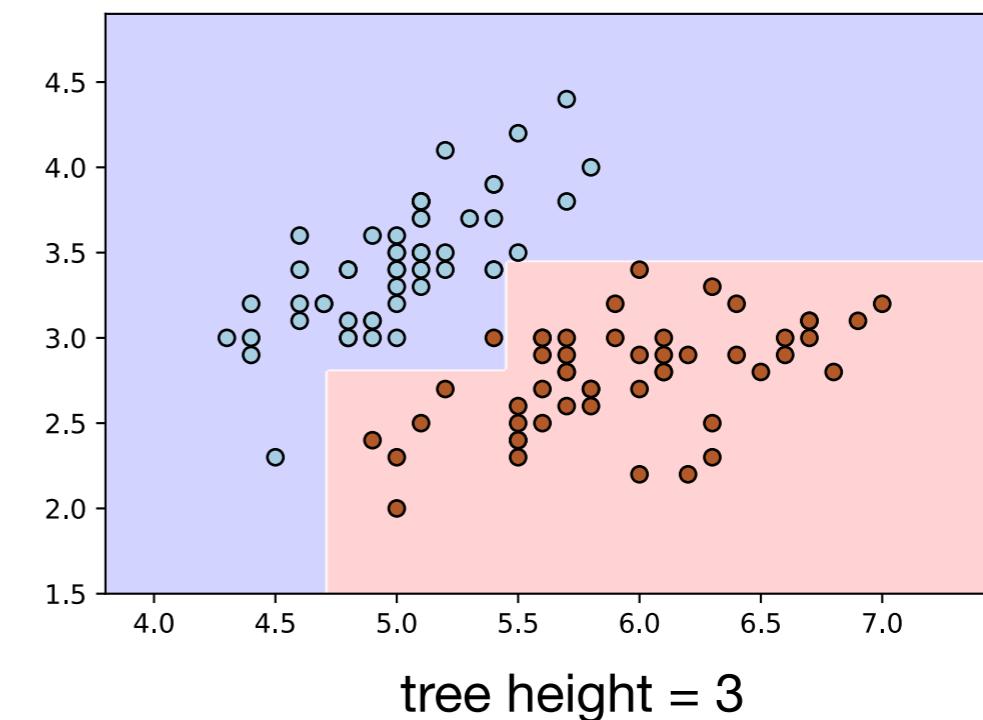
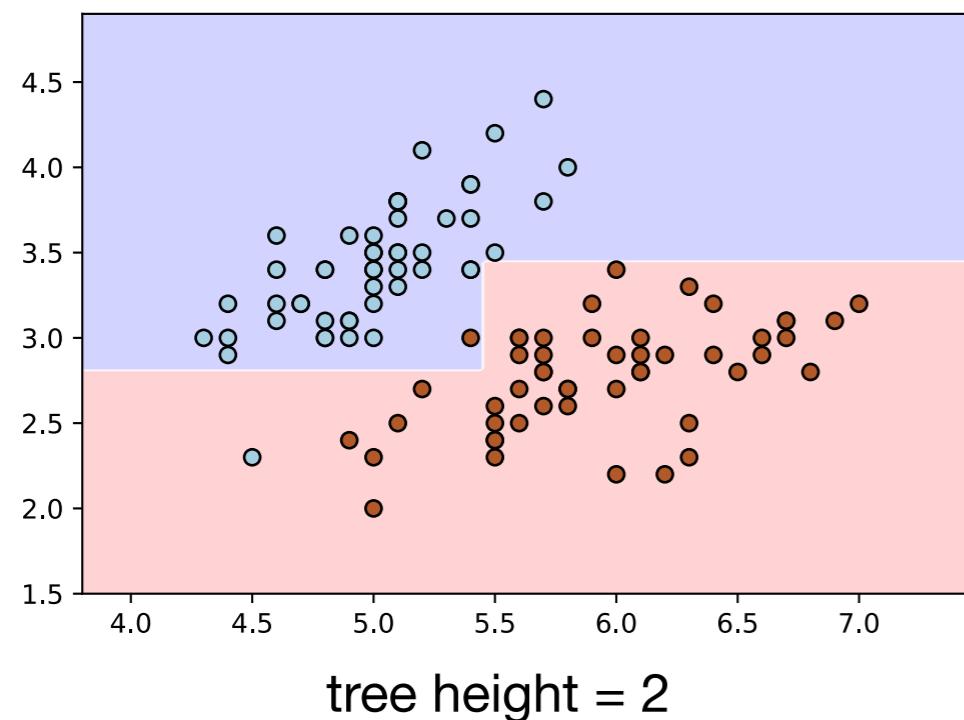


$X[0] \leq 5.45$   
gini = 0.5  
samples = 100  
value = [50, 50]

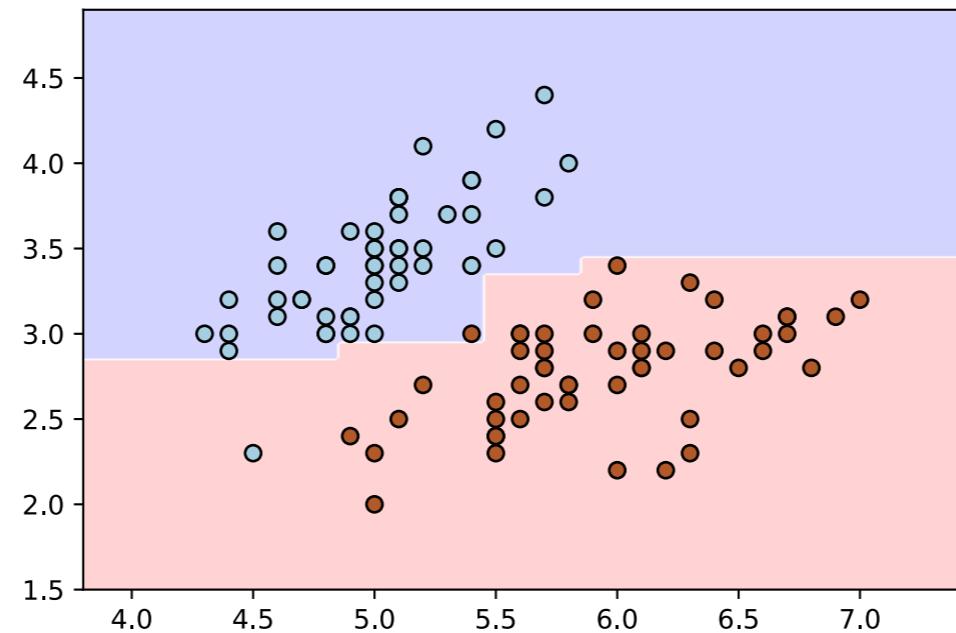
gini = 0.208  
samples = 51  
value = [45, 6]

gini = 0.183  
samples = 49  
value = [5, 44]

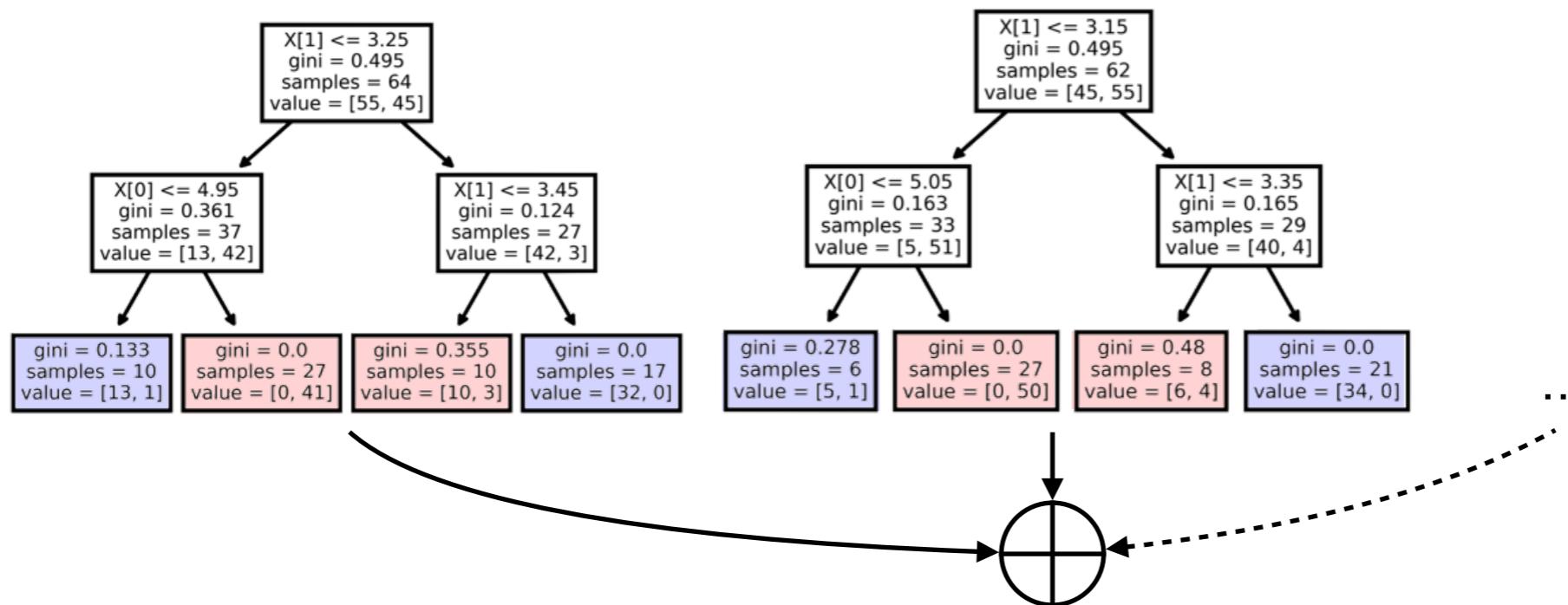
Decision tree with height 1



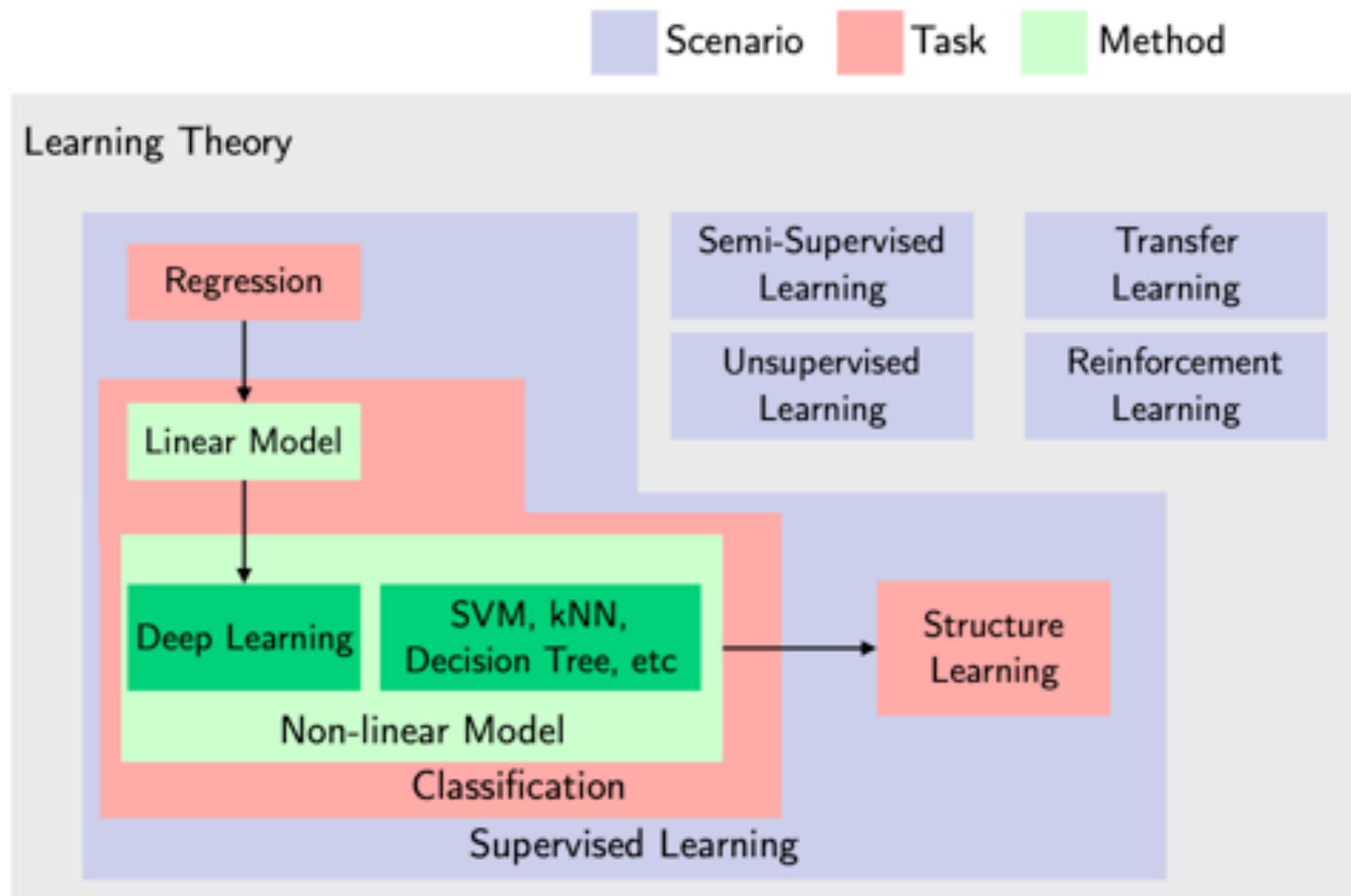
# Random Forest



number of trees = 10, tree height = 2



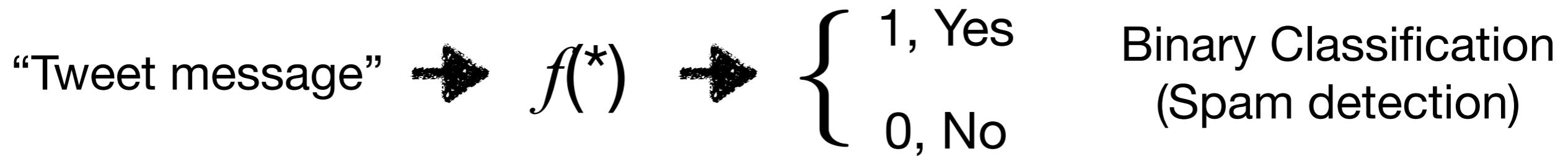
# Learning Map



# Supervised Learning

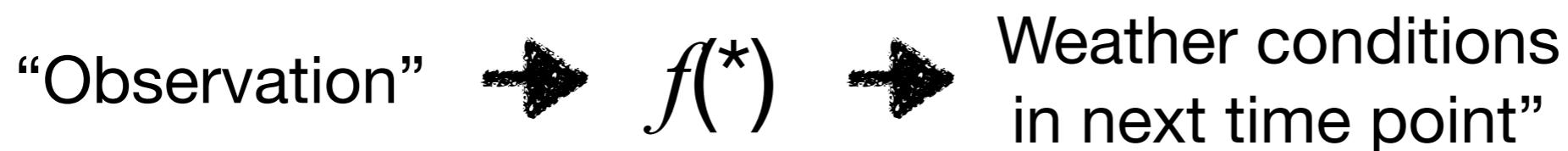
- **Classification**

- Each element in the sample is labeled as belonging to some class. No order among classes.



- **Prediction**

- Elements in the sample have the inherent relationships to weather condition at some time point.



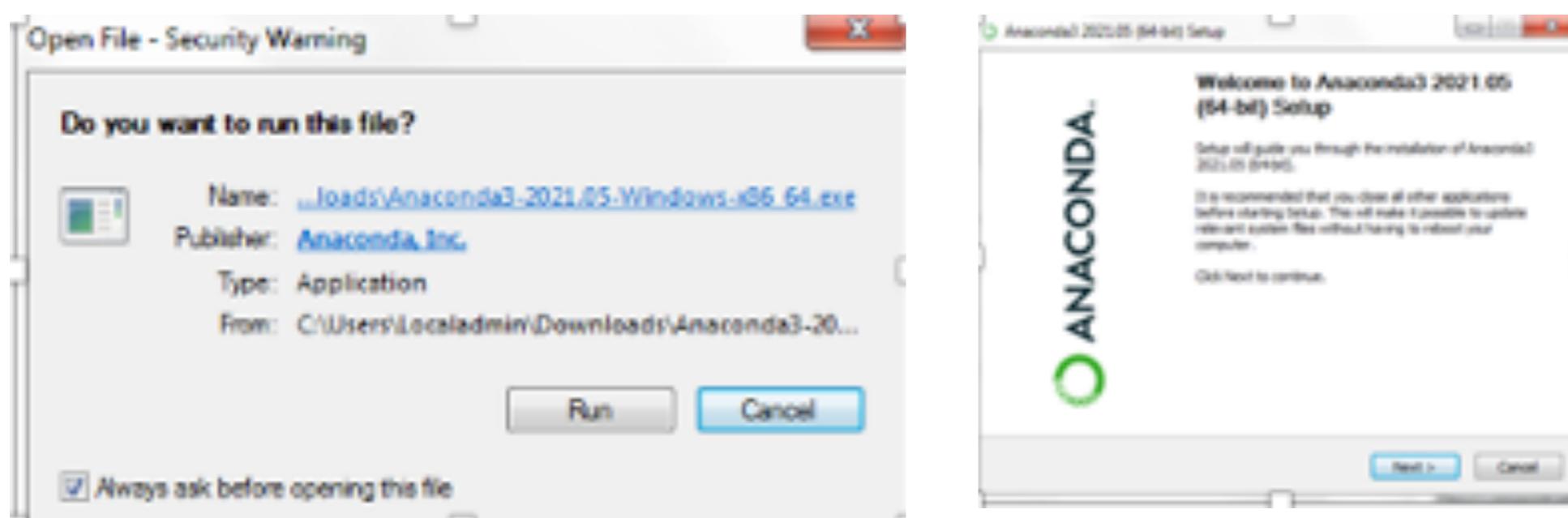
# Before starting, we need to know Python

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- Python provides a set of libraries including different ML packages
- Standard libraries provide the ready-to-use implementation of algorithms
- The scikit-learn is the one we will use in this course

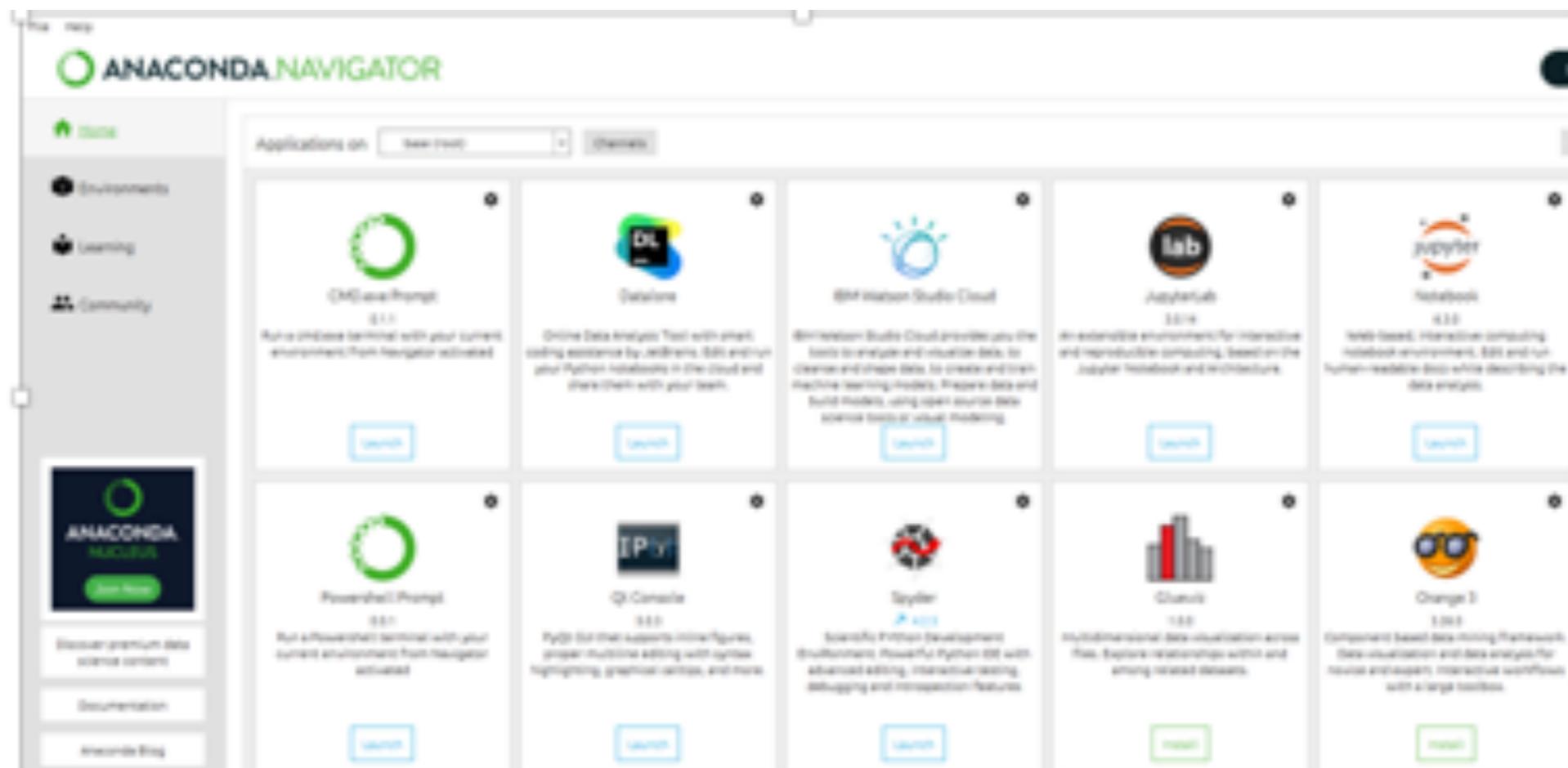
# Installing Anaconda Navigator

1. Browse <https://docs.anaconda.com/anaconda/install/windows/>
2. Click on Download the Anaconda installer
  - Check your OS bit version: Start button->Settings->System->About: Device specification System Type
  - Click on (your\_OS\_bit\_version)-Bit Graphical Installer, e.g., 64-Bit Graphical Installer, and click on save (will take a while for downloading)
3. Double click the installer to launch and click on Run for installation
4. Click on Next -> I Agree -> Next ->Next->Install (for default settings)



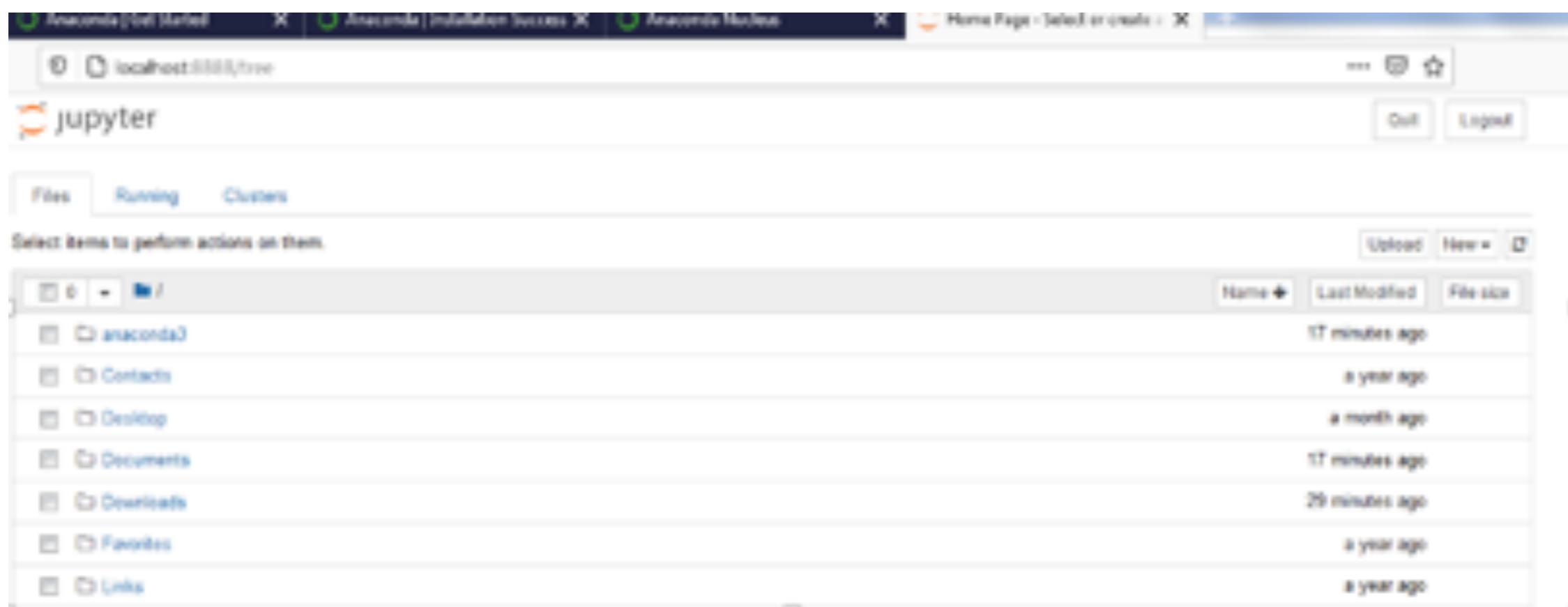
# Installing Anaconda Navigator (Continuing...)

5. Click Next->Next->Finish to complete the installation (registration is not essential for operation).
6. Open Anaconda Navigator: It will pop up an icon in the status bar.
7. Click on the icon and click on the launch button of Jupyter Notebook.



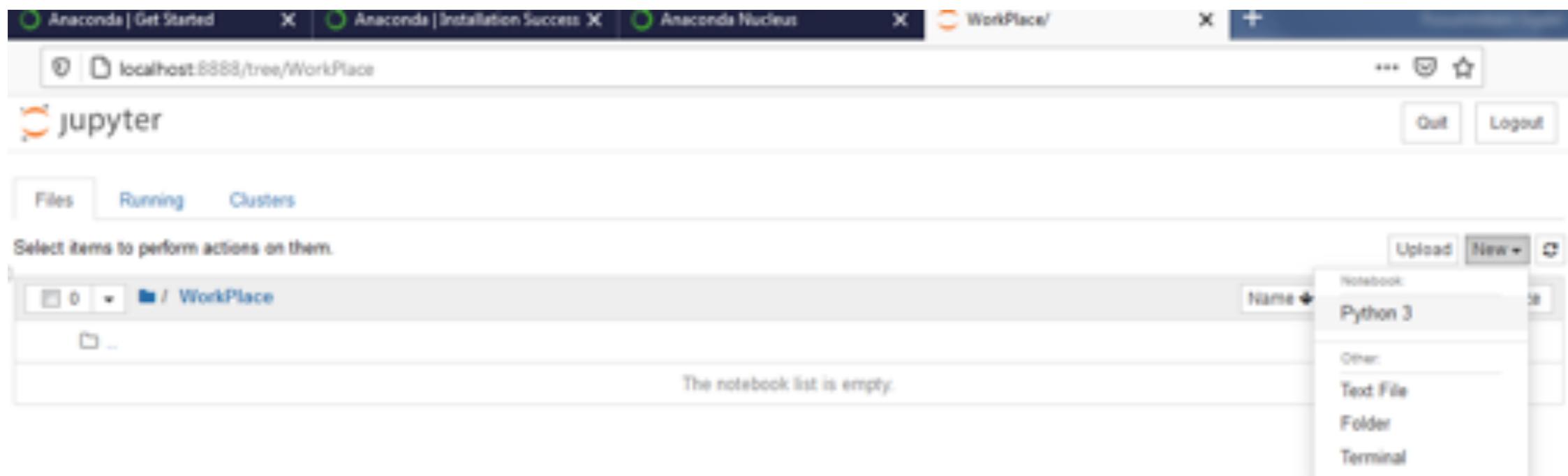
# Installing Anaconda Navigator (Continuing...)

8. It will open the browser and show your files and directory (folders) from C:\Users\Your\_user\_account.
9. For the time being, create a working directory at C:\Users\Your\_user\_account\[yourWorkingDirectory]



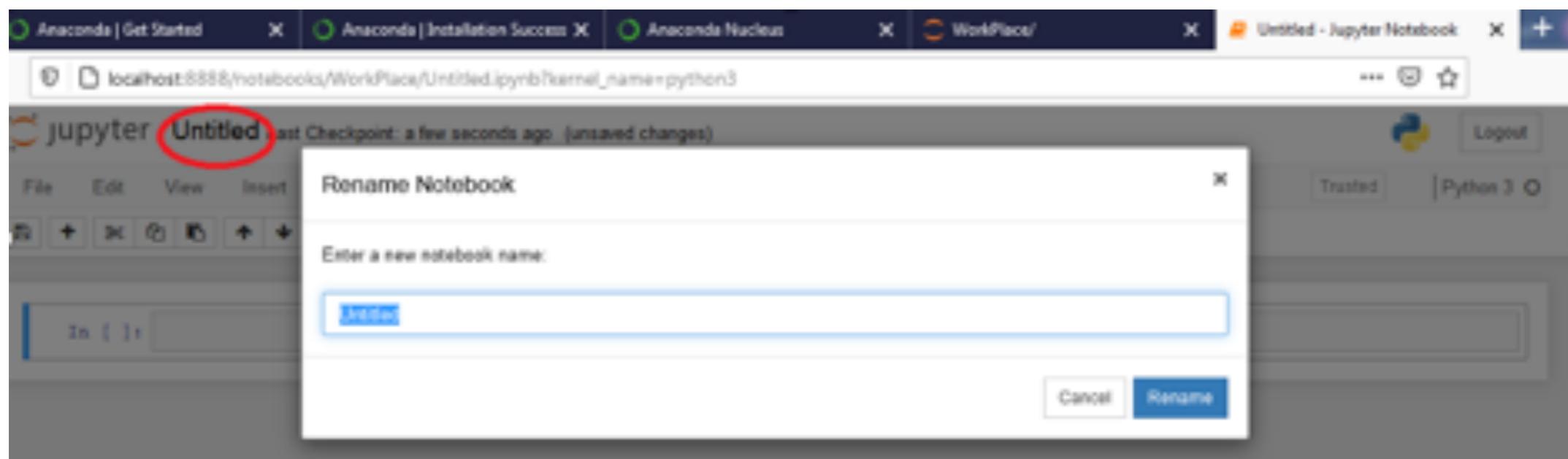
# Installing Anaconda Navigator (Continuing...)

10. Click on your working directory (in my case, it is ‘workPlace’). It will take you to a new window.
11. Click on the New dropdown button (on the right side) and click on the Python 3.

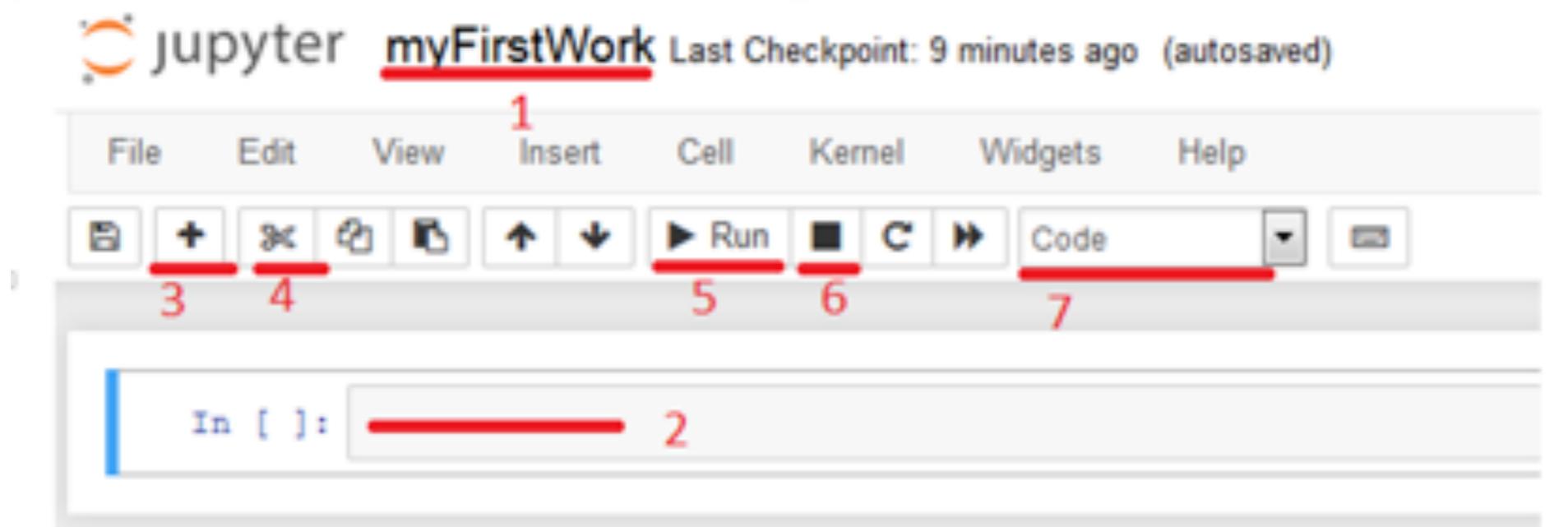


# Installing Anaconda Navigator (Continuing...)

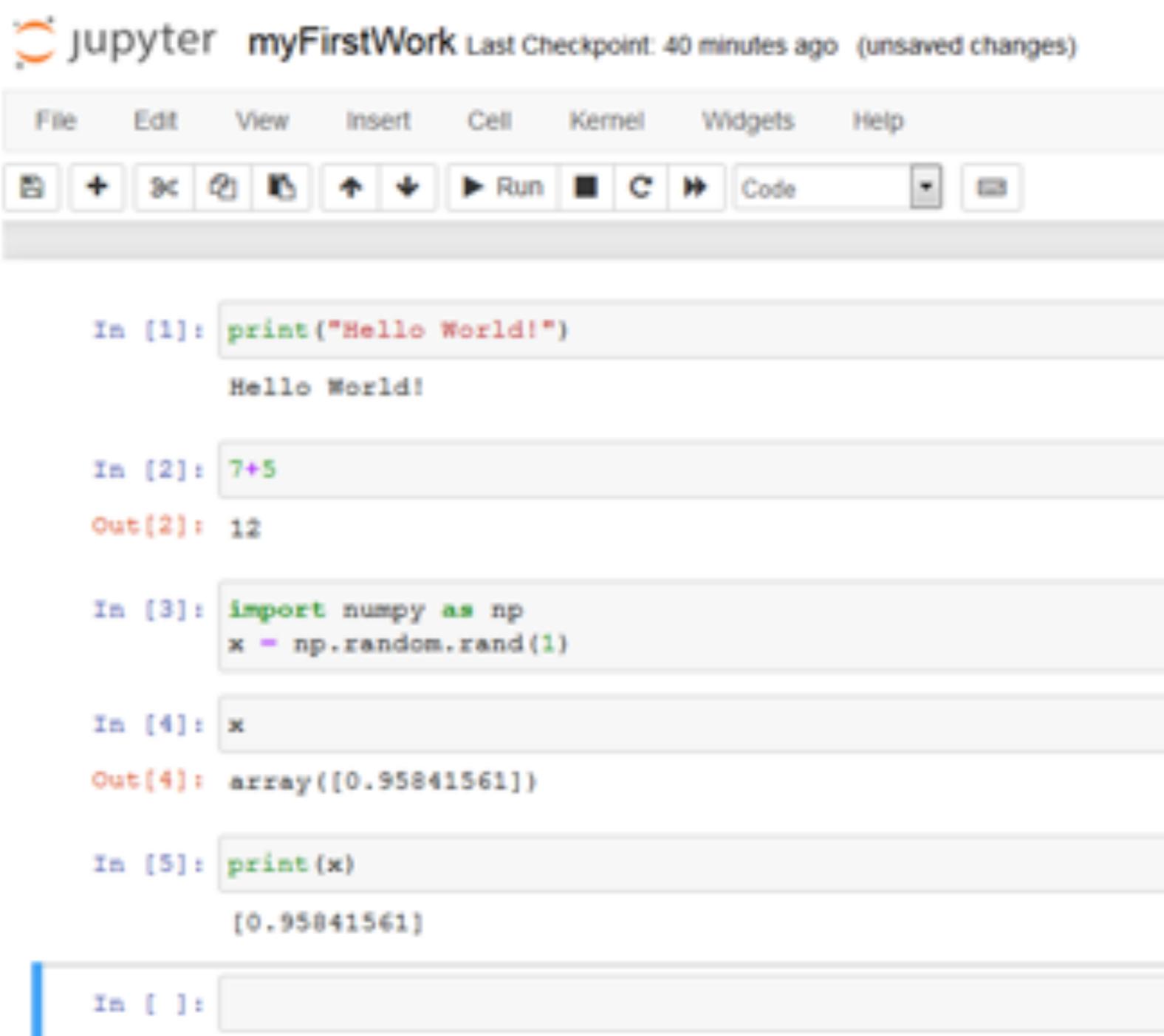
12. It will open a new page in the browser with the Untitled – Jupyter Notebook page. To change the name, click on the Untitled label (on the top left) and rename your file.



# Frequently Used buttons



# Examples

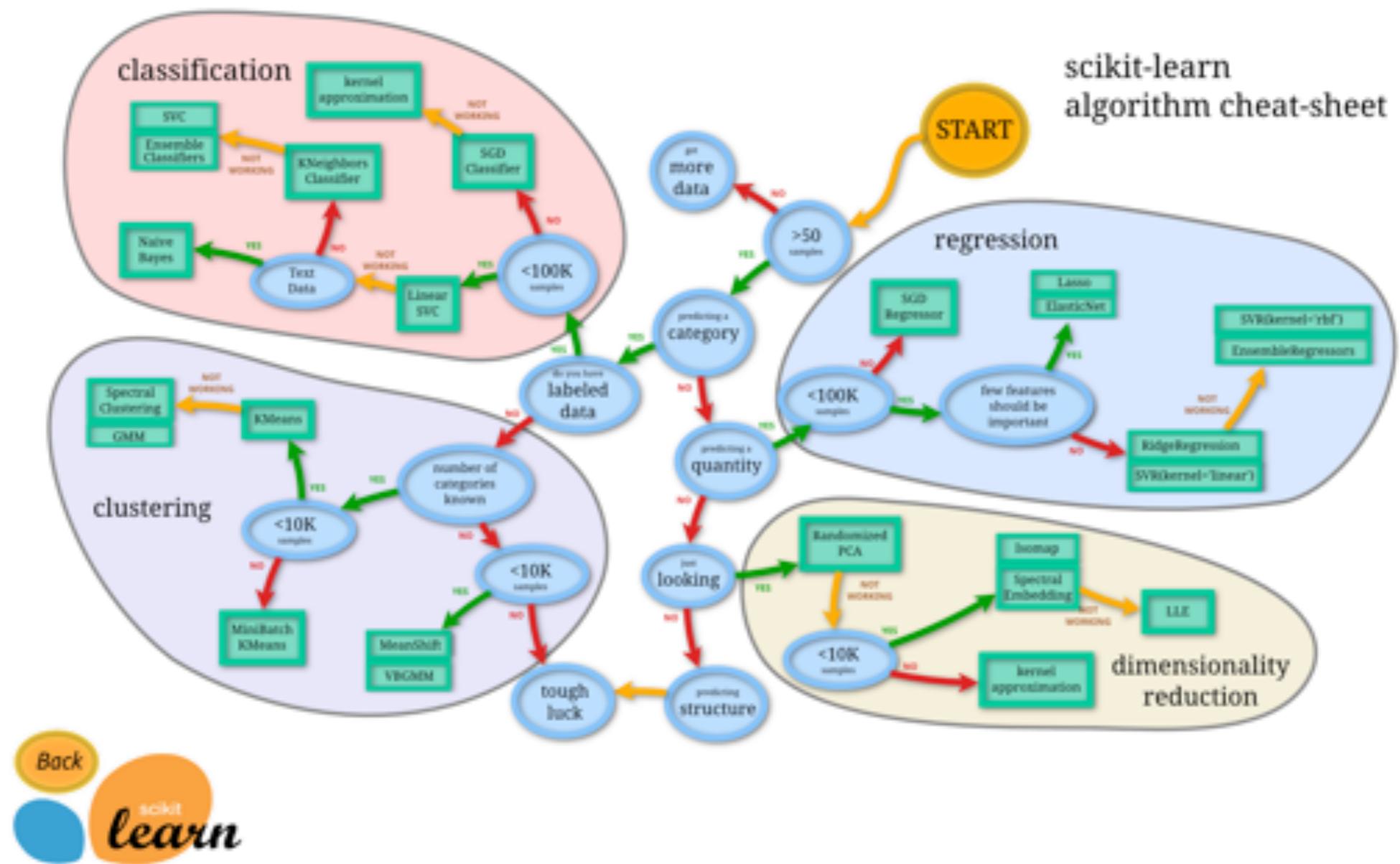


The screenshot shows a Jupyter Notebook interface with the title "jupyter myFirstWork Last Checkpoint: 40 minutes ago (unsaved changes)". The menu bar includes File, Edit, View, Insert, Cell, Kernel, Widgets, and Help. Below the menu is a toolbar with icons for file operations, cell creation, run, and code execution.

The notebook contains the following code cells:

- In [1]: `print("Hello World!")`  
Hello World!
- In [2]: `7+5`  
Out[2]: 12
- In [3]: `import numpy as np  
x = np.random.rand(1)`
- In [4]: `x`  
Out[4]: array([0.95841561])
- In [5]: `print(x)`  
[0.95841561]
- In [ ]:

# Scikit-learn



Source: [https://scikit-learn.org/stable/tutorial/machine\\_learning\\_map/index.html](https://scikit-learn.org/stable/tutorial/machine_learning_map/index.html)

# Example 1

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```
from sklearn import svm

X = [[0, 1], [1, 2], [2, 1], [2, 3], [1, 3], [2, 2]]

y = ['a', 'a', 'b', 'b', 'a', 'b']

clf = svm.SVC()

clf.fit(X, y)

result1 = clf.predict([[3, 1]])

print(result1)

result2 = clf.predict([[0, 2]])

print(result2)

['b']
['a']
```

## Example 2

```
from sklearn import svm
from sklearn.datasets import load_iris
#iris dataset contains 150 samples, each has 4 features
X, y = load_iris(return_X_y = True)
...
Parameter 'return_X_y = True' is required in
load_iris() function to get the sample and label data in
seperate variables.
...
print("The size of the sample:", X.shape)
print("First 5 samples:\n", X[0:5])
print("First 5 labels:\n", y[0:5])
clf = svm.SVC()
clf.fit(X, y)
result = clf.predict(X[45:55])
print("Predicted labels\n", result)
print("Actual labels\n", y[45:55])
```

The size of the sample: (150, 4)  
First 5 samples:  
[[5.1 3.5 1.4 0.2]  
 [4.9 3. 1.4 0.2]  
 [4.7 3.2 1.3 0.2]  
 [4.6 3.1 1.5 0.2]  
 [5. 3.6 1.4 0.2]]  
First 5 labels:  
[0 0 0 0 0]  
Predicted labels  
[0 0 0 0 0 1 1 1 1 1]  
Actual labels  
[0 0 0 0 0 1 1 1 1 1]

## More Resources

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Please check  
[https://people.cmix.louisiana.edu/yuan/2022\\_Summer\\_Tutorial\\_Courses.html](https://people.cmix.louisiana.edu/yuan/2022_Summer_Tutorial_Courses.html)  
for more examples

# Q&A

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Thank You!