

#### Tutorial for Assignment 1

COMP3314
Machine Learning

Xi CHEN

## Tutorial for Assignment 1

- Basics of Python
- Requirements of Assignment 1
- Guidelines for Assignment 1

- Python is an interpreted, high-level and general-purpose programming language [1]. The usage of Python is like that of other scripting languages, such as R and MATLAB.
- Currently, Python is one of the most popular programming languages [2].



<sup>[1]</sup> https://en.wikipedia.org/wiki/Python\_(programming\_language)

<sup>[2]</sup> https://spectrum.ieee.org/top-programming-languages-2022/

- Python provides many useful packages (also know as libraries or wheels) to process data, build machine learning models, do scientific computing, etc.
- Python is the first choice in various deep learning frameworks, such as PyTorch and TensorFlow.
  - O <u>NumPy</u>
  - SciPy library
  - scikit-learn
  - O Matplotlib
  - O <u>Pandas</u>
  - O <u>Tensorflow</u>
  - O <u>Jupyter Notebook</u>

- AlexNet, the first deep CNN published in 2012, is written in tens of thousands lines of C++.
- Nowadays, we could use 5-10 lines of code to build and train a CNN model with the help of some highly integrated python packages like <u>fast.ai.</u>

```
path = untar_data(URLs.PETS)/'images'

def is_cat(x): return x[0].isupper()

dls = ImageDataLoaders.from_name_func(
    path, get_image_files(path), valid_pct=0.2, seed=42,
    label_func=is_cat, item_tfms=Resize(224))

learn = vision_learner(dls, resnet34, metrics=error_rate)
learn.fine_tune(1)
```

- Packages that might be used in Assignment 1
  - scikit-learn
    - Use some implemented classifiers to train and predict on your data.
    - Tutorial: Documentation
  - o pandas
    - Load and process your dataset.
    - Tutorial: 10 minutes to pandas
  - Matplotlib
    - Draw figures for visualize your results.
    - Tutorial: Quick start for Matplotlib
  - $\circ$  NumPy
    - Manipulate and do computations for matrix.
    - Tutorial: Quick start for NumPy

- How to edit and run python programs?
  - Recommendation:
    - Google Colab <a href="https://colab.research.google.com">https://colab.research.google.com</a>
    - Advantage: easy to use, most of the packages are already installed.
  - For those who are familiar with Python:
    - Use <u>Anaconda/Miniconda</u> to manage your environments.
    - Use <u>Jupyter notebook</u>/ <u>PyCharm</u>/ <u>VSCode</u> to edit and run your code.
    - Advantage: manage the project on your own computer.

Requirements of Assignment 1

## Requirements of Assignment 1

- Where to download the assignment?
  - Course Moodle Assignment 1 (4 files with questions and some template code).



- What to submit?
  - Completed python notebook with executed outputs.
  - Name the file as using your uid, xxxx.zip. For example: 3009666.zip.
- Where to submit?
  - Course Moodle Assignment 1.
- When to submit?
  - o Before 16 Oct (11:59 PM).
- Do not copy! Both the student who copies and the student who offers his/her work for copying shall be penalized.

- Quiz (20 %) 2 equally weighted written quizzes
  - O Q1: 18 Oct, 12:30 pm 1:20 pm, written
  - o Q2: 29 Nov, 12:30 pm 1:20 pm, written
- Assignments (30 %) 3 equally weighted programming assignments
  - A1: 30 Sep 16 Oct (11:59 PM)
     A2: 28 Oct 13 Nov
  - o A3: 25 Nov 11 Dec

**COMP 3314** 

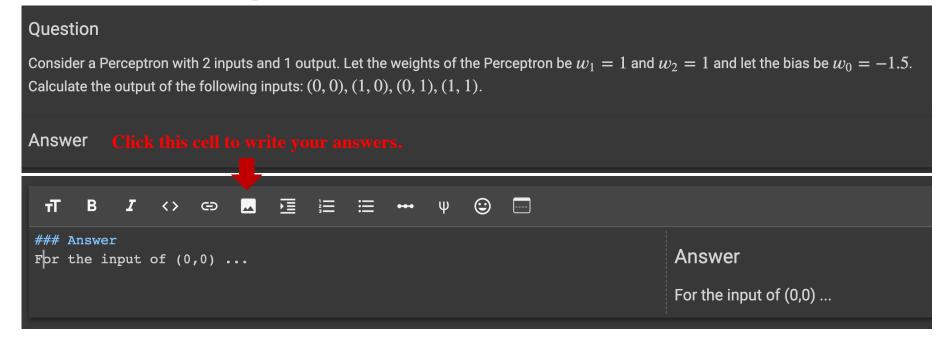
- As: 23 Nov 11 Dec Late submission policy:
  - 20% deduction within 24 hours, 50% deduction within 48 hours
  - no accept beyond 48 hours, unless extreme emergency
- Final examination (50 %)
- o Written, 120 minutes exam
- Candidates are permitted to refer to the following electronic/printed materials in the examination: textbook, lecture slides, assignment handout and sample solutions, and self-
- made notes
- Internet searching is NOT allowed

- Demos of using Google Colab for Assignment 1
  - Step1: download the files (4 python notebooks) from Moodle.
  - Step2: visit <a href="https://colab.research.google.com/">https://colab.research.google.com/</a>
  - Step3: choose "Upload" to upload hw1\_q1\_written.ipynb
  - Step4: answer the questions in this file.
  - Step5: execute all the code blocks to print the results.
  - Step6: save and download this finished .ipynb file
  - Step7: repeat Step 3-6 for all assignment files.
  - Step8: put the finished 4 files in one .zip, and name it using your uid, like 3009666.zip
  - Step9: submit the .zip on Moodle.

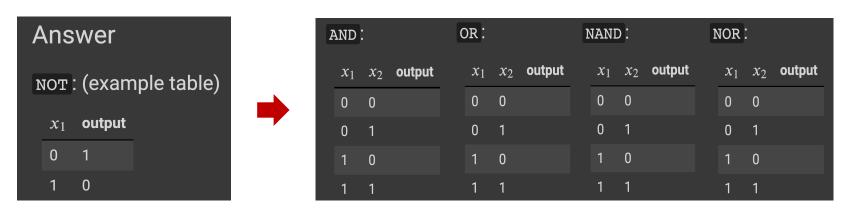
- Assignment 1: Overview
  - HW1-Q1: Written Questions (50 points)
  - HW1-Q2: Perceptron Boolean Operators (10 points)
  - HW1-Q3: Decision Boundary (20 points)
  - HW1-Q4: Hand-written Digits Classification (20 points)

- HW1-Q1: Written Questions(50 points)
  - Q1-1: Perceptron Basics (5 points)
  - Q1-2: Boolean Operators (5 points)
  - Q1-3: Parity Check (5 points)
  - Q1-4 Support Vectors (5 points)
  - Q1-5: Entropy (5 points)
  - Q1-6: Decision Tree (25 points)

- HW1-Q1-1: Perceptron Basics (5 points)
  - Write your answer in the given cell using using Markdown grammar and Latex math equations



- HW1-Q1-2: Boolean Operators (5 points)
  - Filling the tables with right output value for different Boolean operators.



- HW1-Q1-3: Parity Check (5 points)
  - $\circ$  Explanation: the input could only be 0/1. For example, if the 3 inputs are (0,1,1), the number of 1 equals to 2. As 2 is even, the output is expected to be 1.
  - O You are expected to calculate the weights for the perceptron that can do parity check, or you should prove why the perceptron can not.

#### Question

The parity problem returns 1 if the number of inputs that are 1 is even, and 0 otherwise.

Can a perceptron learn this problem for 3 inputs?

#### **Answer**

- HW1-Q1-4: Support Vectors (5 points)
  - First, run the given code(show in right) to draw the separating line.

plt.show()

• Then, answer the questions after observing the data points.

```
Question
```

Suppose that the following are a set of point in two classes:

- Class1: (1, 1), (1, 2), (2, 1)
- Class2: (0,0), (1,0), (0,1)
- Classz. (0, 0), (1, 0), (0, 1)
- 1. Plot them and find the optimal separating line. What are the support vectors?
- 2. What is the meaning of support vectors?

```
%matplotlib inline
import matplotlib.pyplot as plt
import numpy as np
# select setosa and versicolor
c1 = np.array([[1, 1], [1, 2], [2, 1]])
c2 = np.array([[0, 0], [1, 0], [0, 1]])
# plot data
plt.figure(figsize=(5, 5))
plt.scatter(c1[:, 0], c1[:, 1], color='red', marker='o', label='class1')
plt.scatter(c2[:, 0], c2[:, 1], color='blue', marker='x', label='class2')
x = np.linspace(0, 1., 5)
plt.plot(x, 1 - x, 'b--')
x = np.linspace(0, 2., 5)
plt.plot(x, 2 - x, 'r--')
x = np.linspace(0, 1.5, 5)
plt.plot(x, 1.5 - x, 'y-')
plt.xlim(-0.2, 2.5)
plt.ylim(-0.2, 2.5)
plt.grid(True)
plt.tight layout()
```

- HW1-Q1-5: Entropy (5 points)
  - Calculate the entropy according to formulations and explain the meaning.

```
Question
Suppose that the probability of five events:
   • P(\text{first}) = 0.5
    • P(\text{second}) = 0.125
    • P(\text{third}) = 0.125
    • P(\text{fourth}) = 0.125
    • P(\text{fifth}) = 0.125
Calculate the entropy and write down in words what this means.
Answer
```

- HW1-Q1-6: Decision Tree (25 points)
  - O Given several data samples, compute the Gini impurity and the information gain.
  - O Know how to select features to make decision.

#### Question

The new energy vehicle (NEV) is the growing trend in the automotive industry to replace traditional gas-powered vehicles.

Consider a dataset of customer preference for vehicles. Here are the possible values for each feature:

• Engine: {Gas, NEV}

Style: {Sedan, SUV}

• Price: {Regular, Luxury}

Note that samples with the same features can have different labels. If the leaf node of a decision tree is not pure, the majority vote is used to determine the output label.

Now, you want to build a decision tree to predict the preference of a customer. In particular, you want to know which feature (among Engine, Style, and Price) is the most important feature to predict the preference. In other words, which feature should be the root node of the decision tree to maximize the information gain?

#### Your tasks:

- 1. Compute Gini impurity at the root node. (5 points)
- 2. For the 3 features (Engine, Style, and Price), compute the information gain if that feature is used split the root node. (15 points)
- 3. Conclude which feature is the most important feature to predict the preference as it maximizes the information gain. (5 points)

Here is the dataset:				
Engine	Style	Price	Preference?	
Gas	Sedan	Regular	Yes	
Gas	Sedan	Regular	No	
Gas	Sedan	Luxury	No	
Gas	Sedan	Luxury	No	
Gas	SUV	Regular	Yes	
Gas	SUV	Luxury	Yes	
Gas	SUV	Luxury	No	
NEV	Sedan	Regular	Yes	
NEV	Sedan	Regular	Yes	
NEV	Sedan	Luxury	Yes	
NEV	Sedan	Luxury	Yes	
NEV	SUV	Regular	No	
NEV	SUV	Regular	Yes	
NEV	SUV	Regular	Yes	
NEV	SUV	Luxury	Yes	
NEV	SUV	Luxury	No	

- HW1-Q2: Perceptron Boolean Operators (10 points)
  - O The demo code for "NOT Operator" is provided, do not change it
  - Implement other operators referring to the demo code.

```
1. NOT Operator
[1] class PerceptronNOT:
        def init (self):
            self.w0 = 0.5
            self.w1 = -1
        def decision function(self, z):
            return 1 if z >= 0 else 0
        def forward(self, x1):
            z = self.w0 + self.w1 * x1
            phi z = self.decision function(z)
            return phi z
    model = PerceptronNOT()
    for x1 in [0, 1]:
        print(f"NOT({x1}) = {model.forward(x1)}")
    NOT(0) = 1
    NOT(1) = 0
```

```
2. AND Operator
[ ] # Your code here:
3. or Operator
[ ] # Your code here:
4. NAND Operator
[ ] # Your code here:
5. NOR Operator
[ ] # Your code here:
```

- HW1-Q3: Decision Boundary (20 points)
  - A lot of supporting codes are provided, run the cells one by one. Do not change them!

```
[ ] import matplotlib.pyplot as plt
    from matplotlib.colors import ListedColormap
    import numpy as np
    import pandas as pd
    Note: Do not change the code in this cell.
    class Perceptron(object):
        def __init__(self, eta=0.01, n_iter=10):
                eta (float, optional): Learning rate. Defaults to 0.01.
                n iter (int, optional): Number of iterations. Defaults to 10.
            self.eta = eta
            self.n iter = n iter
        def fit(self, xs, ys):
            Fit training data.
                xs (array-like): Training vectors, shape = (n samples, n features).
                ys (array-like): Target values, shape = (n samples,).
```

```
Note: Do not change the code in this cell.
def fetch dataset():
    Download and get a subset the UCI Iris dataset.
    Returns:
        (xs, ys), where xs has shape (100, 2) and ys has shape (100,).
    # Download dataset
    url = "https://archive.ics.uci.edu/ml/machine-learning-databases/iris/iris.data"
    df = pd.read csv(url, header=None)
    df.tail()
    # Select setosa and versicolor
    num samples = 100
    vs = df.iloc[0:num samples, 4].values
    ys = np.where(ys == "Iris-setosa", -1, 1)
    # Extract sepal length and petal length
    xs = df.iloc[0:num samples, [0, 2]].values
    return xs, ys
```

- HW1-Q3: Decision Boundary (20 points)
  - The supporting codes would guide you to:
    - Train a perceptron on given dataset.
    - Visualize the dataset and decision regions.
    - Draw a random decision boundary
  - What you need to do:
    - Write code in the last cell to compute the actual decision boundary for the trained perceptron.

- HW1-Q4: Hand-written Digits Classification (20 points)
  - We provide the template code for perceptron.
  - O You are required to use several other classifiers like LR, SVM, KNN for digits classification, and analyze the results.

```
Classifier #1 Perceptron
     # Example code, including training and testing, to observe the accuracies.
     from sklearn.linear model import Perceptron
     # Tune the eta0 hyperparameter.
     eat0 list = [0.0001, 0.001, 0.01, 0.1, 1]
     # Your code here.
     accuracies = []
    for eta0 in eat0 list:
         model = Perceptron(max iter=100, tol=1e-3, eta0=eta0)
         model.fit(xs train, ys train)
         ys pred = model.predict(xs test)
         accuracy = get accuracy(ys test, ys pred)
         accuracies.append(accuracy)
     for eta0, accuracy in zip(eat0 list, accuracies):
         print(f"eta0 = {eta0:.4f}, accuracy = {accuracy:.4f}")

        ← eta0 = 0.0001, accuracy = 0.9500

     eta0 = 0.0010, accuracy = 0.9167
     eta0 = 0.0100, accuracy = 0.9389
    eta0 = 0.1000, accuracy = 0.9333
    eta0 = 1.0000, accuracy = 0.9333
```

```
▼ Classifier #2 Logistic Regression

[ ] # Your code, including training and testing, to observe the accuracies.

from sklearn.linear_model import LogisticRegression

# Tune the C hyperparameter.

C_list = [1e-3, 0.001, 0.01, 1, 10, 100]

# Your code here.
```

```
Classifier #6 KNN

[ ] # Your code, including training and testing, to observe the accuracies.
    from sklearn.neighbors import KNeighborsClassifier

# Tune the n_neighbors hyperparameter.
    n_neighbors_list = [2, 3, 5, 10, 20]

# Your code here.
```

#### Further Questions

- Mr. Xi Chen (email: xchen2@cs.hku.hk)
- Office: HW-RSC or zoom (<a href="https://hku.zoom.us/my/xavier.xichen">https://hku.zoom.us/my/xavier.xichen</a>)
- Consultation hours: Wednesday, 10:00 am 12:00 pm

- Mr. Yixing Lao (email: laoyx@connect.hku.hk)
- Office: HW-RSC or zoom (<a href="https://hku.zoom.us/my/laoyixing">https://hku.zoom.us/my/laoyixing</a>)
- Consultation hours: Tuesday, 2:00 pm 4:00 pm
- Please send an email before our meeting

Q & A