# COMP 532 Machine Learning and Bioinspired Optimisation

Lecture 3



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## Overview

- Mathematical Preliminaries
- Python Basics

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- Mathematical Preliminaries
- Python Basics

## **Probability basics**

- A random variable X represents outcomes or states of the world
- We will write p(x) to mean Probability(X = x)
- Sample space: The space of all possible outcomes (may be discrete, continuous, or mixed)
- p(x) is the probability mass (density) function
  - Assigns a number to each point in sample space
  - Non-negative, sums (integrates) to 1
  - Intuitively: how often does x occur, how much do we believe in x

## Probability basics

- Random Experiment Experiment that results in different outcomes despite being in similar conditions
  - Tossing of a coin, throwing of a dice, rainfall amount

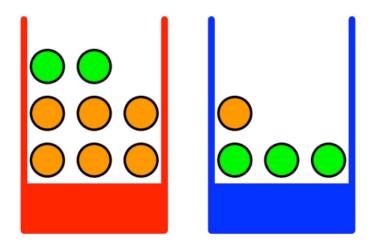
- Sample Space: Set of all possible outcomes of a random experiment
  - Tossing of a coin once: S = {H, T} (head and tail)
  - Tossing coin twice: S = {HH, HT, TH, TT}

# Probability basics

- Joint Probability
- Sum Rule
- Conditional Probability
- Product Rule
- Bayes' Rule

# A Simple Example

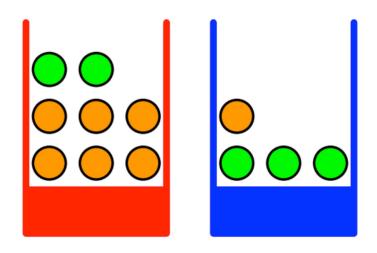
- Two Baskets Red & Blue
- Each basket has some fruits oranges and apples (green)
- All fruits in a basket are equally available
- Red Basket 6 oranges, 2 apples
- Blue Basket 1 orange, 3 apples
- Let P(B=r) = 0.4 and P(B=b) = 0.6
- Random variables:
  - B: {b, r}; F:{o,a}



- What is the probability of picking an orange?
- What is the probability that I pick red basket given that the fruit I pick was an orange?

# A Simple Example

- Assume N = 100 trials
- P(B=r) = 0.4 and P(B=b) = 0.6



	F = 0	F = a	
B = r	6/8 x 40 = 30	2/8 x 40 = 10	40
B = b	1/4 x 60 = 15	3/4 x 60 = 45	60
	45	55	

# Joint Probability

- Assume N = 100 trials
- P(B=r) = 0.4 and P(B=b) = 0.6

	F = 0	F = a	
B = r	30	10	40
B = b	15	45	60
	45	55	

• Joint Probability: The probability (that X will take a value  $x_i$  and Y will take a

value 
$$y_i$$
)  $P(X = x_i, Y = y_i)$   
 $P(B=r, F=o) = 30/100 = 0.3$ 

 $y_j$ 

	$x_i$	
	$n_{ij}$	
	<i>C</i> ;	

#### Generalize

• Let the number of trials that  $X = x_i$  and  $Y = y_i$  be  $n_{ij}$ .

Then,  $P(X = x_i, Y = y_i) = \frac{n_{ij}}{N}$ , where N is the total number of trials

• P(B=b, F=a) = ?

## Sum Rule

- Assume N = 100 trials
- P(B=r) = 0.4 and P(B=b) = 0.6

	F = 0	F = a	
B = r	30	10	40
B = b	15	45	60
	45	55	

What is the probability that I pick oranges?

$$\circ$$
 P(F=o) = 45/100 = 0.45

 $y_j$ 

$x_i$				
		$n_{ij}$		

#### Generalize

- Then,  $P(X = x_i) = \frac{c_i}{N}$  (marginal probability)
- However,  $c_i = \sum_j n_{ij}$
- Then,  $P(X = x_i) = \sum_j \frac{n_{ij}}{N}$

## Sum Rule

- Assume N = 100 trials
- P(B=r) = 0.4 and P(B=b) = 0.6

	F = 0	F = a	
B = r	30	10	40
B = b	15	45	60
	45	55	

What is the probability that I pick oranges?

$$\circ$$
 P(F=o) = 45/100 = 0.45

$x_i$				
		$n_{ij}$		

#### Generalize

- Then,  $P(X = x_i) = \frac{c_i}{N}$  (marginal probability)
- However,  $c_i = \sum_j n_{ij}$
- Then,  $P(X = x_i) = \sum_j \frac{n_{ij}}{N}$
- We have  $P(X = x_i, Y = y_i) = \frac{n_{ij}}{N}$
- Then  $P(X = x_i) = \sum_j P(X = x_i, Y = y_i)$  Sum rule of Probability

# **Conditional Probability**

- Assume N = 100 trials
- P(B=r) = 0.4 and P(B=b) = 0.6

	F = 0	F = a	
B = r	30	10	40
B = b	15	45	60
	45	55	

• What is the probability that the basket I picked is red given that the fruit was orange: P(B = r|f=o) = 30/45  $x_i$ 

 $y_j$   $n_{ij}$   $c_i$ 

- What is the probability that Y will take the value  $y_j$  given that X will take the value  $x_i$
- Then  $P(Y = y_i | X = x_i) = \frac{n_{ij}}{c_i}$

## **Product Rule**

- Assume N = 100 trials
- P(B=r) = 0.4 and P(B=b) = 0.6

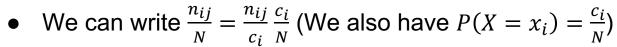
	F = 0	F = a	
B = r	30	10	40
B = b	15	45	60
	45	55	

**Conditional Probability** 

 $P(Y = y_i | X = x_i) = \frac{n_{ij}}{c}$ 

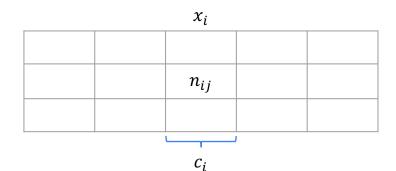
Joint Probability

$$P(X = x_i, Y = y_i) = \frac{n_{ij}}{N}$$



 $y_i$ 

Thus  $P(X = x_i, Y = y_i) = P(Y = y_i | X = x_i) P(X = x_i)$ 



## Rules Simplified

- Sum rule  $P(X = x_i) = \sum_i P(X = x_i, Y = y_i)$
- Produce rule  $P(X = x_i, Y = y_i) = P(Y = y_i | X = x_i)P(X = x_i)$

#### Simply,

- Sum rule  $P(X) = \sum_{y} P(X, Y)$
- Produce rule P(X,Y) = P(Y|X)P(X)

# Bayes' Rule

- Produce Rule P(A,B)=P(A|B)P(B)
- Similarly, P(B,A)=P(B|A)P(A)
- Since P(A,B)= P(A|B)P(B)
- Then we have P(A|B)P(B) = P(B|A)P(A)

$$P(A|B) = P(A) \frac{P(B|A)}{P(B)}$$
 Likelihood  
Posterior Prior

## **Python Basics**

- Python is an interpreted, high-level and general- purpose programming language.
  - Created by Guido van Rossum and first released in 1991
- Has a lot of : <a href="https://www.python.org">https://www.python.org</a>
  - Image Processing
  - Computer Vision
  - Machine Learning
  - Deep Learning
  - Optimization
  - Signal Processing
  - ..

## Why Python

- Open source
- Ease of coding, "Code as plain English" is Python's primary goal.
- Most commonly used, a bigger community. There are a lot of blog posts and online resources regarding Python + OpenCV
- Vast libraries for machine learning, computer vision
- ...

Python vs. C++ vs. Java vs. Matlab

## Tools to use Python

#### Anaconda

Anaconda refers to an open source Python distribution, which includes more than 180 scientific packages and their dependencies, e.g., conda, Python, Jupyter Notebook.

#### VSCode/Pycharm

 An Integrated Development Environment (IDE) for Python. It has a set of tools that can help users improve their efficiency when developing in Python language.
 Such as debugging, syntax highlighting, project management, code jump, smart prompt, auto completion, etc.

#### Jupyter Notebook

- Jupyter Notebook is a web-based application for interactive computing. It can be applied to the whole process of development: code writing, running the code and displaying the results.
- Install the package directly in Jupyter Notebook, such as: pip install pytorch

## Tools to use Python

#### Google Colab

- Colaboratory is a free Jupyter Notebook environment that requires no setup and runs on the Cloud (write, run and share code)
- No configuration required
- Free use of GPU, TPU
- Easy sharing
   Install the package directly in Jupyter Notebook, such as: pip install tensorfolw
- Online platform: <a href="https://colab.research.google.com/notebooks/intro.ipynb">https://colab.research.google.com/notebooks/intro.ipynb</a>
  - Similar to Jupyter Notebook

#### Data structures in Python

- Basic
  - o int
  - o float
  - o bool

#### Ordered

- o list: [2, 3, 3]
- o array: np.array([2, 3, 3]) # using numpy
- o tuple: (1,3,6,10)
- string: 'hello'

#### Unordered

- set: set('abracadabra')
- o dict: {'jack': 4098, 'sape': 4139}

#### Useful tricks

- Array
  - Vector is stored as 1D array [1 2 3 4 5 6]
  - Matrix is 2D array [[1 2 3]
     [4 5 6]]
- The index is from 0
  - For example, let A = [1,2,3,4,5,6] then A[0] = 1
- Array in Numpy import numpy as np

$$a1 = np.array([3, 4, 5, 3])$$

Matrix in Numpy

import numpy as np

m1 = np.random.rand(2,2)

#### Function in Python

Creating a function: In Python a function is defined using the def keyword.

```
def my_function():
    print("Hello from a function")
```

Calling a function

```
def my_function():
    print("Hello from a function")
my_function()
```

Arguments

```
def my_function(fname):
    print(" My name is " + fname)
my_function("Shan")
```

#### Syntax in Python

• if...elif...else

```
a = 200
b = 33
if b > a:
   print("b is greater than a")
elif a == b:
   print("a and b are equal")
else:
   print("a is greater than b")
```

• for

```
fruits = ["apple", "banana", "cherry"]
for x in fruits:
   print(x)
```

while

```
i=1
while i < 6:
    print(i) i += 1</pre>
```

Debugging in Python

• Debugging with pdb: easy, powerful, no extra IDE needed

https://www.youtube.com/watch?v=bHx8A8tbj2c

Debugging with PyCharm: easy, but not free

# More you can explore

Check out more in w3schools, and practice!

https://www.w3schools.com/python/

# Summary

- Probability basics
- Introduction to Python