# Introduction – Part 1

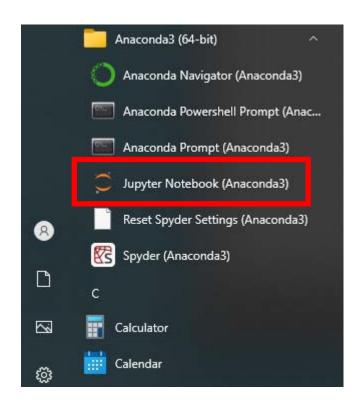
# How to install and use Jupyter Notebook

Go to <a href="https://www.anaconda.com/download">https://www.anaconda.com/download</a>

Download Anaconda3

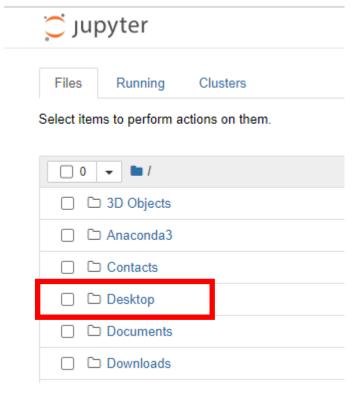
**Install Anaconda3** 

**Click Jupyter Notebook** 

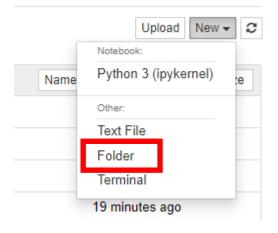




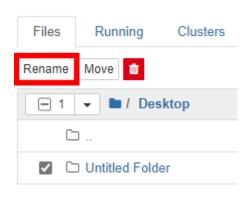
# **Click Desktop**

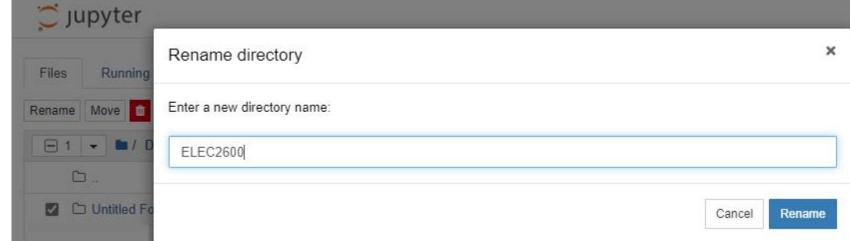


### **Create a Folder**

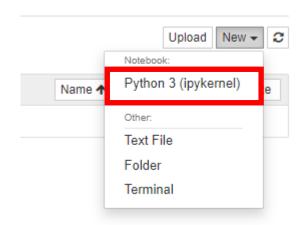


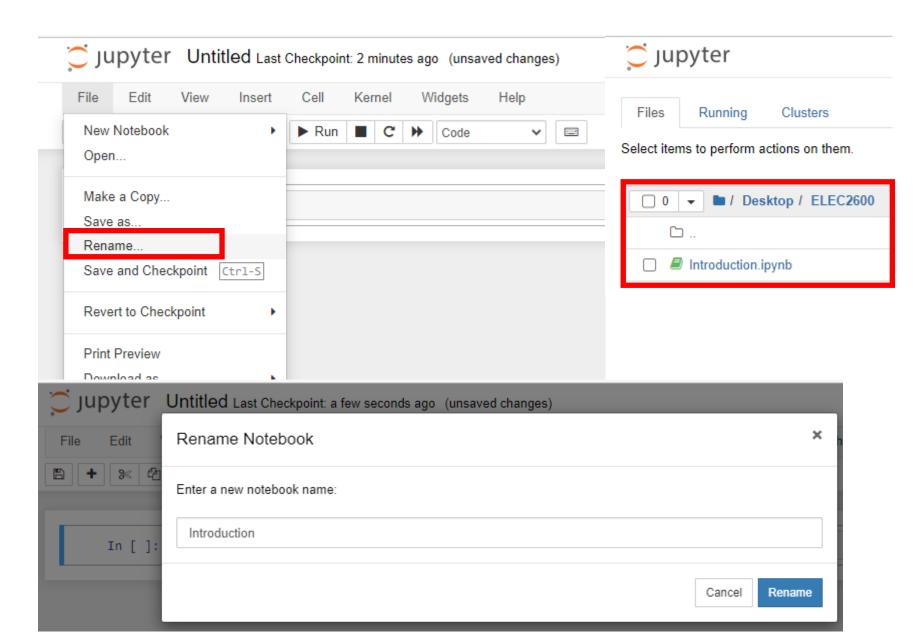
### Rename the Folder



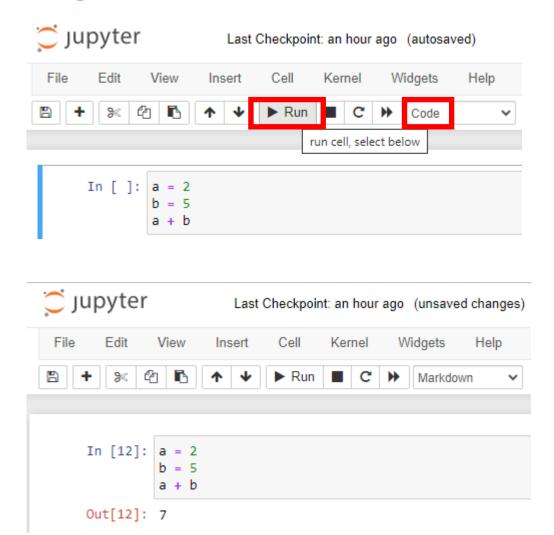


### Create a new notebook Rename the notebook

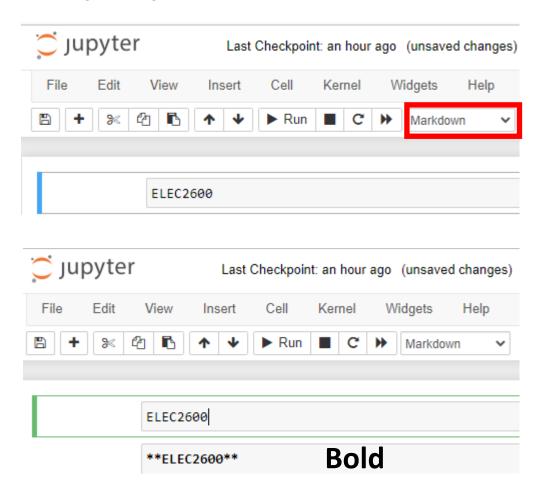




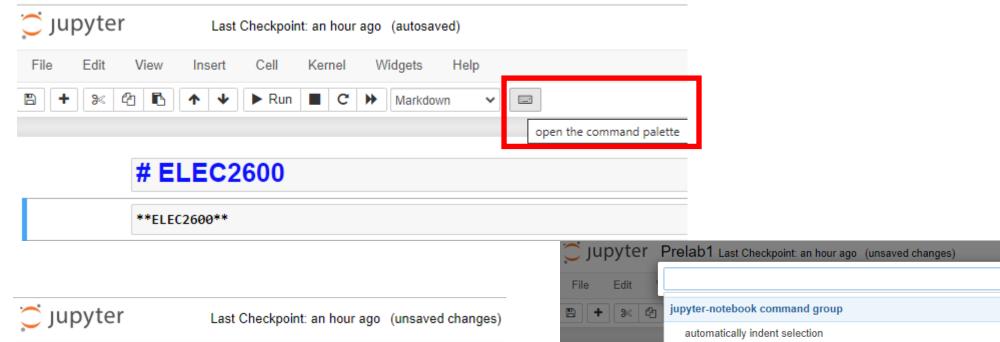
# **Program code and Run cell**



# **Text (Bold)**



# **Text (Heading)**



ELEC2600

Insert

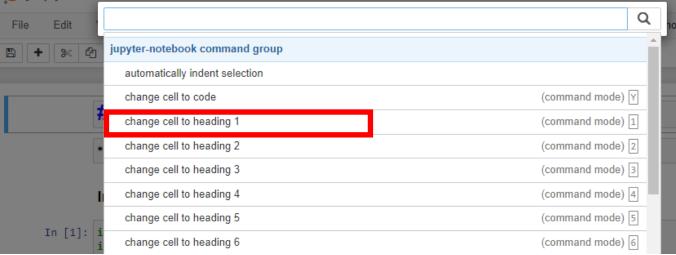
Cell

Kernel

run cell, select below

Widgets

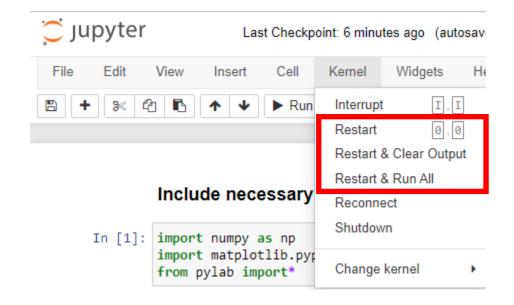
C → Markdown

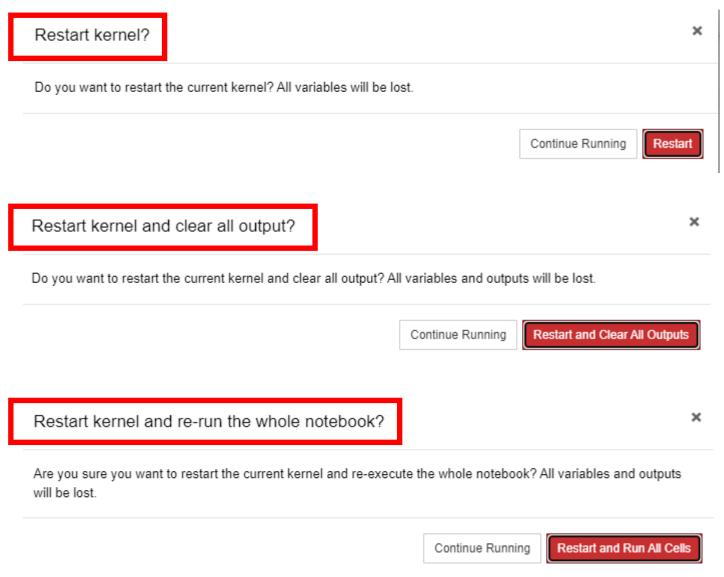


**ELEC2600** 

View

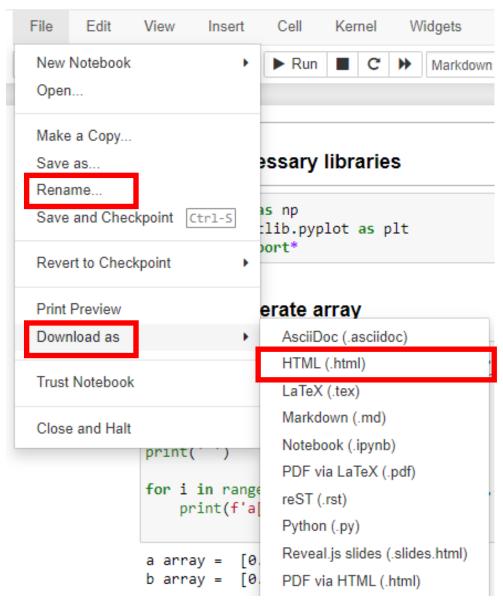
### Kernel



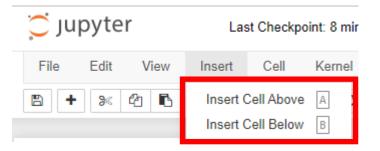


### File

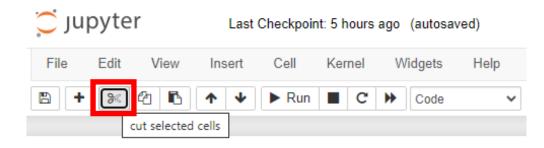




#### Insert



### Delete a cell



### Ex. 06

a = np.array([1, 4, 5])

b = np.array([1, 4, 5, 5, 4, 1, 1, 1])

```
for i in range(a.size):
   print(f'a\{[i]\} = ', a[i])
   compare = np.equal(b, a[i])
   print(compare)
   count = np.cumsum(compare)
   print(count)
   print(' ')
a[0] = 1
[ True False False False True True]
[1 1 1 1 1 2 3 4]
a[1] = 4
[False True False False True False False]
[0 1 1 1 2 2 2 2]
a[2] = 5
[False False True True False False False]
[0 0 1 2 2 2 2 2]
```

np.equal

np.cumsum

How to obtain the relative frequency versus *n*?

How to plot the relative frequency versus *n*?

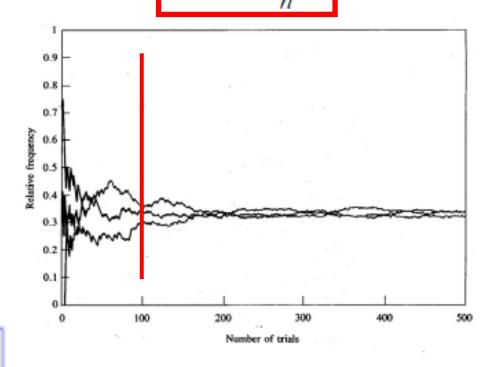
# **Relative frequency**

- Let  $N_1(n)$ ,  $N_2(n)$  and  $N_3(n)$  be the *number of times* that we pick balls 1, 2, and 3 in n trials (*events*).
- □ Define the *relative frequency* of the outcome k as  $f_k(n)$
- This experiment exhibits statistical regularity: as n increases, the relative frequency approaches a constant value

$$\lim_{n\to\infty} f_k(n) = p_k$$

where  $p_k$  indicates the probability of outcome k.

Provides a key connection between measurement of physical quantities and probability models!

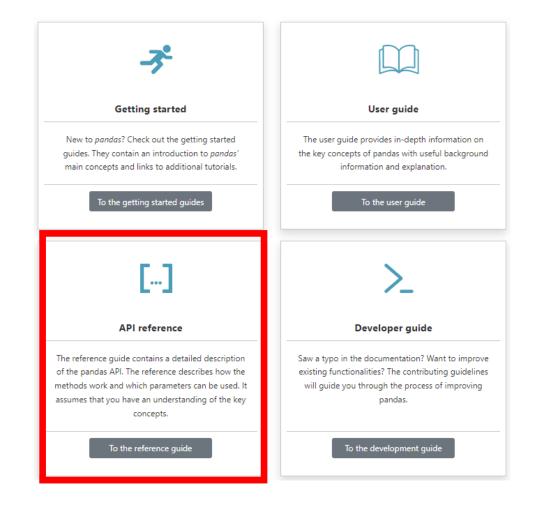


# Introduction – Part 2

# How to handle excel file

Go to <a href="https://pandas.pydata.org/pandas-docs/stable/index.html">https://pandas.pydata.org/pandas-docs/stable/index.html</a>

Click API reference



### Click DateFrame



Getting started User Guide

API reference

Development Release notes

Input/output

General functions

Series

#### DataFrame

pandas.DataFrame

pandas.DataFrame.index

pandas.DataFrame.columns

pandas.DataFrame.dtypes

pandas.DataFrame.info

# pandas.DataFrame.groupby

DataFrame.groupby(by=None, axis=0, level=None, as\_index=True, sort=True, group\_keys=True, observed=False, dropna=True) # [source]

Group DataFrame using a mapper or by a Series of columns.

A groupby operation involves some combination of splitting the object, applying a function, and combining the results. This can be used to group large amounts of data and compute operations on these groups.

# Click and read the description of the followings (groupby, count and sum)

# Function application, GroupBy & window

DataFrame.apply (func[, axis, raw,])	Apply a function along an axis of the DataFrame.
<pre>DataFrame.applymap (func[, na_action])</pre>	Apply a function to a Dataframe elementwise.
DataFrame.pipe (func, *args, **kwargs)	Apply chainable functions that expect Series or DataFrames.
DataFrame.agg ([func, axis])	Aggregate using one or more operations over the specified axis.
DataFrame.aggregate ([func, axis])	Aggregate using one or more operations over the specified axis.
DataFrame.transform (func[, axis])	Call func on self producing a DataFrame with the same axis shape as self.
DataFrame.groupby ([by, axis, level,])	Group DataFrame using a mapper or by a Series of columns.
DataFrame.rolling (window[, min_periods,])	Provide rolling window calculations.
DataFrame.expanding ([min_periods, axis, method])	Provide expanding window calculations.
DataFrame.ewm ([com, span, halflife, alpha,])	Provide exponentially weighted (EW) calculations.

### Computations / descriptive stats

DataFrame.abs ()	Return a Series/DataFrame with absolute numeric value of each element.
DataFrame.all ([axis, bool_only, skipna])	Return whether all elements are True, potentially over an axis.
DataFrame.any (*[, axis, bool_only, skipna])	Return whether any element is True, potentially over an axis.
DataFrame.clip ([lower, upper, axis, inplace])	Trim values at input threshold(s).
DataFrame.corr ([method, min_periods,])	Compute pairwise correlation of columns, excluding NA/null values.
DataFrame.corrwith (other[, axis, drop,])	Compute pairwise correlation.
DataFrame.count ([axis, numeric_only])	Count non-NA cells for each column or row.
	Count non-IVA cells for each column or row.
DataFrame.sem ([axis, skipna, ddof, numeric_only])	Return unbiased standard error of the mean over requested axis.
DataFrame.sem ([axis, skipna, ddof, numeric_only])  DataFrame.skew ([axis, skipna, numeric_only])	Return unbiased standard error of the mean
	Return unbiased standard error of the mean over requested axis.

### Ex.01 – Groupby and Count

#### print(data) Class Price Index fruit fruit fruit 10 fruit 10 fruit vegetable vegetable vegetable vegetable vegetable 10 fruit 12 fruit 13 vegetable 14 vegetable vegetable 15 vegetable 16

### Ex.01 - Groupby and Count

```
n_class = data.groupby('Class')['Price'].count()
print(n_class)
print('')

print('Number of fruits = ', n_class['fruit'])
print('')

print('Total number of products', sum(n_class))
```

```
Class
fruit 7
vegetable 9
Name: Price, dtype: int64
```

# **Relative frequency?**

```
Number of fruits = 7
```

Total number of products 16

# Ex.02 – Groupby and Sum

```
print(data)
print(' ')

n_class = data.groupby('Class')['Price'].sum()
print(n_class)
```

Total price of fruit = ?

Total price of vegetable = ?

	Class	Price	
Index			
1	fruit	5	
2	fruit	5	
3	fruit	10	
4	fruit	10	
5	fruit	3	
6	vegetable	1	
7	vegetable	1	
8	vegetable	5	
9	vegetable	2	
10	vegetable	4	
11	fruit	5	
12	fruit	5	
13	vegetable	1	
14	vegetable	5	
15	vegetable	2	
16	vegetable	4	

```
Class
fruit 43
vegetable 25
Name: Price, dtype: int64
```

### Ex.03 – Create a new column

```
data['data_price'] = data['Price'] > 4
print(data)
print(' ')

print('The sum of data_price = ', sum(data['data_price']))
print(' ')

n_fruit_4 = data.groupby('Class')['data_price'].sum()
print(n_fruit_4)
```

### **True = 1 False = 0**

fruit (Price > 4) + vegetable (Price > 4) = 8

fruit (Price > 4) = ?

vegetable (Price > 4) = ?

Class	Price	data_price
fruit	5	True
fruit	5	True
fruit	10	True
fruit	10	True
fruit	3	False
vegetable	1	False
vegetable	1	False
vegetable	5	True
vegetable	2	False
vegetable	4	False
- fruit	5	True
fruit	5	True
vegetable	1	False
vegetable	5	True
vegetable	2	False
vegetable	4	False
	fruit fruit fruit fruit fruit vegetable vegetable vegetable vegetable vegetable vegetable vegetable vegetable vegetable	fruit 5 fruit 10 fruit 10 fruit 3 vegetable 1 vegetable 5 vegetable 2 vegetable 4 fruit 5 fruit 5 fruit 5 vegetable 1 vegetable 2 vegetable 2

The sum of data\_price = 8

```
Class
fruit 6
vegetable 2
Name: data_price, dtype: int64
```

**Event A** 

**Event B** 

Prior probability of event A

Prior probability of event B

P(A)

P(B)

**Event A and Event B both occur** 

 $P(A \cap B)$ 

**Conditional probability** 

$$P(A|B) = \frac{P(A \cap B)}{P(B)}$$

What is the prior probability of fruit?

What is the probability  $P[(Price > 4) \cap Fruit]$ ?

What is the conditional probability P[(Price > 4) | Fruit]?