



## Qbus1040 Python notes

Foundations of Business Analytics (University of Sydney)



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# Python Basic

Print:

```
Print('qbus1040')
```

List:

```
numbers = [4, 7, 2, 0]
```

```
to_buy = ['carrot', 'fruit']
```

Float → decimal number

String formation:

```
print('Pi to 2 decimal places: {:.2f}'.format(pi))
```

Calculator/Math library:

+ addition, - subtraction, \* multiplication, / division, % modulus, \*\* power

Import math

```
Print(math.factorial(4)) → 24
```

Time library:

```
import time
```

```
start = time.time()
```

```
# Your program
```

```
end = time.time()
```

```
elapsed_time = end - start
```

Numpy library:

**Input: collect information from user**

```
variable = input('message_displayed_to_user ')
```

- Don't forget that input values are always strings!
- You might need to convert to *int()* , *float()* , *np.fromstring(variable, dtype=int, sep=' ')* when handling numeric values.

```
variable = input(message_displayed_to_user)
```

```
e.g. x = int(input('Enter your number: '))
```

- *np.fromstring* → convert input to vector

*np.random.seed(0)* → every time generate same set of random #

*np.random.rand(int/float/nth)*(low (inclusive), high (exclusive), size)

*np.inner* → inner product between 2 vectors ( $a^T b$ ) → scalar

*np.array* → store vectors and matrices

*np.shape/ a.shape()* → show dimensions

*np.concatenate((a,b, axis=0 or 1))*

```
np.array([2, 1 ,3])
```

```
array[index]
```

```
array[start:end]
```

- start index inclusive, end exclusive
- *A[2:4, :3]* → starting from 2 to 3, starting from 0 up to 2

**range(end) - end is exclusive**

**range(start, end)**

**range(start, end, step\_size)**

**np.zeros(dimension)**

- *Np.zeros((n,1))* → print zero matrix
- *Np.zeros(n)* → print n-vector

**np.ones(dimension)**

*.min()*, *.max()*, *.mean()*, *.sum()*

*.argmin()*, *.argmax()* → show the index of the min/max values

*np.isclose()* → check whether two values are similar

**if condition:**

```
# code you execute if condition is true
```

**else:**

```
# code you execute if condition is false
```

**Elif statement**

```
if condition_1:
    # code to execute if condition 1 is true
elif condition_2:
    # code to execute if condition 2 is true
elif condition_3:
    # code to execute if condition 3 is true
...
else:
    # code to execute if none of the conditions are true
```

```
np.eye(n)
np.tril (matrix)
np.triu (matrix)
np.diag(np.array([3.7, 2.5, -1.2, 4.5]))
```

**Brackets:**

- () Used for functions such as print()
- [] Used for lists.
- {} Used for string formatting.

**Root mean Square(RMS)**

```
def rms(x):
    norm = np.sqrt(np.inner(x,x))
    sqrt_n = np.sqrt(x.shape[0])
    rms = norm/sqrt_n

    return rms
```

Tut 7 Gram-schmidt  $A \rightarrow Q$   
 Tut 8 QR factorisation  $A \rightarrow Q, R$   
 Tut 9 Back Sub  $Rx = b \rightarrow$  get x  
 Tut 10 Solve via Back Sub  $A, b \rightarrow$  get x

### Back substitution: $Rx = b$

- Get x from R, b

```
def back_sub(R, b):
    n, n = R.shape
    x = np.zeros(n)
    for i in range(n):
        j = n - 1 - i
        x[j] = (b[j] - np.inner(R[j], x)) / R[j, j]
    return x
```

Solve via Back Substitution:

- Get  $\hat{x}$  from  $R, Q^T b$
- $A \rightarrow Q, R$
- $B \rightarrow Q^T b$
- Perform back sub ( $R, Q^T b$ )

$$\begin{aligned}\hat{x} &= A^+ b = R^{-1} Q^T b \\ R \hat{x} &= R R^{-1} Q^T b \\ R \hat{x} &= Q^T b\end{aligned}$$

```
def solve_via_back_sub(A, b):
    linearly_independent, Q_transpose = gram_schmidt(A.T) → why on rows?
    R = Q_transpose @ A
    Q = Q_transpose.T
    x = back_sub(R, Q_transpose @ b)
    return x
```

```
def polyfit(x, y, degree):
    A = vandermonde(x, degree+1)
    theta_hat = solve_via_back_sub(A, y)
    return theta_hat
```

```
def polyeval(x, theta_hat, degree):
    A = vandermonde(x, degree+1)
    f_hat_x = A @ theta_hat
    return f_hat_x
```

## Least Squares Data Fitting

- Constant model:  
 $A = \text{np.ones}((y\_d.\text{shape}[0], 1))$
- Straight-line fit:  
 $A = \text{np.concatenate}((\text{ones}, x\_d), \text{axis}=1)$
- De-trended time series:

```
year = [1980, 1985, 1990, 1995, 2000, 2005, 2010, 2015]
location = [0, 5, 10, 15, 20, 25, 30, 35]
plt.xticks(location, year)
```

- Polynomial fit:  
`A = func.vandermonde(x, degree+1)`  
`theta_hat = func.solve_via_back_sub(A, y_d)`
  - Sometimes there are errors with python, when some value in A are too large, the `p_inv` of A will be too small, python can't handle so we use solve via back sub, instead of `np.linalg.pinv(A) @ y_d`

- Auto Regressive model:  
`A = np.zeros((T-M, M))` → no. of predictions corresponds to # of rows in zeros

```
for i in range(M):
    A[:, i] = z[M-1-i:T-1-i]
    # A[:, 0] = z[M-1-0:T-1-0]
    # A[:, 1] = z[M-1-1:T-1-1]
    # A[:, 2] = z[M-1-2:T-1-2]
y_d = z[M:].reshape(-1, 1) → true data (after memory)
```

-----

`np.linspace(start, stop, number)` → generate a vector of equally spaced x values for testing

- The easiest way to visualise your model is to "plot points".
- This involves constructing a new matrix A for lots of different x values, and then obtaining the corresponding  $y^{\wedge}$  and then joining the dots!

`theta_hat = np.linalg.pinv(A) @ y_d` → find model parameter  
`f_hat_x = A @ theta_hat` → estimate/predicted  $y_d$

## Loading Data

`Fold1 = np.load('fold1.npy')`  
`red_df = pd.read_csv('red-wine.csv').to_numpy()` → convert to np array

## Extracting Data

- `x.reshape(-1, 1)`
  - This function reshape a 1D array into 2D array with one column, and as many rows as needed to contain all elements
  - Convert a row vector (array) to column vector
- `x_data = data[:, 0]` #extracting all rows but only 1st col  
`y_data = data[:, 1]` #extracting all rows but only 2nd col
- `allfolds = [0, 1, 2, 3, 4]`  
`selected = allfolds[1:]` #starting from 1 (inclusive till the end)  
`[1, 2, 3, 4]`
- `f = ['fold0', 'fold1', 'fold2', 'fold3', 'fold4']`  
`i = 2`  
`print(f[i:] + f[i+1:])`  
#selecting all folds, except fold i  
`['fold0', 'fold1', 'fold3', 'fold4']`

## Cross Validation

- Split data into training and test data
- Obtain theta hat from train data (x,y)
- Use theta hat obtained from train data, generate  $f_{\text{hat}_x}$  for test data (`A_test @ theta_hat`)
- Compare predicted  $f_{\text{hat}_y}$  to true test  $y$  → generate residuals, rmse

#calculate the rmse of the test/train data of that split  
`def rmse_split(all_folds, split, degree):`

```
train = np.concatenate(all_folds[:split] + all_folds[split+1:], axis=0)
test = all_folds[split]
#pull out all training except for fold split, and make split the test d
```

```
train_x = train[:, 0]; train_y = train[:, 1]
test_x = test[:, 0]; test_y = test[:, 1]
```

```

theta_hat = funcs.polyfit(train_x, train_y, degree)

test_y_hat = funcs.polyeval(test_x, theta_hat, degree)
test_residuals = test_y - test_y_hat
test_rmse = funcs.rms(test_residuals)

train_y_hat = funcs.polyeval(train_x, theta_hat, degree)
train_residuals = train_y - train_y_hat
train_rmse = funcs.rms(train_residuals)

return test_rmse, train_rmse

#Test function
all_folds = [fold0, fold1, fold2, fold3, fold4]
split = 0
degree = 1
test = rmse_split(all_folds, split, degree)

#creating for-loop calculate all rmses of all splits
rmses = np.zeros(5)
for i in range(5):
    split = i
    test = rmse_split(all_folds, split, degree)
    rmses[i] = test

rms_cv = funcs.rms(rmses)

```

## Visualizing Data (Matplotlib)

Import matplotlib.pyplot as plt

```

plt.figure(figsize=(9, 4)) → create new figure with width9, height4
plt.subplot(1, 2, 1)
plt.subplot(1, 2, 2)

```

```

plt.scatter(x values, y values, s=10) → scatter plots
plt.plot(x values, y values, 'o-') → line plots

```

- label = ' $y^{\{d_1\}}$ ' →  $y^{d_1}$

```

plt.axhline(1, color='red') → adding straight line

```

```

plt.quiver(0, 0, 3, 4, angles='xy', scale_units='xy', scale=1) → showing vector displacement

```

```

plt.grid(); plt.axis('square') → help visualisation
plt.xlim([x_min, x_max]), plt.ylim([y_min, y_max])
plt.xlabel('x1'); plt.ylabel('x2')
plt.title('Example scatter graph')
plt.legend()
plt.savefig('filename')

```

