

# Python for Science and Engg: Matrices & Least Square Fit

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# Outline

1 Matrices

2 Least Squares Fit

3 Summary

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# Matrices: Introduction

All matrix operations are done using **arrays**

# Matrices: Initializing

```
In []: c = array([[11, 12, 13],  
                  [21, 22, 23],  
                  [31, 32, 33]])
```

```
In []: c
```

```
Out []:
```

```
array([[11, 12, 13],  
       [21, 22, 23],  
       [31, 32, 33]])
```

# Initializing some special matrices

```
In []: ones((3,5))
```

```
Out []:
```

```
array([[ 1.,  1.,  1.,  1.,  1.],  
       [ 1.,  1.,  1.,  1.,  1.],  
       [ 1.,  1.,  1.,  1.,  1.]])
```

```
In []: ones_like([1, 2, 3, 4])
```

```
Out []: array([1, 1, 1, 1])
```

```
In []: identity(2)
```

```
Out []:
```

```
array([[ 1.,  0.],  
       [ 0.,  1.]])
```

Also available **zeros**, **zeros\_like**

# Accessing elements

```
In []: c
```

```
Out []:
```

```
array([[11, 12, 13],  
       [21, 22, 23],  
       [31, 32, 33]])
```

```
In []: c[1][2]
```

```
Out []: 23
```

```
In []: c[1,2]
```

```
Out []: 23
```

```
In []: c[1]
```

```
Out []: array([21, 22, 23])
```

# Changing elements

```
In []: c[1,1] = -22
```

```
In []: c
```

```
Out []:
```

```
array([[ 11,  12,  13],  
       [ 21, -22,  23],  
       [ 31,  32,  33]])
```

```
In []: c[1] = 0
```

```
In []: c
```

```
Out []:
```

```
array([[11, 12, 13],  
       [ 0,  0,  0],  
       [31, 32, 33]])
```

How to access one **column**?



# Slicing

```
In []: c[:,1]
```

```
Out []: array([12,  0, 32])
```

```
In []: c[1,:]
```

```
Out []: array([0, 0, 0])
```

```
In []: c[0:2,:]
```

```
Out []:
```

```
array([[11, 12, 13],  
       [ 0,  0,  0]])
```

```
In []: c[1:3,:]
```

```
Out []:
```

```
array([[ 0,  0,  0],  
       [31, 32, 33]])
```

# Slicing ...

```
In []: c[:2,:]
```

```
Out []:
```

```
array([[11, 12, 13],  
       [ 0,  0,  0]])
```

```
In []: c[1:,:]
```

```
Out []:
```

```
array([[ 0,  0,  0],  
       [31, 32, 33]])
```

```
In []: c[1:,:2]
```

```
Out []:
```

```
array([[ 0,  0],  
       [31, 32]])
```

# Striding

```
In []: c[:, :2, :]
```

```
Out []:
```

```
array([[11, 12, 13],  
       [31, 32, 33]])
```

```
In []: c[:, ::2]
```

```
Out []:
```

```
array([[11, 13],  
       [ 0,  0],  
       [31, 33]])
```

```
In []: c[:, :2, ::2]
```

```
Out []:
```

```
array([[11, 13],  
       [31, 33]])
```

# Shape of a matrix

```
In []: c
```

```
Out []:
```

```
array([[11, 12, 13],  
       [ 0,  0,  0],  
       [31, 32, 33]])
```

```
In []: c.shape
```

```
Out []: (3, 3)
```

Shape specifies shape or dimensions of a matrix

# Elementary image processing

```
In []: a = imread('lena.png')
```

```
In []: imshow(a)
```

```
Out []: <matplotlib.image.AxesImage object at 0xa0...
```

**imread** returns an array of shape (512, 512, 4) which represents an image of 512x512 pixels and 4 shades.

**imshow** renders the array as an image.

# Slicing & Striding Exercises

- Crop the image to get the top-left quarter
- Crop the image to get only the face
- Resize image to half by dropping alternate pixels

# Solutions

```
In []: imshow(a[:256,:256])
```

```
Out []: <matplotlib.image.AxesImage object at 0xb6
```

```
In []: imshow(a[200:400,200:400])
```

```
Out []: <matplotlib.image.AxesImage object at 0xb7
```

```
In []: imshow(a[:,::2],[:,::2])
```

```
Out []: <matplotlib.image.AxesImage object at 0xb7
```

# Transpose of a Matrix

```
In []: a = array([[ 1,  1,  2, -1],  
....:           [ 2,  5, -1, -9],  
....:           [ 2,  1, -1,  3],  
....:           [ 1, -3,  2,  7]])
```

```
In []: a.T
```

```
Out []:
```

```
array([[ 1,  2,  2,  1],  
       [ 1,  5,  1, -3],  
       [ 2, -1, -1,  2],  
       [-1, -9,  3,  7]])
```



# Matrix Addition

```
In []: b = array([[3, 2, -1, 5],  
                  [2, -2, 4, 9],  
                  [-1, 0.5, -1, -7],  
                  [9, -5, 7, 3]])
```

```
In []: a + b
```

```
Out []:
```

```
array([[ 4. ,  3. ,  1. ,  4. ],  
       [ 4. ,  3. ,  3. ,  0. ],  
       [ 1. ,  1.5, -2. , -4. ],  
       [10. , -8. ,  9. , 10. ]])
```

# Elementwise Multiplication

```
In []: a*b
```

```
Out []:
```

```
array([[ 3. ,  2. , -2. , -5. ],  
       [ 4. , -10. , -4. , -81. ],  
       [-2. ,  0.5,  1. , -21. ],  
       [ 9. , 15. , 14. , 21. ]])
```

# Matrix Multiplication

```
In []: dot(a, b)
```

```
Out []:
```

```
array([[ -6. ,   6. ,  -6. ,  -3. ],  
       [-64. ,  38.5, -44. ,  35. ],  
       [ 36. , -13.5,  24. ,  35. ],  
       [ 58. , -26. ,  34. , -15. ]])
```

# Inverse of a Matrix

```
In []: inv(a)
```

```
Out []:
```

```
array([[ -0.5 ,  0.55, -0.15,  0.7 ],  
       [ 0.75, -0.5 ,  0.5 , -0.75],  
       [ 0.5 , -0.15, -0.05, -0.1 ],  
       [ 0.25, -0.25,  0.25, -0.25]])
```

Try this:  $I = \text{dot}(a, \text{inv}(a))$

# Determinant and sum of all elements

```
In []: det(a)
```

```
Out []: 80.0
```

```
In []: sum(a)
```

```
Out []: 12
```

# Eigenvalues and Eigen Vectors

```
In []: e = array([[3,2,4],[2,0,2],[4,2,3]])
```

```
In []: eig(e)
```

```
Out []:
```

```
(array([-1.,  8., -1.]),  
 array([[ -0.74535599,  0.66666667, -0.1931126 ],  
        [ 0.2981424 ,  0.33333333, -0.78664085],  
        [ 0.59628479,  0.66666667,  0.58643303]]))
```

```
In []: eigvals(e)
```

```
Out []: array([-1.,  8., -1.])
```

# Outline

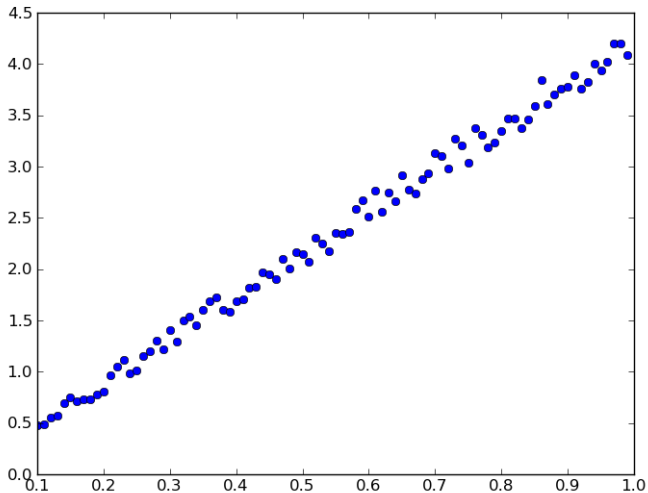
1 Matrices

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3 Summary

# $L$ vs. $T^2$ - Scatter

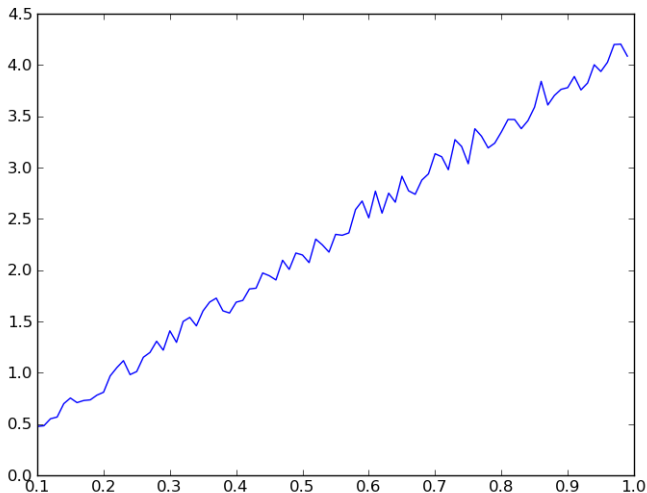
Linear trend visible.





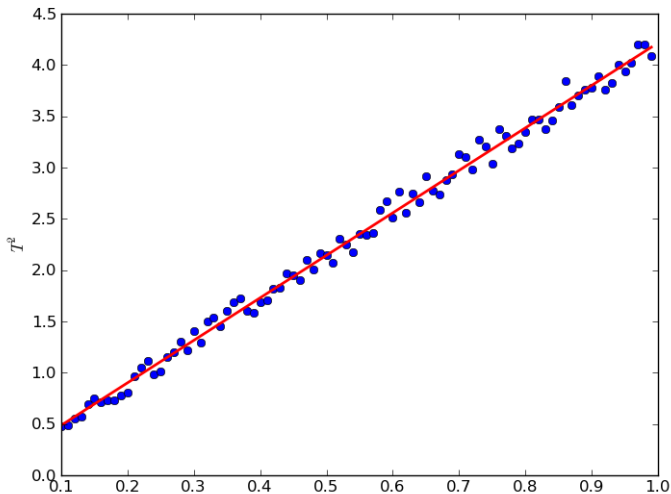
# $L$ vs. $T^2$ - Line

This line does not make any mathematical sense.



# $L$ vs. $T^2$ - Least Square Fit

This is what our intention is.



# Matrix Formulation

- We need to fit a line through points for the equation  $T^2 = m \cdot L + c$
- In matrix form, the equation can be represented as

$$T_{sq} = A \cdot p, \text{ where } T_{sq} \text{ is } \begin{bmatrix} T_1^2 \\ T_2^2 \\ \vdots \\ T_N^2 \end{bmatrix}, \text{ A is } \begin{bmatrix} L_1 & 1 \\ L_2 & 1 \\ \vdots & \vdots \\ L_N & 1 \end{bmatrix} \text{ and}$$

$$p \text{ is } \begin{bmatrix} m \\ c \end{bmatrix}$$

- We need to find  $p$  to plot the line

# Getting $L$ and $T^2$

```
In []: L = []  
In []: t = []  
In []: for line in open('pendulum.txt'):  
.....     point = line.split()  
.....     L.append(float(point[0]))  
.....     t.append(float(point[1]))  
.....  
.....
```

# Getting $L$ and $T^2$ ...

```
In []: L = array(L)
```

```
In []: t = array(t)
```

```
In []: tsq = t*t
```

# Generating A

```
In []: A = array([L, ones_like(L)])  
In []: A = A.T
```

# lstsq...

- Now use the **lstsq** function
- Along with a lot of things, it returns the least squares solution

```
In []: result = lstsq(A,tsq)
In []: coef = result[0]
```

# Least Square Fit Line ...

We get the points of the line from `coef`

```
In []: Tline = coef[0]*L + coef[1]
```

```
In []: Tline.shape
```

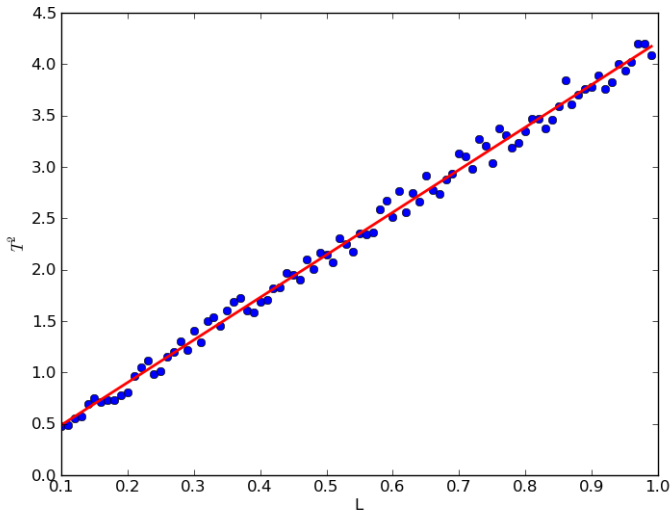
- Now plot `Tline` vs. `L`, to get the Least squares fit line.

```
In []: plot(L, Tline, 'r')
```

```
In []: plot(L, tsq, 'o')
```



# Least Squares Fit



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# What did we learn?

- Matrices
  - Initializing
  - Accessing elements
  - Slicing and Striding
  - Transpose
  - Addition
  - Multiplication
  - Inverse of a matrix
  - Determinant
  - Eigenvalues and Eigen vector
- Least Square Curve fitting