

Interpolation and Least Squares

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Abstract—Through examples, this manual introduces methods for polynomial interpolation and curve fitting like Newton, Lagrange's and least squares methods. Python codes are provided for all these methods.

Solution: The following code generates the coefficients

1.121, 0.002, 0.0005, -0.0015, 0.002, -0.0025
(1.1)

1 NEWTON'S INTERPOLATION FORMULA

Theorem 1.1. Newton's interpolation formula for a function $f(x)$ is given by [1], [2]

$$f(x) = \sum_{k=0}^{\infty} \Delta^k y^u C_k \quad (0.1)$$

where

$${}^u C_k = \frac{u(u-1)(u-2)\cdots(u-k+1)}{k!} \quad (0.2)$$

$$\Delta^n y = \sum_{k=0}^n \binom{n}{k} (-1)^{n-k} y_k \quad (0.3)$$

$$u = \frac{x-a}{h} \quad (0.4)$$

with a as the initial point and h , the step-size.

Problem 1. For the table in Table I, obtain (0.3) for $n = 0, 1, \dots, 5$.

x	0	0.001	0.002	0.003	0.004	0.005
y	1.121	1.123	1.1255	1.127	1.128	1.1285

TABLE I

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```
#Forward differences
coefficients
#for Newton's Interpolation
Formula
import numpy as np
from scipy.misc import comb

def diff_f(n,y):
    k = np.asarray(range(n
+1))
    coeff = comb(n,k)
    return np.sum(coeff
*(-1)**(n-k)*y[0:n
+1])

x = np.array([0,0.001,0.002,
\
0.003,0.004,0.005])
y = np.array
([1.121,1.123,1.1255
\
,1.127,1.128,1.1285])
n = 5

for k in range(n+1):
    print(diff_f(k,y))
```

Problem 2. Obtain the interpolating polynomial $P_5(u)$.

Solution: From (1.1) and (0.1),

$$\begin{aligned}
 P_5(u) = & 1.121 + u \times .002 \\
 & + \frac{u(u-1)}{2}(.0005) + \frac{u(u-1)(u-2)}{3!} \times (-.0015) \\
 & + \frac{u(u-1)(u-2)(u-3)}{4!}(.002) + \\
 & \frac{u(u-1)(u-2)(u-3)(u-4)}{5!} \times (-.0025).
 \end{aligned} \tag{2.1}$$

Problem 3. Using (2.1), interpolate the value of the function at $x = 0.0045$.

Solution: From (0.4),

$$u = \frac{x-a}{h} = \frac{0.0045-0}{0.001} = 4.5 \tag{3.1}$$

The following code results in

$$P_5(u) = 1.128400390625 \tag{3.2}$$

```

#Forward differences
coefficients
#for Newton's Interpolation
Formula
import numpy as np
from scipy.misc import comb
import mpmath as mp

def diff_f(n,y):
    k = np.asarray(range(n
        +1))
    coeff = comb(n,k)
    return np.sum(coeff
        *(-1)**(n-k)*y[0:n
        +1])

x = np.array([0,0.001,0.002,
    \
    0.003,0.004,0.005])
y = np.array
    
```

```

([1.121,1.123,1.1255
    \
    ,1.127,1.128,1.1285])
n = 5
u = 4.5
ans = 0
for k in range(n+1):
    ans += diff_f(k,y)*mp.
        binomial(u,k)

print(ans)
    
```

2 LAGRANGE'S INTERPOLATION FORMULA

Theorem 2.1. Given a set of data points

$$(x_0, y_0), \dots, (x_j, y_j), \dots, (x_k, y_k)$$

where no two x_j are the same, [3],

$$f(x) = \sum_{j=0}^k y_j \prod_{\substack{0 \leq m \leq k \\ m \neq j}} \frac{x - x_m}{x_j - x_m} \tag{3.3}$$

Problem 4. Using the following data [4], find by (3.3), the value of $f(x)$ at $x = 10$

x	9.3	9.6	10.2	10.4	10.8
y	11.4	12.8	14.7	17	19.8

TABLE II

Solution: The following code gives

$$f(10) = 13.1978451179 \tag{4.1}$$

```

#Forward differences
coefficients
#for Newton's Interpolation
Formula
import numpy as np
from scipy.interpolate import
    lagrange
from numpy.polynomial.
    polynomial import Polynomial

x = np.array([9.3,9.6,10.2,
    
```

```

\
10.4,10.8])
y = np.array([11.4,12.8,14.7
\
,17,19.8])

poly = lagrange(x, y)
p = Polynomial(poly).coef
print(np.polyval(p,10))

```

Problem 5. Using (3.3), find x if $f(x) = 16$.

Solution: Interchange x and y in the previous code and evaluate $f(16)$.

3 LEAST SQUARES

Use the following data for the following problems.

$$(x_i, y_i) = (0, 5), (2, 4), (4, 1), (6, 6), (8, 7). \quad (5.1)$$

Problem 6. Obtain a matrix solution to find a line of best fit from the given data.

Solution: Let

$$y = mx + c \quad (6.1)$$

be the equation of the line. Using the data in (5.1) and (6.1) results in the matrix equation

$$\mathbf{Az} = \mathbf{y} \quad (6.2)$$

where

$$\mathbf{A} = \begin{pmatrix} x_1 & 1 \\ x_2 & 1 \\ \vdots & \vdots \\ x_5 & 1 \end{pmatrix}, \mathbf{z} = \begin{pmatrix} m \\ c \end{pmatrix}, \mathbf{y} = \begin{pmatrix} y_1 \\ y_2 \\ \vdots \\ y_5 \end{pmatrix}, \quad (6.3)$$

The solution to (6.2) is obtained as

$$\mathbf{z} = (\mathbf{A}^T \mathbf{A})^{-1} \mathbf{A}^T \mathbf{y} \quad (6.4)$$

Problem 7. Obtain \mathbf{z} in (6.4) and sketch (6.1) using this information.

Solution: The following code results in

$$\mathbf{z} = \begin{pmatrix} 0.3 \\ 3.4 \end{pmatrix} \quad (7.1)$$

and plots Fig. 7.

```

import numpy as np
import matplotlib.pyplot as plt

x = np.matrix([0,2,4,
\
6,8])
y = np.matrix([5,4,1
\
,6,7])

A = np.column_stack([x.T,np.
ones([5,1])])

z = np.linalg.inv(A.T*A)*A.T*y.
T
yhat=A*z
x = np.array(x)[0]
y = np.array(y)[0]
yhat = np.array(yhat.T)[0]
plt.plot(x,yhat,label='estimate
')
plt.plot(x,y,'o',label='data')
plt.xlabel('$x$')
plt.ylabel('$y$')
plt.legend(loc='best',prop={'
size':11})
plt.savefig('../figs/ls_line.
eps')
plt.show()

```

Problem 8. Obtain the estimate for $y = ax^2 + bx + c$.

Solution: The following code results in

$$\mathbf{z} = \begin{pmatrix} 0.21428571 \\ -1.41428571 \\ 5.11428571 \end{pmatrix} \quad (8.1)$$

and plots Fig. 8.

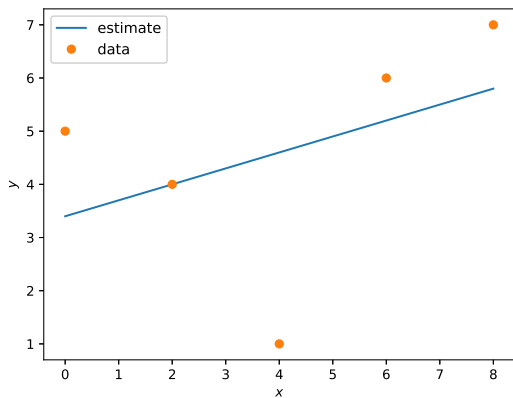


Fig. 7: Linear Estimate.

```
eps')
plt.show()
```

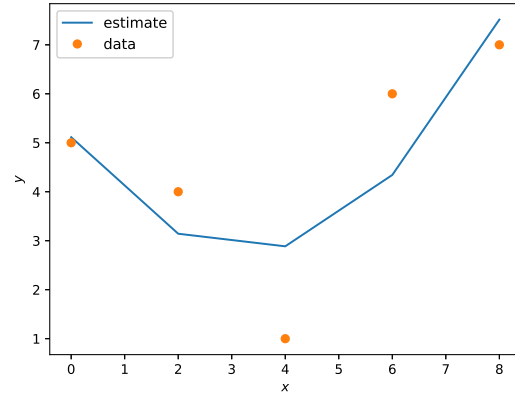


Fig. 8: Quadratic Estimate.

```
import numpy as np
import matplotlib.pyplot as plt

x = np.matrix([0,2,4,
\
6,8])
x2=np.square(x)
y = np.matrix([5,4,1
\
,6,7])

A = np.column_stack([x2.T,x.T,
np.ones([5,1])])
z = np.linalg.inv(A.T*A)*A.T*y.
T
print(z)
yhat=A*z
x = np.array(x)[0]
y = np.array(y)[0]
yhat = np.array(yhat.T)[0]
plt.plot(x,yhat,label='estimate
')
plt.plot(x,y,'o',label='data')
plt.xlabel('$x$')
plt.ylabel('$y$')
plt.legend(loc='best',prop={'
size':11})
plt.savefig('../figs/ls_quad.
```

Problem 9. Write a program to find a curve of the form $y = ae^{bx}$ from the given data.

Solution: Taking logarithms,

$$\ln y = \ln a + bx \quad (9.1)$$

Solution: Note that (9.1) has the same form as (6.1). The following code results in

$$\begin{pmatrix} a \\ b \end{pmatrix} = \begin{pmatrix} 1.05540067 \\ 1.13099846 \end{pmatrix} \quad (9.2)$$

and plots Fig. 9.

```
import numpy as np
import matplotlib.pyplot as plt

x = np.matrix([0,2,4,
\
6,8])
y = np.matrix([5,4,1
\
,6,7])
ylog = np.log(y)
A = np.column_stack([x.T,np.
ones([5,1])])
z = np.linalg.inv(A.T*A)*A.T*
ylog.T
```

```

print(np.exp(z[0]), z[1])
yloghat=A*z
yhat = np.exp(yloghat)
x = np.array(x)[0]
y = np.array(y)[0]
yhat = np.array(yhat.T)[0]
plt.plot(x, yhat, label='estimate
')
plt.plot(x, y, 'o', label='data')
plt.xlabel('$x$')
plt.ylabel('$y$')
plt.legend(loc='best', prop={'
size':11})
#plt.savefig(' ../figs/ls_exp.
eps ')
plt.show()

```

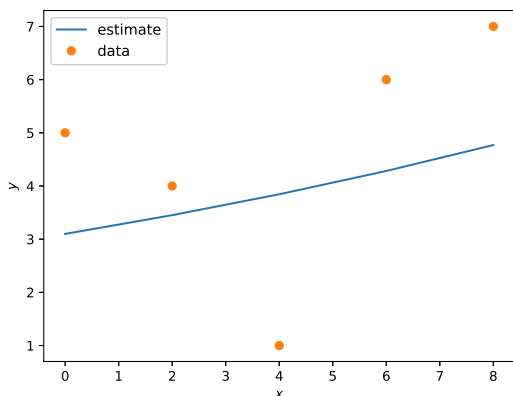


Fig. 9: Exponential Estimate.

```

import numpy as np
import matplotlib.pyplot as plt

x = np.matrix([0.5, 2, 4,
\
6, 8])
y = np.matrix([5, 4, 1
\
, 6, 7])
ylog = np.log(y)
xlog = np.log(x)
A = np.column_stack([xlog.T, np.
ones([5, 1])])
z = np.linalg.inv(A.T*A)*A.T*
ylog.T
print(np.exp(z[0]), z[1])
yloghat=A*z
yhat = np.exp(yloghat)
x = np.array(x)[0]
y = np.array(y)[0]
yhat = np.array(yhat.T)[0]
plt.plot(x, yhat, label='estimate
')
plt.plot(x, y, 'o', label='data')
plt.xlabel('$x$')
plt.ylabel('$y$')
#plt.legend(loc='best', prop={'
size':11})
plt.savefig(' ../figs/ls_pow.eps
')
plt.show()

```

Problem 10. Write a program to find a curve of the form $y = ax^b$ from the given data.

Solution: Taking logarithms,

$$\ln y = \ln a + b \ln x \quad (10.1)$$

Solution: Note that (10.1) has the same form as (6.1). The following code results in

$$\begin{pmatrix} a \\ b \end{pmatrix} = \begin{pmatrix} 1.00450417 \\ 1.34195488 \end{pmatrix} \quad (10.2)$$

and plots Fig. 10.

REFERENCES

- [1] Wikipedia. [Online]. Available: https://en.wikipedia.org/wiki/Finite_difference
- [2] NPTEL. [Online]. Available: <http://www.nptel.ac.in/courses/122104018/node109.html>
- [3] Wikipedia. [Online]. Available: https://en.wikipedia.org/wiki/Lagrange_polynomial
- [4] NPTEL. [Online]. Available: <http://www.nptel.ac.in/courses/122104018/node113.html>

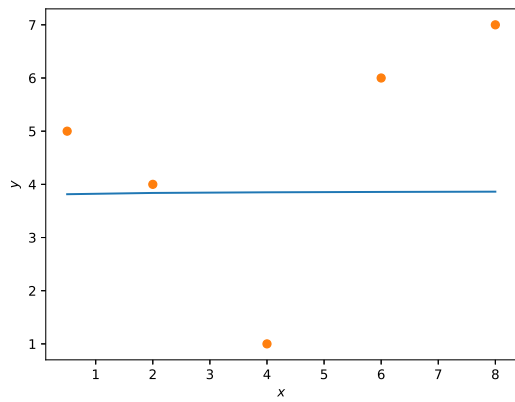


Fig. 10: Estimate.