

hw12

April 7, 2024

```
[1]: import numpy as np
import numpy.linalg as la
```

```
[4]: r = { "student01": {"% exercises": 20, "score": 55},
          "student02": {"% exercises": 100, "score": 100},
          "student03": {"% exercises": 90, "score": 100},
          "student04": {"% exercises": 70, "score": 70},
          "student05": {"% exercises": 50, "score": 75},
          "student06": {"% exercises": 10, "score": 25},
          "student07": {"% exercises": 30, "score": 60}
        }
```

```
[7]: M = np.array([[r[s]["% exercises"], r[s]["score"]] for s in r.keys()])
M
```

```
[7]: array([[ 20,  55],
           [100, 100],
           [ 90, 100],
           [ 70,  70],
           [ 50,  75],
           [ 10,  25],
           [ 30,  60]])
```

```
[85]: from itertools import chain, combinations
def nset(n, iterable):
    s = list(iterable)
    return list(combinations(s, n))
```

```
[86]: nset(3, ['a', 'b', 'c', 'd', 'e'])
```

```
[86]: [('a', 'b', 'c'),
       ('a', 'b', 'd'),
       ('a', 'b', 'e'),
       ('a', 'c', 'd'),
       ('a', 'c', 'e'),
       ('a', 'd', 'e'),
       ('b', 'c', 'd'),
       ('b', 'c', 'e'),
```

```

('b', 'd', 'e'),
('c', 'd', 'e')]

```

```

[89]: correct = { 'a': 150,
                  'b': 125,
                  'c': 275,
                  'd': 80,
                  'e': 300
                }

def mkestimates(correct):
    keys = correct.keys()
    return { (x,y,z): np.round(correct[x] + correct[y] + correct[z] + rng.
    ↪uniform(-2,2),2)
            for (x,y,z) in nset(3,keys) }

```

```

[91]: mass_estimates = mkestimates(correct)

mass_estimates

```

```

[91]: {('a', 'b', 'c'): 551.03,
      ('a', 'b', 'd'): 353.19,
      ('a', 'b', 'e'): 574.36,
      ('a', 'c', 'd'): 506.1,
      ('a', 'c', 'e'): 724.92,
      ('a', 'd', 'e'): 531.9,
      ('b', 'c', 'd'): 478.21,
      ('b', 'c', 'e'): 701.98,
      ('b', 'd', 'e'): 504.75,
      ('c', 'd', 'e'): 653.07}

```

```

[93]: # as a vector...
b=[mass_estimates[k] for k in mass_estimates.keys() ]
b

```

```

[93]: [551.03, 353.19, 574.36, 506.1, 724.92, 531.9, 478.21, 701.98, 504.75, 653.07]

```

```

[95]: def sbv(i,n):
        return np.array([1 if j ==i else 0 for j in range(n)])

def sbvalpha(elem,ls):
    return sbv(list(ls).index(elem),len(ls))

```

```

M = np.array([sbvalpha(x,correct.keys()) + sbvalpha(y,correct.keys()) +
↳sbvalpha(z,correct.keys()) for (x,y,z) in mass_estimates])
b = np.array([mass_estimates[x] for x in mass_estimates.keys()])
(M,b)

```

```

[95]: (array([[1, 1, 1, 0, 0],
            [1, 1, 0, 1, 0],
            [1, 1, 0, 0, 1],
            [1, 0, 1, 1, 0],
            [1, 0, 1, 0, 1],
            [1, 0, 0, 1, 1],
            [0, 1, 1, 1, 0],
            [0, 1, 1, 0, 1],
            [0, 1, 0, 1, 1],
            [0, 0, 1, 1, 1]]),
      array([551.03, 353.19, 574.36, 506.1 , 724.92, 531.9 , 478.21, 701.98,
            504.75, 653.07]))

```

```

[97]: (M,b)

```

```

[97]: (array([[1, 1, 1, 0, 0],
            [1, 1, 0, 1, 0],
            [1, 1, 0, 0, 1],
            [1, 0, 1, 1, 0],
            [1, 0, 1, 0, 1],
            [1, 0, 0, 1, 1],
            [0, 1, 1, 1, 0],
            [0, 1, 1, 0, 1],
            [0, 1, 0, 1, 1],
            [0, 0, 1, 1, 1]]),
      array([551.03, 353.19, 574.36, 506.1 , 724.92, 531.9 , 478.21, 701.98,
            504.75, 653.07]))

```

```

[102]: M.transpose() @ M

```

```

[102]: array([[6, 3, 3, 3, 3],
            [3, 6, 3, 3, 3],
            [3, 3, 6, 3, 3],
            [3, 3, 3, 6, 3],
            [3, 3, 3, 3, 6]])

```

```

[74]: aa=la.lstsq(M,b,rcond=None)
      aa

```

```

[74]: (array([149.6235 , 125.3465 , 275.0235 , 79.79183333,
            300.5365  ]),
      array([8.07883383]),

```

```
5,  
array([4.24264069, 1.73205081, 1.73205081, 1.73205081, 1.73205081]))
```

```
[117]: def f(x,v0,h0):  
        return -9.8*x**2/2 + v0*x + h0  
  
def ff(x):  
    return f(x,10,200)  
  
[ (x,ff(x)) for x in np.arange(0,7.55,.05)]
```

```
[117]: [(0.0, 200.0),  
(0.05, 200.48775),  
(0.1, 200.951),  
(0.15000000000000002, 201.38975),  
(0.2, 201.804),  
(0.25, 202.19375),  
(0.30000000000000004, 202.559),  
(0.35000000000000003, 202.89975),  
(0.4, 203.216),  
(0.45, 203.50775),  
(0.5, 203.775),  
(0.55, 204.01775),  
(0.6000000000000001, 204.236),  
(0.65, 204.42975),  
(0.7000000000000001, 204.599),  
(0.75, 204.74375),  
(0.8, 204.864),  
(0.8500000000000001, 204.95974999999999),  
(0.9, 205.031),  
(0.9500000000000001, 205.07775),  
(1.0, 205.1),  
(1.05, 205.09775),  
(1.1, 205.071),  
(1.1500000000000001, 205.01975),  
(1.2000000000000002, 204.944),  
(1.25, 204.84375),  
(1.3, 204.719),  
(1.35, 204.56975),  
(1.4000000000000001, 204.396),  
(1.4500000000000002, 204.19774999999998),  
(1.5, 203.975),  
(1.55, 203.72775),  
(1.6, 203.456),  
(1.6500000000000001, 203.15975),  
(1.7000000000000002, 202.839),  
(1.75, 202.49375),
```

(1.8, 202.124),
 (1.85, 201.72975),
 (1.9000000000000001, 201.311),
 (1.9500000000000002, 200.86775),
 (2.0, 200.4),
 (2.0500000000000003, 199.90775),
 (2.1, 199.391),
 (2.15, 198.84975),
 (2.2, 198.284),
 (2.25, 197.69375),
 (2.3000000000000003, 197.079),
 (2.35, 196.43975),
 (2.4000000000000004, 195.77599999999998),
 (2.45, 195.08775),
 (2.5, 194.375),
 (2.5500000000000003, 193.63774999999998),
 (2.6, 192.876),
 (2.6500000000000004, 192.08974999999998),
 (2.7, 191.279),
 (2.75, 190.44375),
 (2.8000000000000003, 189.584),
 (2.85, 188.69975),
 (2.9000000000000004, 187.791),
 (2.95, 186.85775),
 (3.0, 185.9),
 (3.0500000000000003, 184.91774999999998),
 (3.1, 183.911),
 (3.1500000000000004, 182.87974999999997),
 (3.2, 181.82399999999998),
 (3.25, 180.74375),
 (3.3000000000000003, 179.63899999999998),
 (3.35, 178.50975),
 (3.4000000000000004, 177.356),
 (3.45, 176.17775),
 (3.5, 174.975),
 (3.5500000000000003, 173.74775),
 (3.6, 172.49599999999998),
 (3.6500000000000004, 171.21974999999998),
 (3.7, 169.91899999999998),
 (3.75, 168.59375),
 (3.8000000000000003, 167.24399999999997),
 (3.85, 165.86974999999998),
 (3.9000000000000004, 164.47099999999998),
 (3.95, 163.04775),
 (4.0, 161.6),
 (4.05, 160.12775),
 (4.1000000000000005, 158.63099999999997),

(4.15, 157.109749999999996),
(4.2, 155.564),
(4.25, 153.993749999999998),
(4.3, 152.399),
(4.3500000000000005, 150.779749999999998),
(4.4, 149.135999999999997),
(4.45, 147.467749999999997),
(4.5, 145.774999999999998),
(4.55, 144.05775),
(4.6000000000000005, 142.315999999999997),
(4.65, 140.549749999999996),
(4.7, 138.758999999999996),
(4.75, 136.94375),
(4.8000000000000001, 135.103999999999996),
(4.8500000000000005, 133.23975),
(4.9, 131.350999999999997),
(4.95, 129.43775),
(5.0, 127.499999999999999),
(5.0500000000000001, 125.537749999999996),
(5.1000000000000005, 123.550999999999996),
(5.15, 121.539749999999997),
(5.2, 119.503999999999999),
(5.25, 117.44375),
(5.3000000000000001, 115.358999999999995),
(5.3500000000000005, 113.249749999999995),
(5.4, 111.115999999999999),
(5.45, 108.957749999999998),
(5.5, 106.774999999999998),
(5.5500000000000001, 104.567749999999993),
(5.6000000000000005, 102.335999999999996),
(5.65, 100.079749999999996),
(5.7, 97.798999999999998),
(5.75, 95.493749999999998),
(5.8000000000000001, 93.163999999999996),
(5.8500000000000005, 90.809749999999998),
(5.9, 88.430999999999998),
(5.95, 86.027749999999997),
(6.0, 83.6),
(6.0500000000000001, 81.147749999999995),
(6.1000000000000005, 78.670999999999994),
(6.15, 76.169749999999996),
(6.2, 73.643999999999995),
(6.25, 71.09375),
(6.3000000000000001, 68.518999999999992),
(6.3500000000000005, 65.919749999999996),
(6.4, 63.2959999999999935),
(6.45, 60.64775),

```
(6.5, 57.974999999999994),
(6.5500000000000001, 55.277749999999994),
(6.6000000000000005, 52.555999999999926),
(6.65, 49.809749999999998),
(6.7, 47.038999999999999),
(6.75, 44.243749999999998),
(6.8000000000000001, 41.423999999999995),
(6.8500000000000005, 38.579749999999996),
(6.9, 35.7109999999999956),
(6.95, 32.817749999999996),
(7.0, 29.899999999999977),
(7.0500000000000001, 26.957749999999947),
(7.1000000000000005, 23.990999999999993),
(7.15, 20.999749999999977),
(7.2, 17.983999999999952),
(7.25, 14.943749999999966),
(7.3000000000000001, 11.878999999999905),
(7.3500000000000005, 8.789749999999997),
(7.4, 5.675999999999931),
(7.45, 2.537749999999996),
(7.5, -0.625)]
```

```
[107]: x = np.arange(0,20,.5)
x
```

```
[107]: array([ 0. ,  0.5,  1. ,  1.5,  2. ,  2.5,  3. ,  3.5,  4. ,  4.5,  5. ,
          5.5,  6. ,  6.5,  7. ,  7.5,  8. ,  8.5,  9. ,  9.5, 10. , 10.5,
          11. , 11.5, 12. , 12.5, 13. , 13.5, 14. , 14.5, 15. , 15.5, 16. ,
          16.5, 17. , 17.5, 18. , 18.5, 19. , 19.5])
```

```
[134]: #height_estimates = { x: np.round(ff(x) + rng.uniform(-5,5),2) for x in np.
      ↪ arange(0,7.55,.25) }
height_estimates
```

```
[134]: {0.0: 199.6,
        0.25: 202.96,
        0.5: 207.23,
        0.75: 208.29,
        1.0: 207.47,
        1.25: 203.96,
        1.5: 199.18,
        1.75: 202.91,
        2.0: 204.29,
        2.25: 196.31,
        2.5: 195.71,
        2.75: 187.89,
        3.0: 187.61,
```

```
3.25: 177.12,  
3.5: 171.07,  
3.75: 171.89,  
4.0: 158.68,  
4.25: 152.64,  
4.5: 146.7,  
4.75: 138.52,  
5.0: 127.27,  
5.25: 122.38,  
5.5: 103.97,  
5.75: 96.91,  
6.0: 83.08,  
6.25: 67.34,  
6.5: 55.75,  
6.75: 45.42,  
7.0: 25.33,  
7.25: 14.67,  
7.5: -1.45}
```

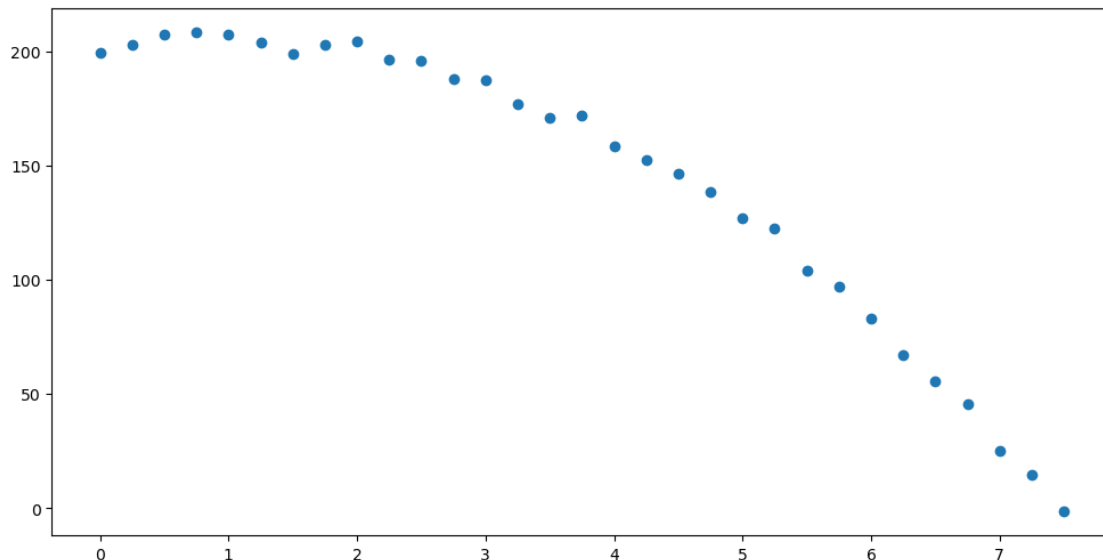
```
[136]: height_estimates = {0.0: 199.6,  
                           0.25: 202.96,  
                           0.5: 207.23,  
                           0.75: 208.29,  
                           1.0: 207.47,  
                           1.25: 203.96,  
                           1.5: 199.18,  
                           1.75: 202.91,  
                           2.0: 204.29,  
                           2.25: 196.31,  
                           2.5: 195.71,  
                           2.75: 187.89,  
                           3.0: 187.61,  
                           3.25: 177.12,  
                           3.5: 171.07,  
                           3.75: 171.89,  
                           4.0: 158.68,  
                           4.25: 152.64,  
                           4.5: 146.7,  
                           4.75: 138.52,  
                           5.0: 127.27,  
                           5.25: 122.38,  
                           5.5: 103.97,  
                           5.75: 96.91,  
                           6.0: 83.08,  
                           6.25: 67.34,  
                           6.5: 55.75,  
                           6.75: 45.42,
```



```
7.0: 25.33,  
7.25: 14.67,  
7.5: -1.45}
```

```
[197]: import matplotlib.pyplot as plt  
  
## lists x,y have been populated; lets plot the points  
def plot_data(x,y):  
    fig, ax = plt.subplots(figsize=(12,6))  
    return ax.plot(x,y,"o")  
  
x1 = np.array(list(height_estimates.keys()))  
y1 = np.array(list(height_estimates.values()))  
plot_data(x1,y1)
```

```
[197]: [<matplotlib.lines.Line2D at 0x7f8f4f957fd0>]
```



```
[144]: Mf = np.array([[t**2, t, 1] for t in height_estimates.keys() ])  
Mb = np.array([height_estimates[k] for k in height_estimates.keys()])  
  
coeffs = la.lstsq(Mf,Mb,rcond=None)[0]  
coeffs
```

```
[144]: array([ -4.87577207,   9.52224214, 201.12954545])
```

```
[153]: MMf = np.array([[t, 1] for t in height_estimates.keys() ])  
MMb = np.array([height_estimates[k] + (9.8/2)*k**2 for k in height_estimates.  
    ↪keys()])
```

```
coeffs_alt = la.lstsq(MMf,MMb,rcond=None)[0]
coeffs_alt
```

```
[153]: array([ 9.70395161, 200.90997984])
```

```
[201]: def mk_func(cc):

    # get the coefficients and report them
    if len(cc) == 3:
        alpha,beta,gamma = cc
    else:
        alpha = -9.8/2
        beta,gamma = cc
    print(f"f(t) = {alpha:.04}*t^2 + {beta:.04}*t + {gamma:.04}")

    # return the linear function determined by these coefficients
    return lambda t:alpha*t**2 + beta*t + gamma

fa = mk_func(coeffs)
fb = mk_func(coeffs_alt)
```

$f(t) = -4.876*t^2 + 9.522*t + 201.1$

$f(t) = -4.9*t^2 + 9.704*t + 200.9$

```
[191]: aaaa = np.arange(5)
fa(aaaa)
(aaaa,x1)
```

```
[191]: (array([0, 1, 2, 3, 4]),
[0.0,
 0.25,
 0.5,
 0.75,
 1.0,
 1.25,
 1.5,
 1.75,
 2.0,
 2.25,
 2.5,
 2.75,
 3.0,
 3.25,
 3.5,
 3.75,
 4.0,
 4.25,
```

```

4.5,
4.75,
5.0,
5.25,
5.5,
5.75,
6.0,
6.25,
6.5,
6.75,
7.0,
7.25,
7.5])

```

```

[174]: def plot_curve_fit(x0,f,x,y):
        # graph the line with slope alpha and y-intercept beta, and plot the data
        ↪points
        #
        fig,ax = plt.subplots(figsize=(12,6))
        ax.plot(x0,f(x0))
        ax.plot(x,y,'o')

        # ax.plot(x0,f(x0))

        return fig,ax

```

```

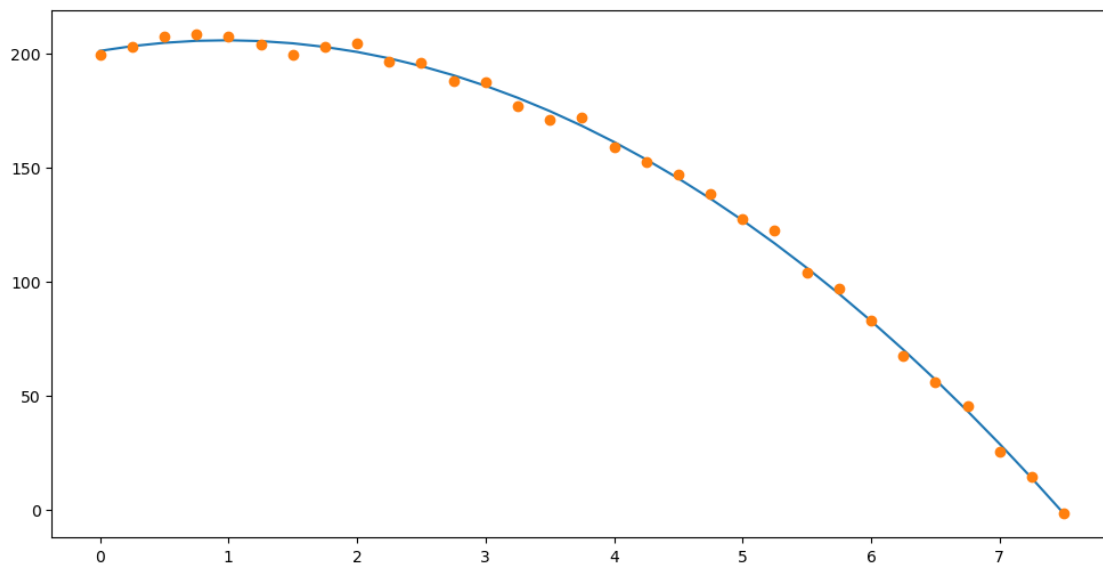
[202]: plot_curve_fit(x1,fa,x1,y1)

```

```

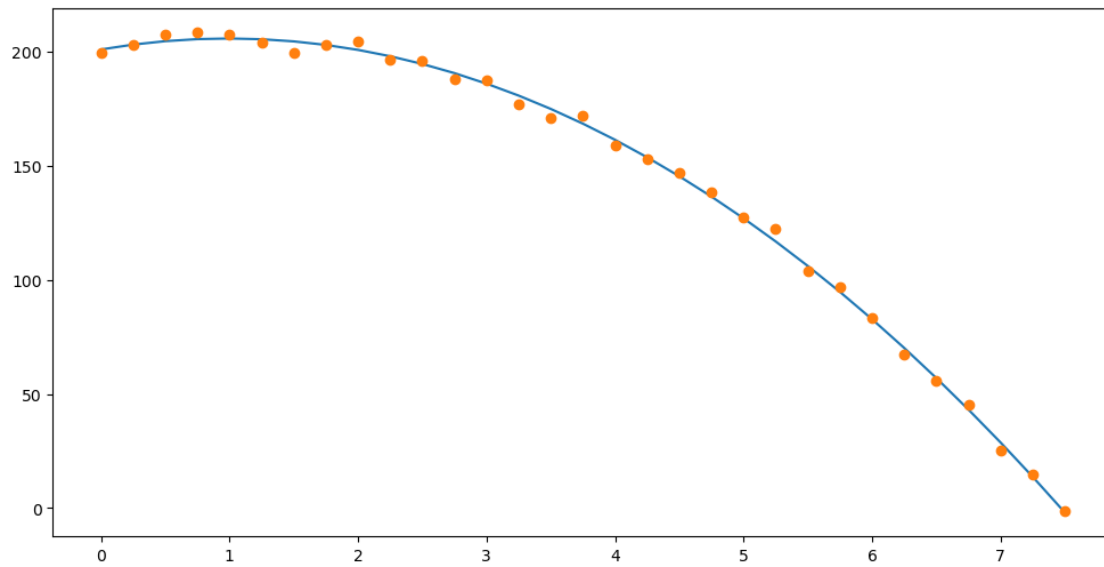
[202]: (<Figure size 1200x600 with 1 Axes>, <Axes: >)

```



```
[203]: plot_curve_fit(x1,fb,x1,y1)
```

```
[203]: (<Figure size 1200x600 with 1 Axes>, <Axes: >)
```



```
[ ]:
```