# ch\_15\_assignment

May 9, 2023

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```
[]: from IPython.core.interactiveshell import InteractiveShell
InteractiveShell.ast_node_interactivity = 'all'
```

## 1 Equation Solving

### 1.1 Importing Modules

```
[]: from scipy import linalg as la
  from scipy import optimize
  import sympy
  sympy.init_printing()
  import numpy as np
  import matplotlib.pyplot as plt
```

#### 1.2 Inner Product

```
import numpy as np

def innerVec(a,b):
    rowA = a.shape[0]
    rowB = b.shape[0]

if (rowA != rowB):
        print('Wrong input data')
        return None

c = 0.0
    for i in range(rowA):
        c += a[i]*b[i]
    return c

if __name__ == '__main__':
```

```
a = np.array([1.1, 1.2, 1.3])
b = np.array([2.1, 2.2, 2.3])

c = innerVec(a, b)
print("Inner product of vectors=", c)
```

Inner product of vectors= 7.94000000000001

```
[]: # InnerVect2.py
import numpy as np

a = np.array([1.1, 1.2, 1.3])
b = np.array([2.1, 2.2, 2.3])

c = np.dot(a, b)
print("inner product = ", c)
```

inner product = 7.94

## 1.3 Square Systems

```
[]: A = sympy.Matrix([[2, 3],[5, 4]])
B = sympy.Matrix([4, 3])

A.rank()

A.condition_number()
sympy.N(_)
A.norm()

[]: 2
[]: \frac{\sqrt{2\sqrt{170} + 27}}{\sqrt{27 - 2\sqrt{170}}}
[]: 7.58240137440151
[]: 3\sqrt{6}
[]: A = np.array([[2, 3], [5, 4]])
b = np.array([4, 3])
np.linalg.matrix_rank(A)
np.linalg.cond(A)
```

```
np.linalg.norm(A)
```

[]: 2

[]: 7.58240137440151

[]: 7.34846922834953

#### 1.3.1 Gaussian Elimination

```
[]: # GaussElimin.py
     111
     x = GaussElimin(A, b)
        Equiation Ax=b
     import numpy as np
     def GaussElimin(A, b):
         n = len(b)
         # Elimination
         for k in range(0, n-1):
             for i in range(k+1, n):
                 if A[i, k] != 0.0:
                     lam = A[i, k] / A[k, k]
                     A[i, k+1:n] = lam * A[k, k+1:n]
                     b[i] -= lam *b[k]
         # Substition
         for k in range(n-1, -1, -1):
             b[k] = (b[k] - np.dot(A[k, k+1:n], b[k+1:n])) / A[k, k]
         return b
```

[]: '\nx = GaussElimin(A, b)\n Equiation Ax=b\n'

```
borg = b.copy()
         x = GaussElimin(A, b)
         print('\nresult: x=', x)
         print('\ncheck: Ax-b =', np.dot(Aorg, x) - borg)
    result: x= [10. 22. 14.]
    check: Ax-b = [1.77635684e-15 7.10542736e-15 -3.55271368e-15]
    1.3.2 LU decomposion
[]: A = sympy.Matrix([[2, 3], [5, 4]])
     b = sympy.Matrix([4, 3])
     L, U, _ = A.LUdecomposition()
     L
     U
     L*U
     x = A.solve(b); x
[ ]: [1 0]
     \frac{5}{2}
        1
[]:[2]
     0 -\frac{7}{2}
[ ]: [2  3]
     |5 4|
[]:[-1]
     2
[]: A = np.array([[2, 3], [5, 4]])
     b = np.array([4, 3])
     P, L, U = la.lu(A)
     L
     P.dot(L.dot(U))
     la.solve(A, b)
[]: array([[1., 0.],
            [0.4, 1.]])
```

## 1.4 Rectangular Systems

```
[]: x_vars = sympy.symbols("x_1, x_2, x_3")
A = sympy.Matrix([[1, 2, 3], [4, 5, 7]])
x = sympy.Matrix(x_vars)
b = sympy.Matrix([7, 8])
sympy.solve(A*x - b, x_vars)
```

$$\left\{x_1: \frac{x_3}{3} - \frac{19}{3}, \ x_2: \frac{20}{3} - \frac{5x_3}{3}\right\}$$

#### 1.4.1 Least-square solution

```
import numpy as np
from scipy import linalg as linalg

x = np.array([0, 1, 2, 3])
y = np.array([-1, 0.2, 0.9, 2.1])

A = np.vstack([x, np.ones(len(x))])
B = A.T
print(np.ones(len(x)))
print('\n', A)
print('\n', B)

sol, r, rank, sv = la.lstsq(B, y)
print(sol, r, rank, sv)
```

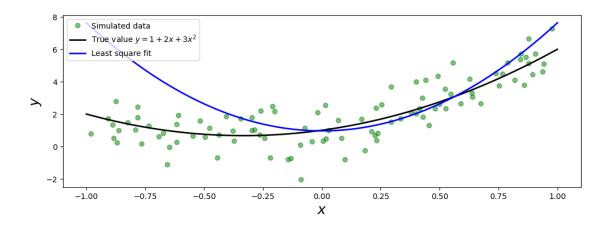
```
[1. 1. 1. 1.]

[[0. 1. 2. 3.]
[1. 1. 1. 1.]]

[[0. 1.]
[1. 1.]
[2. 1.]
[3. 1.]]
[1. -0.95] 0.05 2 [4.10003045 1.09075677]
```

```
[]: import matplotlib.pyplot as plt
     # define true model parameters
     x = np.linspace(-1, 1, 100)
     a, b, c = 1, 2, 3
     y_exact = a + b * x + c * x**2
     # simulate noisy data
     m = 100
     X = 1 - 2 * np.random.rand(m)
     Y = a + b * X + c * X**2 + np.random.randn(m)
     # fit the data to the model using linear legst square
     A = np.vstack([X**0, X**1, X**2])
     sol, r, rank, sv = la.lstsq(A.T, Y)
     y_fit = sol[0] + sol[1] * sol[2] * x**2
     fig, ax = plt.subplots(figsize=(12, 4))
     ax.plot(X, Y, 'go', alpha = 0.5, label = 'Simulated data')
     ax.plot(x, y_exact, 'k', lw=2, label = 'True value $y = 1 + 2x + 3x^2$')
     ax.plot(x, y_fit, 'b', lw=2, label ='Least square fit')
     ax.set_xlabel(r"$x$", fontsize=18)
     ax.set_ylabel(r"$y$", fontsize=18)
     ax.legend(loc=2)
[]: [<matplotlib.lines.Line2D at 0x1663b9820>]
```

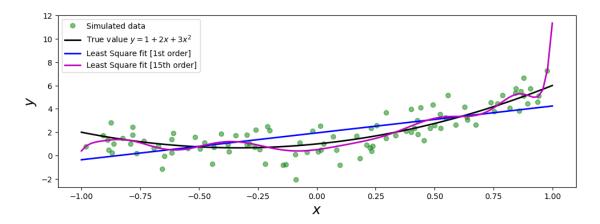
- []: [<matplotlib.lines.Line2D at 0x1663b9a60>]
- []: [<matplotlib.lines.Line2D at 0x1663b9c70>]
- []: Text(0.5, 0, '\$x\$')
- []: Text(0, 0.5, '\$y\$')
- []: <matplotlib.legend.Legend at 0x1663b9c10>



```
[]: # fit the data to the model using linear least square:
     # ist order polynomial
     A = np.vstack([X**n for n in range(2)])
     sol, r, rank, sv = la.lstsq(A.T, Y)
     y_fit1 = sum([s * x**n for n, s in enumerate(sol)])
     # 15th order polynomial
     A = np.vstack([X**n for n in range(16)])
     sol, r, rank, sv = la.lstsq(A.T, Y)
     y_fit15 = sum([s * x**n for n, s in enumerate(sol)])
     fig, ax = plt.subplots(figsize=(12, 4))
     ax.plot(X, Y, 'go', alpha = 0.5, label='Simulated data')
     ax.plot(x, y_exact, 'k', lw=2, label='True value y = 1 + 2x + 3x^2')
     ax.plot(x, y_fit1, 'b', lw=2, label='Least Square fit [1st order]')
     ax.plot(x, y_fit15, 'm', lw=2, label='Least Square fit [15th order]')
     ax.set_xlabel(r"$x$", fontsize=18)
     ax.set_ylabel(r"$y$", fontsize=18)
     ax.legend(loc=2)
```

- []: [<matplotlib.lines.Line2D at 0x166629850>]
- []: [<matplotlib.lines.Line2D at 0x166627910>]
- []: [<matplotlib.lines.Line2D at 0x166629a90>]
- []: [<matplotlib.lines.Line2D at 0x166629ee0>]
- []: Text(0.5, 0, '\$x\$')
- []: Text(0, 0.5, '\$y\$')

#### []: <matplotlib.legend.Legend at 0x1666238b0>



#### 1.5 Eigenvalue Problems

```
 []: \begin{tabular}{ll} eps, delta = sympy.symbols("epsilon, Delta") \\ H = sympy.Matrix([[eps, delta], [delta, -eps]]) \\ H \\ H.eigenvals() \\ H.eigenvects() \\ []: \begin{tabular}{ll} & \epsilon & \Delta \\ \Delta & -\epsilon \\ \\ \end{bmatrix}: \begin{tabular}{ll} & \epsilon & \Delta \\ \Delta & -\epsilon \\ \\ \end{bmatrix}: \begin{tabular}{ll} & \epsilon & \Delta \\ \Delta & -\epsilon \\ \\ \end{bmatrix}: \begin{tabular}{ll} & \epsilon & \Delta \\ \Delta & -\epsilon \\ \\ \end{bmatrix}: \begin{tabular}{ll} & \epsilon & \Delta \\ \Delta & -\epsilon \\ \\ \end{bmatrix}: \begin{tabular}{ll} & \epsilon & \Delta \\ \Delta & -\epsilon \\ \\ \end{bmatrix}: \begin{tabular}{ll} & \epsilon & \Delta \\ \Delta & -\epsilon \\ \\ \end{bmatrix}: \begin{tabular}{ll} & \epsilon & \Delta \\ \Delta & -\epsilon \\ \\ \end{bmatrix}: \begin{tabular}{ll} & \epsilon & \Delta \\ \Delta & -\epsilon \\ \\ \end{bmatrix}: \begin{tabular}{ll} & \epsilon & \Delta \\ \Delta & -\epsilon \\ \\ \end{bmatrix}: \begin{tabular}{ll} & \epsilon & \Delta \\ \Delta & -\epsilon \\ \\ \end{bmatrix}: \begin{tabular}{ll} & \epsilon & \Delta \\ \Delta & -\epsilon \\ \\ \end{bmatrix}: \begin{tabular}{ll} & \epsilon & \Delta \\ \Delta & -\epsilon \\ \\ \end{bmatrix}: \begin{tabular}{ll} & \epsilon & \Delta \\ \Delta & -\epsilon \\ \\ \end{bmatrix}: \begin{tabular}{ll} & \epsilon & \Delta \\ \Delta & -\epsilon \\ \\ \end{bmatrix}: \begin{tabular}{ll} & \epsilon & \Delta \\ \Delta & -\epsilon \\ \\ \end{bmatrix}: \begin{tabular}{ll} & \epsilon & \Delta \\ \Delta & -\epsilon \\ \\ \end{bmatrix}: \begin{tabular}{ll} & \epsilon & \Delta \\ \Delta & -\epsilon \\ \\ \end{bmatrix}: \begin{tabular}{ll} & \epsilon & \Delta \\ \Delta & -\epsilon \\ \\ \end{bmatrix}: \begin{tabular}{ll} & \epsilon & \Delta \\ \Delta & -\epsilon \\ \\ \end{bmatrix}: \begin{tabular}{ll} & \epsilon & \Delta \\ \Delta & -\epsilon \\ \\ \end{bmatrix}: \begin{tabular}{ll} & \epsilon & \Delta \\ \Delta & -\epsilon \\ \\ \end{bmatrix}: \begin{tabular}{ll} & \epsilon & \Delta \\ \Delta & -\epsilon \\ \\ \end{bmatrix}: \begin{tabular}{ll} & \epsilon & \Delta \\ \Delta & -\epsilon \\ \\ \end{bmatrix}: \begin{tabular}{ll} & \epsilon & \Delta \\ \Delta & -\epsilon \\ \\ \end{bmatrix}: \begin{tabular}{ll} & \epsilon & \Delta \\ \Delta & -\epsilon \\ \\ \end{bmatrix}: \begin{tabular}{ll} & \epsilon & \Delta \\ \Delta & -\epsilon \\ \\ \end{bmatrix}: \begin{tabular}{ll} & \epsilon & \Delta \\ \Delta & -\epsilon \\ \\ \end{bmatrix}: \begin{tabular}{ll} & \epsilon & \Delta \\ \end{aligned}: \begin{tabular}{ll} & \epsilon & \Delta \\ \\ \end{bmatrix}: \begin{tabular}{ll} & \epsilon & \Delta \\ \end{aligned}: \begin{tabular}{ll} & \epsilon & \Delta \\ \end{aligned}: \begin{tabular}{ll} & \epsilon &
```

[]: array([-1.75902942, 3.40592034, 13.35310908])

Reference \* Title: Physics Programming Lecture Note (INU) \* Author: Jeongwoo Kim, Ph.D. \* Availability: https://sites.google.com/view/jeongwookim