HW3

July 14, 2019

1 HW3

author: Luis Perez email: luis0@stanford.edu

2 Imports

```
[1]: import numpy as np
import pprint
from scipy import linalg
import seaborn as sns
```

2.1 Problem 1: Memory of a linear, time-invariant system

2.1.1 Part (b)

```
[2]: # Number of samples
    T = 100
    # The input
    u = np.array([
        0.7700,
        0.0600,
        0.0100,
        0.8000,
        0.6600,
        0.9600,
        0.0300,
        0.3200,
        0.9700,
        0.9400,
        0.3600,
        0.4100,
        0.7000,
        0.1700,
        0.1500,
        0.5000,
```

```
0.9600,
0.7300,
0.0800,
0.9000,
0.6800,
0.1100,
0.8500,
0.1900,
0.1600,
0.8900,
0.0700,
0.8300,
0.2600,
0.9700,
0.9100,
0.3800,
0.4100,
0.7000,
0.3200,
0.7200,
0.8600,
0.1700,
0.2600,
0.1900,
0.9100,
0.9800,
0.7900,
0.3300,
0.4300,
0.8200,
0.3800,
0.5300,
0.9000,
0.8300,
0.7900,
0.3600,
0.0300,
0.1700,
0.8500,
0.6800,
0.2100,
0.6400,
0.5000,
0.5100,
0.6000,
0.1900,
0.7400,
```

```
0.9800,
    0.2100,
    0.1900,
    0.3300,
    0.1900,
    0.3900,
    0.5900,
    0.8900,
    0.8500,
    0.5200,
    0.3700,
    0.9600,
    0.8400,
    0.4900,
    0.5400,
    0.8700,
    0.0500,
    0.0500,
    0.3800,
    0.4000,
    0.7800,
    0.4500,
    0.4100,
    0.5300,
    0.8900,
    0.9800,
    0.9300,
    0.1900,
    0.9900,
    0.5800,
    0.6500,
    0.2600,
    0.0900,
    0.0400,
    0.1700,
    0.6800,
    0.3900
])
# The output.
y = np.array([
    1.9663,
    2.5611,
    2.0898,
    1.2280,
    1.2092,
    1.4403,
```

```
1.6056,
1.0081,
1.2782,
2.4515,
2.7882,
1.8628,
1.5105,
2.1557,
2.4086,
1.9909,
1.7859,
2.0328,
1.8014,
1.0792,
1.6591,
2.4087,
2.1433,
2.1237,
1.8788,
1.7544,
2.0878,
1.5912,
1.6593,
1.4437,
1.8176,
2.1338,
1.8734,
1.8083,
2.4560,
2.4541,
2.3812,
2.1748,
1.7084,
1.6541,
1.8287,
2.3495,
2.2745,
1.6575,
1.2851,
1.9121,
2.8243,
2.6995,
2.1301,
2.1108,
2.2646,
2.2775,
2.0367,
```

```
1.9813,
2.3103,
2.6761,
2.1503,
1.1300,
1.1453,
1.7349,
2.0853,
2.0011,
1.6036,
1.9028,
2.3020,
1.7725,
1.4568,
1.7929,
2.0215,
1.8469,
1.4356,
1.3920,
1.5152,
1.4999,
1.7441,
2.6239,
2.9211,
2.4077,
2.0847,
2.5268,
2.3471,
1.9722,
1.9461,
1.9477,
1.7484,
1.1384,
0.9787,
1.5102,
2.1821,
2.3770,
2.2443,
1.8250,
2.6027,
3.0107,
2.8894,
2.2339,
1.9132,
1.8799,
1.8463,
1.6423
```

```
])
    assert y.shape[0] == u.shape[0]
    assert u.shape[0] == T
[3]: def isValidMemory(M, u, y, eps=1e-4):
        """Uses the process described in the homework to check if
        M is a valid memory value for the given inputs u and outputs y.
        # Creates the A matrix described in the handout.
        T = y.shape[0]
        A = linalg.toeplitz(c=u[M-1:T-1], r=np.flip(u[:M]))
        assert u.shape[0] == T
        assert A.shape == (T-M, M)
        # Find the closests hs that work.
        ybar = y[M:]
        hbar = np.dot(np.linalg.inv(np.dot(A.T, A)), np.dot(A.T, ybar))
        return np.linalg.norm(ybar - np.dot(A, hbar)) < eps, hbar</pre>
[4]: def findSmallestValidM(u,y):
        T = u.shape[0]
        for m in range(1, T // 2):
            isValid, candidate = isValidMemory(m,u,y)
            import pdb
            pdb.set_trace
            if isValid:
                return m, candidate
        assert False
[5]: smallestM, weights = findSmallestValidM(u,y)
    print("The memory of our given data is %s." % smallestM)
    pprint.pprint(weights)
   The memory of our given data is 7.
   array([0.63, 0.27, 0.02, 0.37, 0.96, 0.95, 0.46])
   2.2 Problem 3: Sensor integrity monitor
[6]: # Copied from homework assignment.
    A = np.array([
        [1, 2, 3],
        [1, -1, -2],
        [-2, 1, 3],
        [1, -1, -2],
        [1, 1, 0]
    ])
[7]: def findBMatrix(A, eps=1e-3):
```

B = linalg.null_space(A.T).T

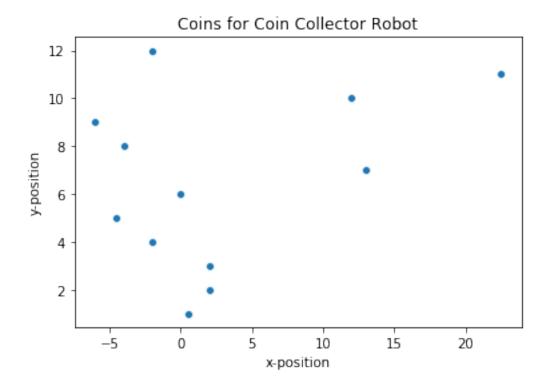
```
B[np.abs(B) < eps] = 0
         B \neq np.abs(B[B != 0]).min()
         return B
 [8]: B = findBMatrix(A)
 [9]: def checkMatrixIsIntegrityMonitor(measureMatrix, integrityMonitor, eps=1e-3):
         M, N = measureMatrix.shape
         # Generate all unit vectors in input space.
         inputBasis = np.identity(N)
         # Assert all consistent values will output zero by monitor.
         for i in range(N):
             measurement = np.dot(measureMatrix, inputBasis[:,i])
             assert np.linalg.norm(np.dot(integrityMonitor, measurement)) < eps</pre>
             # Inconsistent by adding random noise (high likelihood)
             badMeasurement = measurement + np.random.rand(M)
             assert np.linalg.norm(np.dot(integrityMonitor, badMeasurement)) > eps
[10]: checkMatrixIsIntegrityMonitor(A, B)
[11]: B
[11]: array([[ 0.
                        , -5.47259193, 2. , 8.47259193, 1.
            [ 0.
                           6.97831636, 6.39848918, 2.61941742, 3.19924459]])
    2.3 Problem 4: Coin collector robot
    2.3.1 Part (c)
[12]: n=6;
     x = [
         0.5000,
         2.0000,
         2.0000,
        -2.0000,
        -4.5000,
       13.0000,
        -4.0000,
        -6.0000,
        12.0000,
```

[13]: # Plot of the coins.
ax = sns.scatterplot(x=x, y=range(1,2*n + 1))
ax.set_title("Coins for Coin Collector Robot")
ax.set(xlabel='x-position', ylabel='y-position')

22.5000, -2.0000]

[13]: [Text(0, 0.5, 'y-position'), Text(0.5, 0, 'x-position')]

[14]: def generateAMatrix(n, missingRowIndex=None):



```
Generates A as specified in the handout with
        missingRowIndex missing (if set).
        def generateCRow(t):
            c = np.zeros(n)
            for i in range(t // 2):
                    c[i] = 2 + 2*(t - 2*(i+1))
            # If t is odd, we need to add the final push.
            if t % 2 != 0:
                c[t // 2] = 0.5
            return c
        return np.stack([generateCRow(i+1) for i in range(2*n)
                         if missingRowIndex is None or missingRowIndex != i])
[15]: def isCollectible(x, n, missingRowIndex=None, eps=1e-3):
         {\scriptstyle \hookrightarrow } collectible """
        A = generateAMatrix(n, missingRowIndex)
        fhat = np.dot(np.linalg.inv(np.dot(A.T, A)), np.dot(A.T,x))
        if linalg.norm(np.dot(A, fhat) - x) < eps:</pre>
```

```
return fhat
return None

[16]: if isCollectible(x, n):
    print("The coins can be collected.")

else:
    print("The coins cannot be collected.")
```

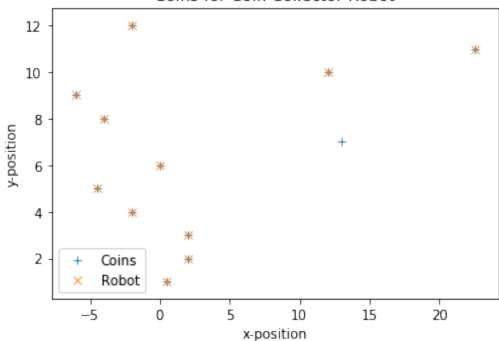
The coins cannot be collected.

```
[17]: def collectAllButOne(x, n):
         """Tries to collect all coins except 1.
         Returns the forces and the index of the bad coin.
         indexes = []
         forces = []
         coins = []
         for i in range(2*n):
             xMissing = np.delete(x, i)
             maybeFhat = isCollectible(xMissing, n, missingRowIndex=i)
             if maybeFhat is not None:
                 indexes.append(i)
                 forces.append(maybeFhat)
                 coins.append((x[i], i+1))
         assert len(indexes) == 1
         assert len(forces) == 1
         assert len(coins) == 1
         return indexes[0], coins[0], forces[0],
[18]: badCoinIndex, badCoin, forcesToCollectAllButOne = collectAllButOne(x, n)
[19]: print("The %s-th coin (1-indexed) is the coin which cannot be collected." %
      →(badCoinIndex + 1))
     print("The bad coin is located at (%s,%s)." % (badCoin))
```

The 7-th coin (1-indexed) is the coin which cannot be collected. The bad coin is located at (13.0,7).

```
ax.set_title("Coins for Coin Collector Robot")
ax.set(xlabel='x-position', ylabel='y-position')
ax.get_figure().savefig("robot_collector")
```





2.4 Problem 6: Quadratic Extrapolation of Time Series using Least-Squares Fit

2.4.1 Part (a)

```
def computeCMatrix(t):
    """
    For a given value of t, computes the C matrix
    as defined in the handout.
    """

PAST_SAMPLES = 10
    F1 = [1] * PAST_SAMPLES
    F2 = [t - i for i in range(PAST_SAMPLES)]
    F3 = [(t - i)**2 for i in range(PAST_SAMPLES)]
    F = np.stack((F1, F2, F3), axis=1)
    t = np.array([1, t + 1, (t + 1)**2])
    C = np.dot(np.dot(t, np.linalg.inv(np.dot(F.T, F))), F.T)
    return C

[24]: def verifiyCIsTimeInvariant(eps=1e-4):
    for t in range(1000):
```

```
C1 = computeCMatrix(t + 1)
             C2 = computeCMatrix(t + 2)
             if not linalg.norm(C1 - C2) < eps:</pre>
                 print(linalg.norm(C1 - C2))
         return computeCMatrix(1)
[25]: computeCMatrix(1)
[25]: array([ 0.9
                       , 0.5
                                    , 0.18333333, -0.05
                                                                            ])
                                             , 0.03333333, 0.3
            -0.26666667, -0.25
                                     , -0.15
[26]: computeCMatrix(1000)
[26]: array([ 0.90000038,  0.4999995 ,  0.18333217, -0.0500016 , -0.20000182,
             -0.26666848, -0.2500016 , -0.15000115, 0.03333284, 0.30000039]) \\
    2.4.2 Part (b)
[27]: def generateTrueTimeSeries():
         # Generates the vector [1, ..., 1000]
         t = np.array(range(1, 1001))
         z = 5 * np.sin(t / 10 + 2) + 0.1 * np.sin(t) + 0.1 * np.sin(2*t - 5)
         return z
[28]: def computeExtrapolationFromValues(values):
         assert len(values) == 10
         c = computeCMatrix(1)
         for _ in range(11, 1001):
             newValue = np.dot(c, np.flip(values[len(values)-10:]))
             values = np.append(values, newValue)
         return values
     def computeEstimatesUsingTrueValues(values):
         c = computeCMatrix(1)
         estimates = values[:10].tolist()
         for i in range(10, 1000):
             estimates.append(np.dot(c, np.flip(values[i-10:i])))
         return np.array(estimates)
[29]: def getRMSE(true, predictions):
         return np.sqrt(np.sum((true[10:] - predictions[10:])**2) / np.sum(true[10:
      →]**2))
[30]: true = generateTrueTimeSeries()
     predsUsingOnlyInitial = computeExtrapolationFromValues(true[:10])
     predsUsingRealValues = computeEstimatesUsingTrueValues(true)
     extrapolateRMSE = getRMSE(true, predsUsingOnlyInitial)
     cheatRMSE = getRMSE(true, predsUsingRealValues)
[31]: cheatRMSE
```

[31]: 0.050960852617736675

```
[32]: MAX = 100

ax = sns.scatterplot(x=range(1,MAX+1), y=true[:MAX], color='blue')

ax = sns.scatterplot(x=range(1,MAX+1), y=predsUsingRealValues[:MAX],

color='red')

ax.set_title("Least Squares Fit on Ten Samples")

ax.set(xlabel='Time (t)', ylabel='Series Value (z)')

ax.get_figure().savefig('least_squares_fit_series')
```



