

HW 12

1) With Friends:

James Zhu (3031793129), Ilya (3031806896).
I worked alone at first, then met Ilya & James.

2) a) Use Gram-Schmidt to find a matrix V whose columns form an orthonormal basis for the column space of V .

$$V = \begin{bmatrix} 1 & 1 & 1 \\ 0 & 1 & 1 \\ 0 & 0 & 1 \\ 0 & 0 & 1 \end{bmatrix}$$

We orthonormalize the matrix:

$$u_1 = v_1$$

$$e_1 = \frac{u_1}{\|u_1\|} = \begin{bmatrix} 1 \\ 0 \\ 0 \\ 0 \end{bmatrix}$$

$$u_2 = v_2 - \frac{\langle v_2, u_1 \rangle}{\langle u_1, u_1 \rangle} u_1$$

$$e_2 = \frac{u_2}{\|u_2\|} = \begin{bmatrix} 0 \\ 1/\sqrt{2} \\ 1/\sqrt{2} \\ 0 \end{bmatrix}$$

$$u_3 = v_3 - \frac{\langle v_3, u_1 \rangle}{\langle u_1, u_1 \rangle} u_1 - \frac{\langle v_3, u_2 \rangle}{\langle u_2, u_2 \rangle} u_2$$

$$e_3 = \frac{u_3}{\|u_3\|} = \begin{bmatrix} 0 \\ 0 \\ 1/\sqrt{2} \\ 1/\sqrt{2} \end{bmatrix}$$

Now, show that

$$V(V^T V)^{-1} V^T w = V(U^T U)^{-1} U^T w, \quad w = [1, -1, 0, -1, 0]^T$$

$$V^T w = \begin{bmatrix} 1 \\ 0 \\ -1 \end{bmatrix}, \quad V^T V = \begin{bmatrix} 1 & 1 & 1 \\ 1 & 3 & 3 \\ 1 & 3 & 5 \end{bmatrix} \rightarrow (V^T V)^{-1} = \begin{bmatrix} 3/2 & -1/2 & 0 \\ -1/2 & 1 & -1/2 \\ 0 & -1/2 & 1/2 \end{bmatrix}$$

$$(V^T V)^{-1} V^T w = \begin{bmatrix} 3/2 \\ 0 \\ -1/2 \end{bmatrix}, \quad V \begin{bmatrix} 3/2 \\ 0 \\ -1/2 \end{bmatrix} = \begin{bmatrix} 1 \\ -1/2 \\ -1/2 \\ -1/2 \end{bmatrix}$$

$$U^T w = \begin{bmatrix} 1 \\ -1 \end{bmatrix}, \quad U^T U = \begin{bmatrix} 1 & 0 & 0 \\ 0 & 2 & 0 \\ 0 & 0 & 2 \end{bmatrix} \rightarrow (U^T U)^{-1} = \begin{bmatrix} 1 & 0 & 0 \\ 0 & 1/2 & 0 \\ 0 & 0 & 1/2 \end{bmatrix}$$

$$(U^T U)^{-1} U^T w = \begin{bmatrix} 1 \\ -1/2 \\ -1/2 \end{bmatrix}, \quad U \begin{bmatrix} 1 \\ -1/2 \\ -1/2 \end{bmatrix} = \begin{bmatrix} 1 \\ -1/2 \\ -1/2 \\ -1/2 \end{bmatrix}$$

$$\therefore \boxed{V(V^T V)^{-1} V^T w = V(U^T U)^{-1} U^T w} = [1, -1/2, -1/2, -1/2]^T$$

b) Orthonormalise matrix V :

$$u_1 = v_1 = \begin{bmatrix} 1 \\ 0 \\ 0 \end{bmatrix}$$

$$u_2 = v_2 - \frac{\langle v_2, u_1 \rangle}{\langle u_1, u_1 \rangle} u_1 = \begin{bmatrix} 1 \\ 0 \\ -1 \end{bmatrix}$$

$$u_3 = v_3 - \frac{\langle v_3, u_1 \rangle}{\langle u_1, u_1 \rangle} u_1 - \frac{\langle v_3, u_2 \rangle}{\langle u_2, u_2 \rangle} u_2 = \begin{bmatrix} 0 \\ 0 \\ 0 \end{bmatrix}$$

Now verify $V(V^T V)^{-1} V^T w = U(U^T U)^{-1} U^T w$:

$$U^T w = \begin{bmatrix} 0 \\ 1 \end{bmatrix}, \quad U^T U = \begin{bmatrix} 3 & 0 \\ 0 & 4 \end{bmatrix}, \quad (U^T U)^{-1} = \begin{bmatrix} 1/3 & 0 \\ 0 & 1/4 \end{bmatrix}$$

$$U(U^T U)^{-1} U^T w = \begin{bmatrix} 1/2 \\ 3/2 \\ -1/2 \end{bmatrix} = V(V^T V)^{-1} V^T w$$

c) 1)
$$V = \begin{bmatrix} 1 & 1 & 1 \\ 0 & 1 & 1 \\ 0 & 0 & 1 \\ 0 & 0 & 1 \end{bmatrix} = \begin{bmatrix} 1 & 0 & 0 \\ 0 & 1/\sqrt{2} & 0 \\ 0 & 1/\sqrt{2} & 0 \\ 0 & 0 & 1/\sqrt{2} \end{bmatrix} \begin{bmatrix} 1 & 1 & 1 \\ 0 & \frac{2}{\sqrt{2}} & \frac{2}{\sqrt{2}} \\ 0 & 0 & \frac{2}{\sqrt{2}} \end{bmatrix}$$

$$= [q_1 \ q_2 \ q_3] \begin{bmatrix} v_{1q_1} & v_{2q_1} & v_{3q_1} \\ 0 & v_{2q_2} & v_{3q_2} \\ 0 & 0 & v_{3q_3} \end{bmatrix} = QR$$

2)
$$V = \begin{bmatrix} 1 & 1 & -1 \\ 0 & 1 & -1 \\ 1 & 0 & 0 \\ 0 & -1 & 1 \end{bmatrix} = \begin{bmatrix} \frac{1}{\sqrt{3}} & \frac{1}{2} \\ 0 & \frac{1}{2} \\ \frac{1}{\sqrt{3}} & 0 \\ \frac{1}{\sqrt{3}} & -\frac{1}{2} \\ 0 & \frac{1}{2} \end{bmatrix} \begin{bmatrix} \frac{3}{\sqrt{3}} & 0 & 0 \\ 0 & 2 & -2 \end{bmatrix} = QR$$

$$3) a) b^k = \ln\left(\frac{\ln(1-p)}{\ln 0.95}\right) + 26.1931$$

$$A_1^k = \ln(AGE^k)$$

$$A_2^k = \ln(TC^k)$$

$$A_3^k = \ln(HDL^k)$$

$$A_4^k = \ln(SBP^k)$$

$$A_5^k = DIA$$

$$A_6^k = SMK$$

$$b^k = x^T A^k$$

$$b = \begin{bmatrix} b_1 \\ \vdots \\ b_n \end{bmatrix}, A = \begin{bmatrix} A_1^1 & A_2^1 & A_3^1 & A_4^1 & A_5^1 & A_6^1 \\ \vdots & \vdots & \vdots & \vdots & \vdots & \vdots \\ A_1^n & A_2^n & A_3^n & A_4^n & A_5^n & A_6^n \end{bmatrix}$$

$$c) \bar{x} = \begin{bmatrix} 2.322 \\ 1.24 \\ -0.668 \\ 2.685 \\ 0.705 \\ 0.513 \end{bmatrix}$$

$$d) E^2 = |b - \hat{b}|^2 = |0.695|$$

e) iPython

f) Linear

g) Perturbation increases error to 77.205

h) iPython

i) iPython

All
also
in
iPython

4) ~~is~~ Python.

5) a) Yes, Yes

b) No, consider $\vec{a} = \begin{bmatrix} 1 \\ 0 \end{bmatrix}$, $\vec{b} = \begin{bmatrix} 0 \\ 1 \end{bmatrix}$,

$$\left. \begin{aligned} \vec{x}_1 &= \vec{a} + \vec{b} = \begin{bmatrix} 1 \\ 1 \end{bmatrix} \\ \vec{x}_2 &= 2\vec{a} + 2\vec{b} = \begin{bmatrix} 2 \\ 2 \end{bmatrix} \end{aligned} \right\} \vec{x}_1 \neq \vec{x}_2.$$

$$\begin{aligned} c) M &= [\vec{x}_1, \vec{x}_2] & V &= \mathbb{Q}^2 \\ &= \begin{bmatrix} 1 & 2 \\ 1 & 2 \end{bmatrix} \end{aligned}$$

$$u_1 = m_1 = \begin{bmatrix} 1 \\ 1 \end{bmatrix}, \quad u_2 = m_1 - \frac{\langle m_2, u_1 \rangle}{\langle u_1, u_1 \rangle} u_1 = \begin{bmatrix} 1 \\ 1 \end{bmatrix} - 2 \begin{bmatrix} 1 \\ 1 \end{bmatrix} = \begin{bmatrix} -1 \\ -1 \end{bmatrix}$$

$$\therefore Q = \begin{bmatrix} 1/\sqrt{2} & -1/\sqrt{2} \\ 1/\sqrt{2} & -1/\sqrt{2} \end{bmatrix}$$

One example of $Q = \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix}$ LOL,

but this is just a 2D example, in the problem of troll sand, the vectors would not necessarily be 2-dimensional, they could be 1000-dimensional. Just something to keep in mind.

d-e) Python.

- 6) Q: Given a set of three vectors $\{\vec{v}_1, \vec{v}_2, \vec{v}_3\}$ that is linearly independent of each other, ~~and all are of unit length~~, find an orthonormal set of vectors $\{\vec{w}_1, \dots, \vec{w}_n\}$.

Solution:

Step 1 Find unit vector \vec{w}_1 such that $\text{span}(\{\vec{w}_1\}) = \text{span}(\{\vec{v}_1\})$.

$$\vec{w}_1 = \frac{\vec{v}_1}{\|\vec{v}_1\|}$$

Step 2 Find \vec{w}_2 s.t. $\text{span}(\{\vec{w}_1, \vec{w}_2\}) = \text{span}(\{\vec{v}_1, \vec{v}_2\})$ and orthogonal to \vec{w}_1 .

$$\vec{e}_2 = \vec{v}_2 - (\vec{v}_2^T \vec{w}_1) \vec{w}_1$$

$$\vec{w}_2 = \frac{\vec{e}_2}{\|\vec{e}_2\|}$$

Step 3 Find \vec{w}_3

$$\vec{e}_3 = \vec{v}_3 - (\vec{v}_3^T \vec{w}_2) \vec{w}_2 - (\vec{v}_3^T \vec{w}_1) \vec{w}_1,$$

$$\vec{w}_3 = \frac{\vec{e}_3}{\|\vec{e}_3\|}$$

And we're done generalizing.

hw12

November 22, 2016

1 Problem Set 11 Code

```
In [144]: %pylab inline
import numpy as np
import matplotlib.pyplot as plt
```

Populating the interactive namespace from numpy and matplotlib

2 The Framingham Risk Score Revisited

2.1 Part a

```
In [145]: # Importing medical data
import scipy.io

# LOADS IN THE MEDICAL DATA IN THE FORM OF A PYTHON DICTIONARY.
# Data credit: CDC http://www.cdc.gov/nchs/nhanes.htm
data = scipy.io.loadmat('CVDdata.mat')

#UNPACKING DATA INTO COLUMN VECTORS
AGE = data['AGE']
TC = data['TC']
HDL = data['HDL']
SBP = data['SBP']
DIA = data['DIABETIC']
SMK = data['SMOKER']
p = data['pNoisy']

# Write expressions for b, A1, A2, A3, A4, A5, A6
# It will help to use the identity  $\log_n(z) = \log(z)/\log(n)$ 

b = np.log(np.log(1 - p)/np.log(0.95)) + 26.1931

A1 = np.log(AGE)
A2 = np.log(TC)
A3 = np.log(HDL)
A4 = np.log(SBP)
A5 = DIA
A6 = SMK
```

2.2 Part b

```
In [146]: # Write expressions for b and A
# the function np.hstack will be helpful for constructing A
```

```
b = np.array(b.ravel())
A = np.hstack((A1,A2,A3,A4,A5,A6))
```

2.3 Part c

In [147]: *# Write an expression for xhat*

```
xhat = np.dot(np.linalg.inv(np.dot(A.T, A)), np.dot(A.T, b))

print("The estimated values for x:" + str(xhat.T))
```

The estimated values for x: [2.34338912 1.2405098 -0.6692395 2.68521474 0.70530453 0.5132915]

2.4 Part d

In [148]: *# the model estimate bhat, and the squared error e2*

```
bhat = np.dot(A, xhat)
e2 = np.linalg.norm(bhat - b)**2

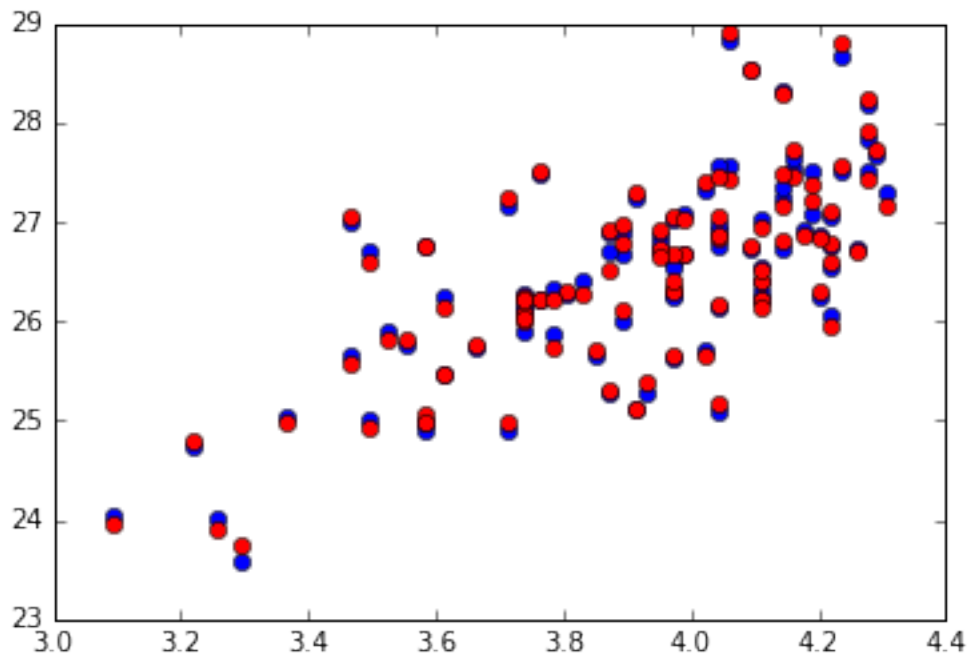
print("E^2 = " + str(e2))
```

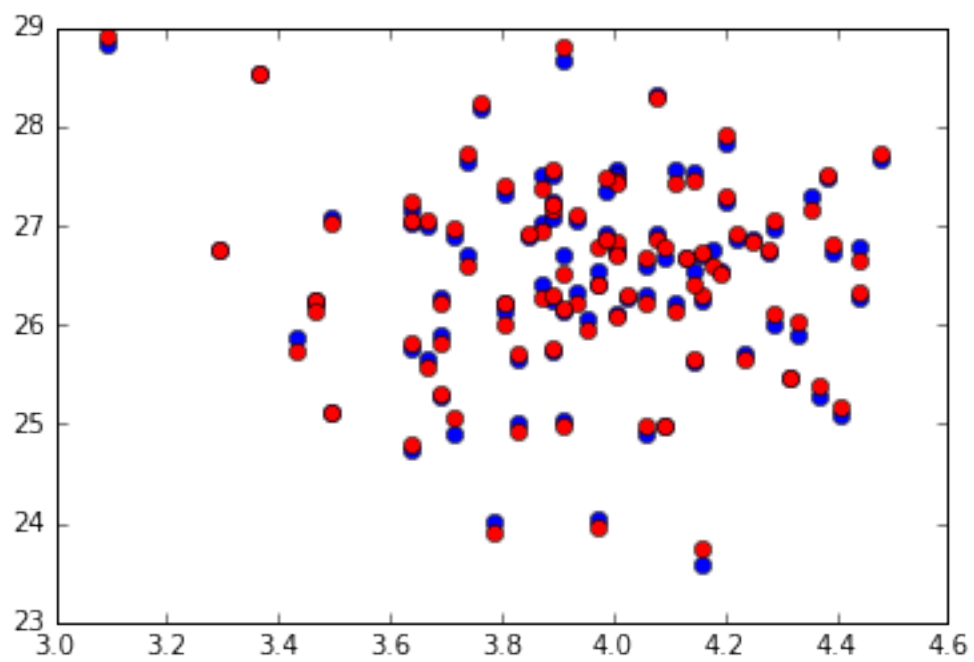
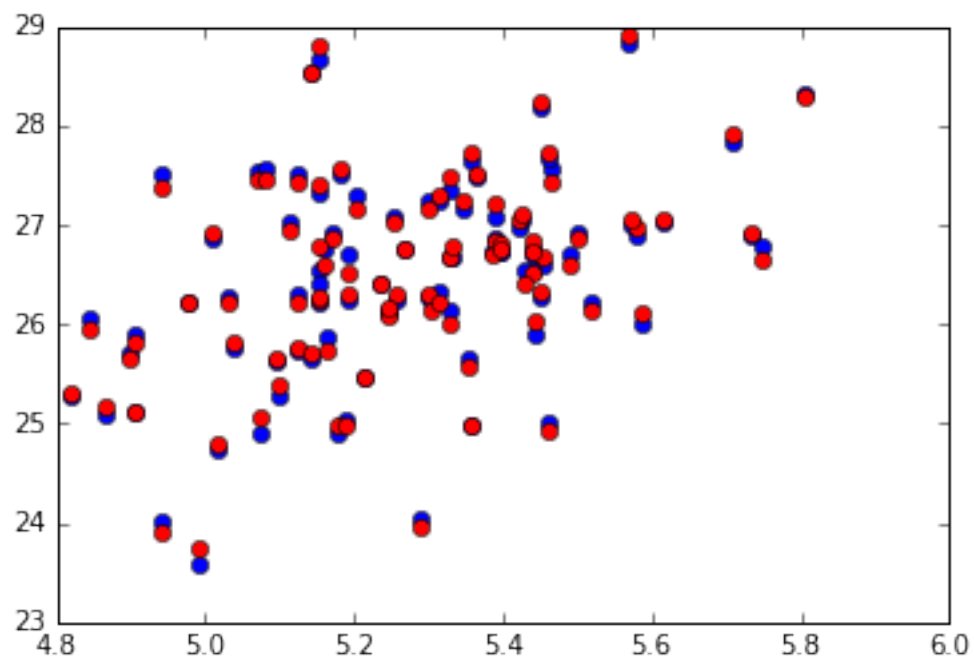
E^2 = 0.695069973457

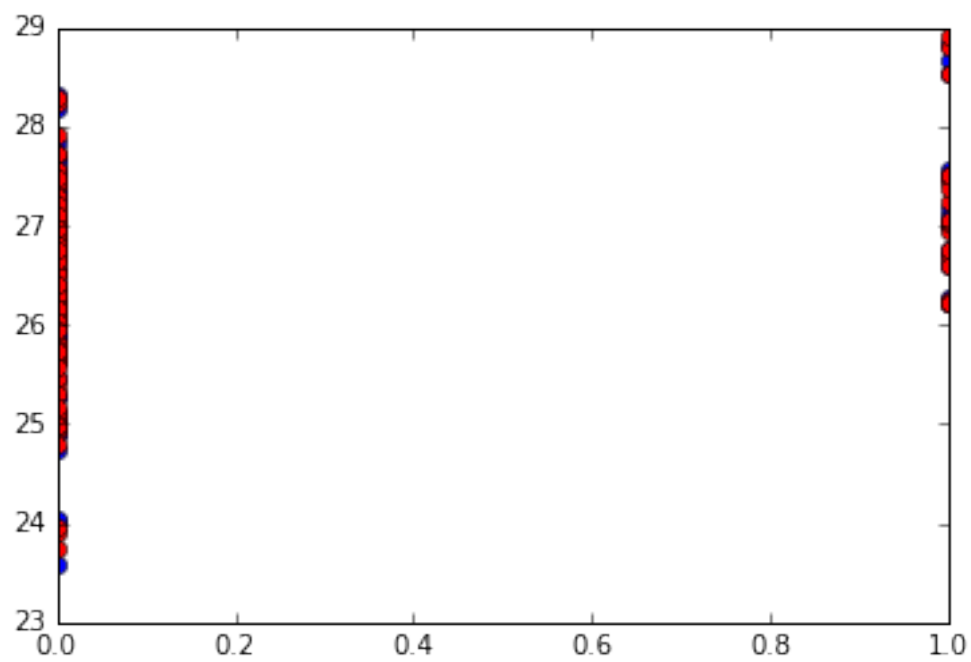
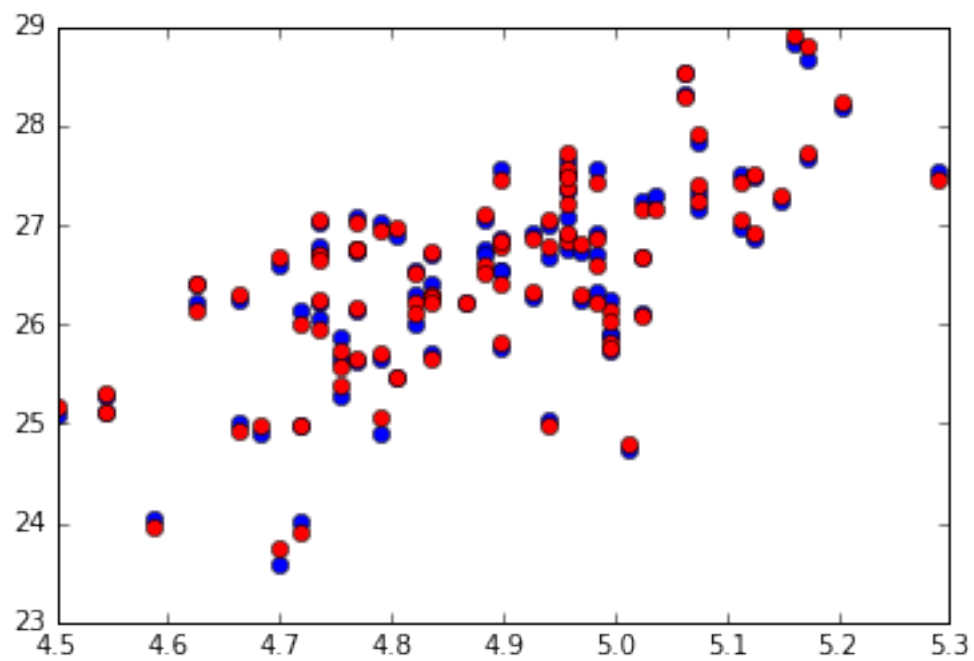
2.5 Part e

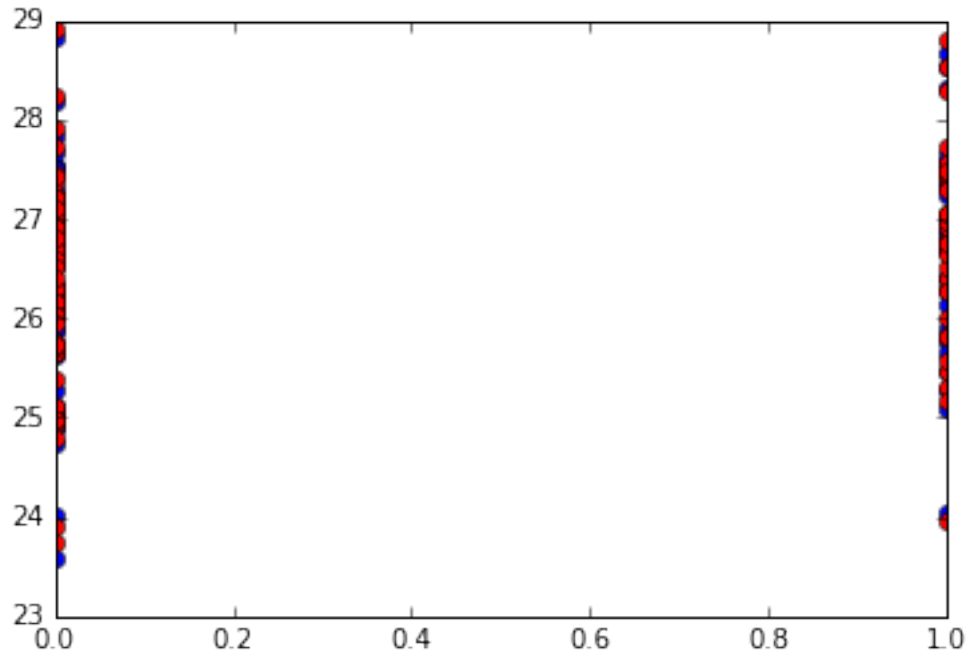
In [151]: *# Linear plots, pick an index below (0,1,2,etc). This code will plot b and bhat vs Ai*

```
for i in range(len(A[0])):
    plt.plot(A[:,i],b,'ob')
    plt.plot(A[:,i],bhat,'or')
plt.show()
```









2.6 Part f

In [152]: *# Here are the values for the test plot*

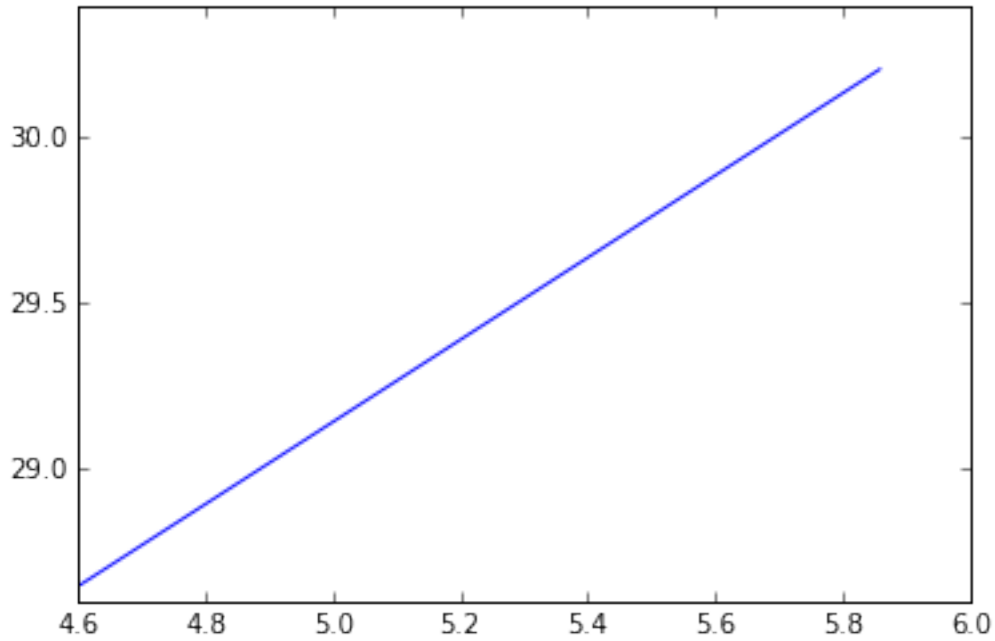
```
age_test = 55
tc_test_vector = np.linspace(100,350,(350-100+1))
hdl_test = 25
sbp_test = 220
dia_test = 1
smk_test = 1
```

```
A2_test = np.zeros(tc_test_vector.size)
b_test = np.zeros(tc_test_vector.size)
```

```
for ind in range(tc_test_vector.size):
    tc_test = tc_test_vector[ind];
    # Use the values for age_test, tc_test, hdl_test, sbp_test, dia_test
    # and smk_test to calculate the next value for b_test (y axis value)
    # and A2_test (x_axis value)
    a = np.array([np.log(age_test), np.log(tc_test), np.log(hdl_test), np.log(sbp_test), dia_
    b_test[ind] = np.dot(a, xhat)
    A2_test[ind] = np.log(tc_test)
```

```
plt.plot(A2_test,b_test,'-b')
```

Out[152]: [



2.7 Part g

In [162]: *# Perturb xhat from the solution above, store into x_perturbed and replot.*

Use the following example expression with different perturbations.

```
x_perturbed = xhat+np.array([.1, 0.2, 0.2, -0.3, 0.1, 0.32])
```

What are the new estimated b values in terms of x_perturbed?

```
b_perturbed = np.dot(A,x_perturbed)
```

Plot again

```
for i in range(len(A[0])):
```

```
    plt.plot(A[:,i],b,'ob')
```

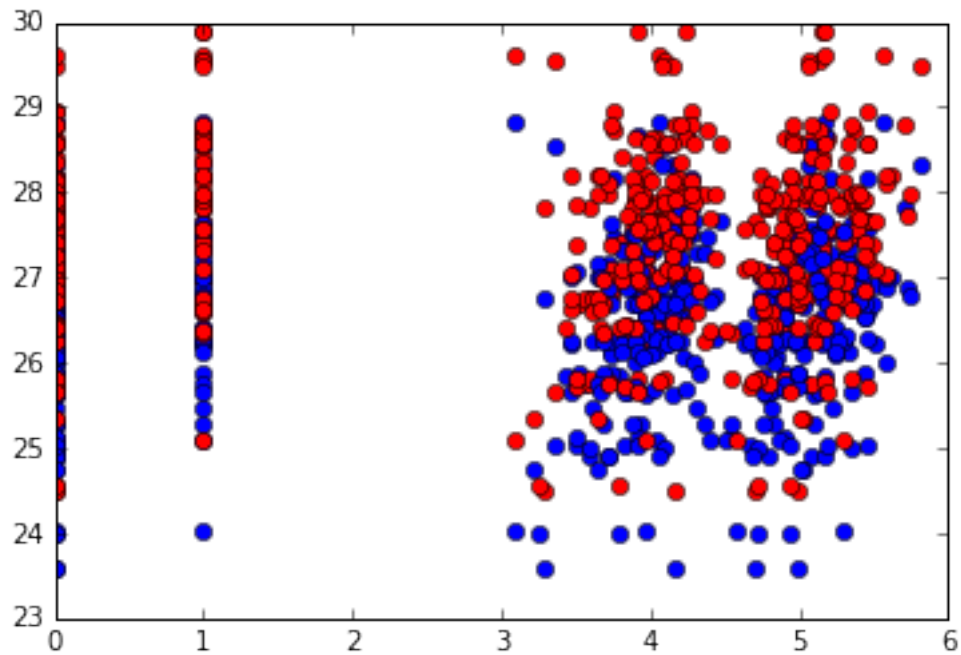
```
    plt.plot(A[:,i],b_perturbed,'or')
```

What is the new sum of squared errors (after perturbing)?

```
e2_perturbed = np.linalg.norm(b_perturbed - b)**2
```

```
print("E^2 after perturbing" + str(e2_perturbed))
```

E² after perturbing:77.3312748655



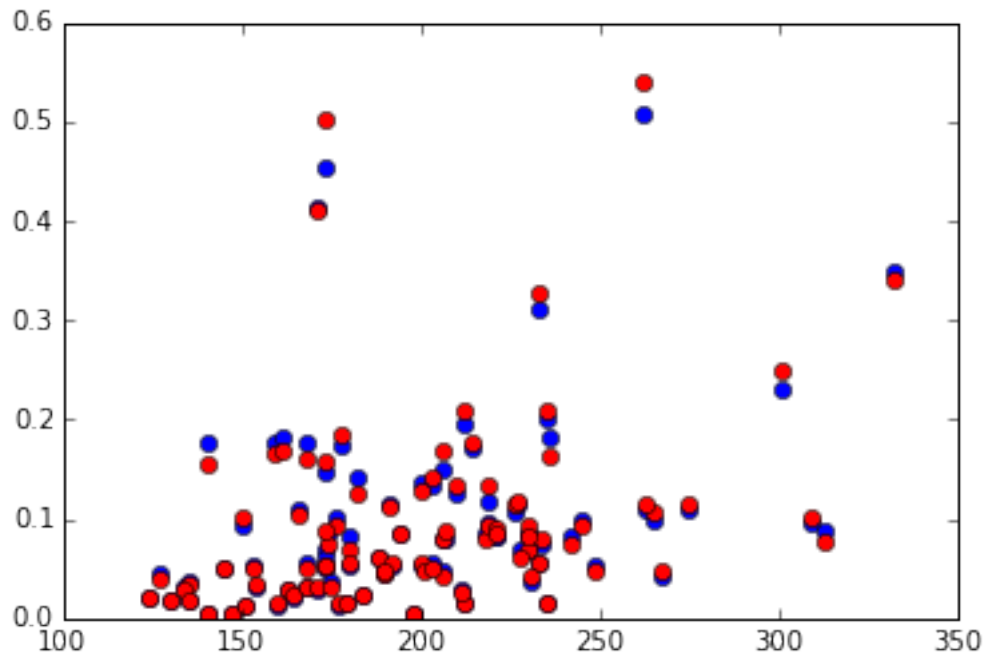
2.8 Part h

In [163]: *# Nonlinear plots, pick an index below (0,1,2,etc). This code will plot b and bhat vs A_i*
i = 1

```
# Write an expression for estimated p values here
p_estimated = 1- np.exp(np.log(0.95)*np.exp(bhat - 26.1931))

plt.plot(np.exp(A[:,i]),p,'ob')
plt.plot(np.exp(A[:,i]),p_estimated,'or')
```

Out[163]: [

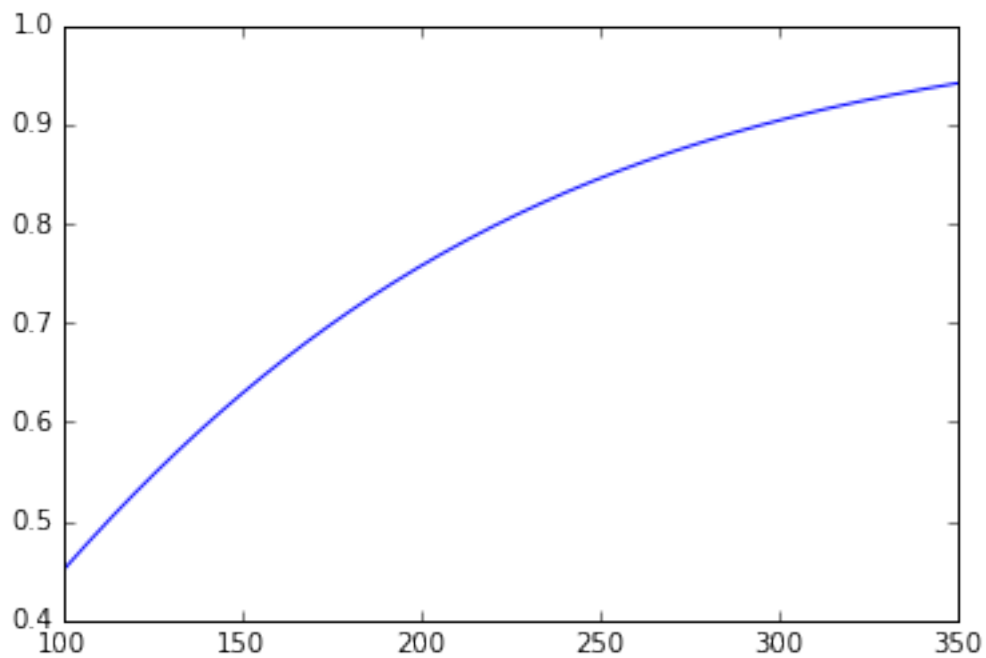


2.9 Part i

```
In [164]: # transform b_test into p_test
p_test = 1- np.exp(np.log(0.95)*np.exp(b_test - 26.1931))

plt.plot(tc_test_vector,p_test,'-b')
```

Out[164]: [<matplotlib.lines.Line2D at 0x106d85358>]



3 Finding Signals in Noise

```
In [165]: # Run this first
%matplotlib inline
import numpy as np
import scipy as sp
import scipy.linalg as la
import pylab as plt
import numpy.random

N = 1000

def rand_vector(n): # returns a random {+1, -1} vector of length n
    return np.random.randint(2, size=n)*2 - 1.0

def rand_normed_vector(n): # returns a random normalized vector of length n
    x = rand_vector(n)
    return x / la.norm(x)

def cross_corr(f, g):
    # returns the cross-correlation (a vector of all the inner-products of 'g' with shifted v
    C = la.circulant(f)
    corr = C.T.dot(g)
    return corr
```

4 (a)

```
In [171]: # generate a random normalized vector for s1
# (running this cell again will generate a new random vector)
s1 = rand_normed_vector(N)

# compute all the inner-products of s1 with shifted versions of s1
# (ie, the cross-correlation of s1 with s1)
corr = cross_corr(s1, s1)

# The inner-product <s1, s1^(1)> is:
print(corr[1])

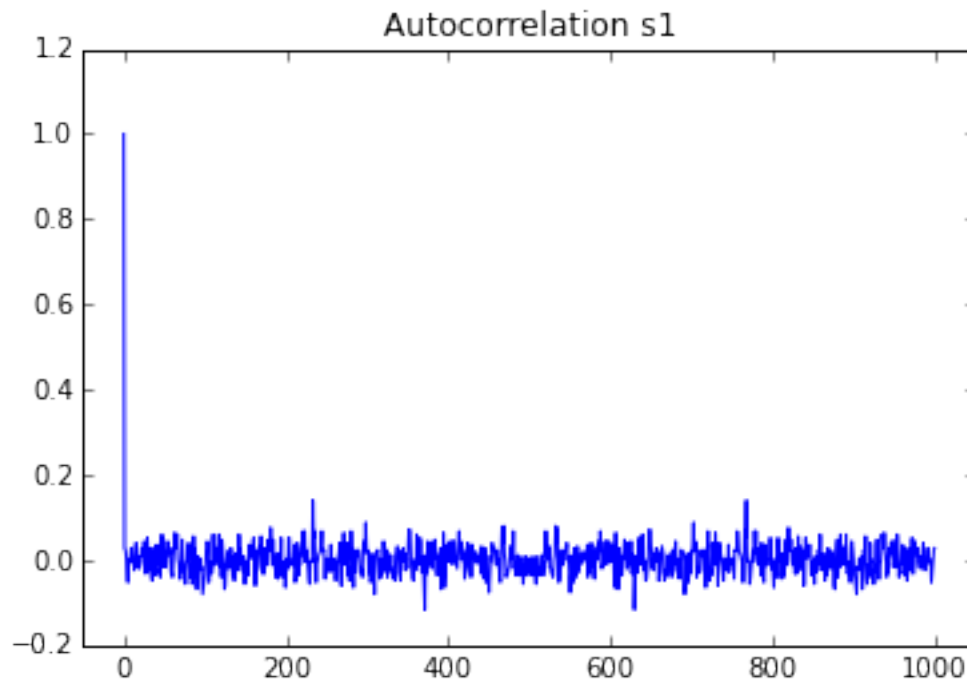
# np.roll circularly shifts the signal
# so the above inner-product could be computed as:
print(np.dot(s1, np.roll(s1,1)))

# Plot the autocorrelation:
plt.title("Autocorrelation s1")
plt.plot(corr)

x1,x2,y1,y2 = plt.axis()
plt.axis([x1-50,x2+50,y1,y2])

plt.show()
```


0.028
0.028



5 (b)

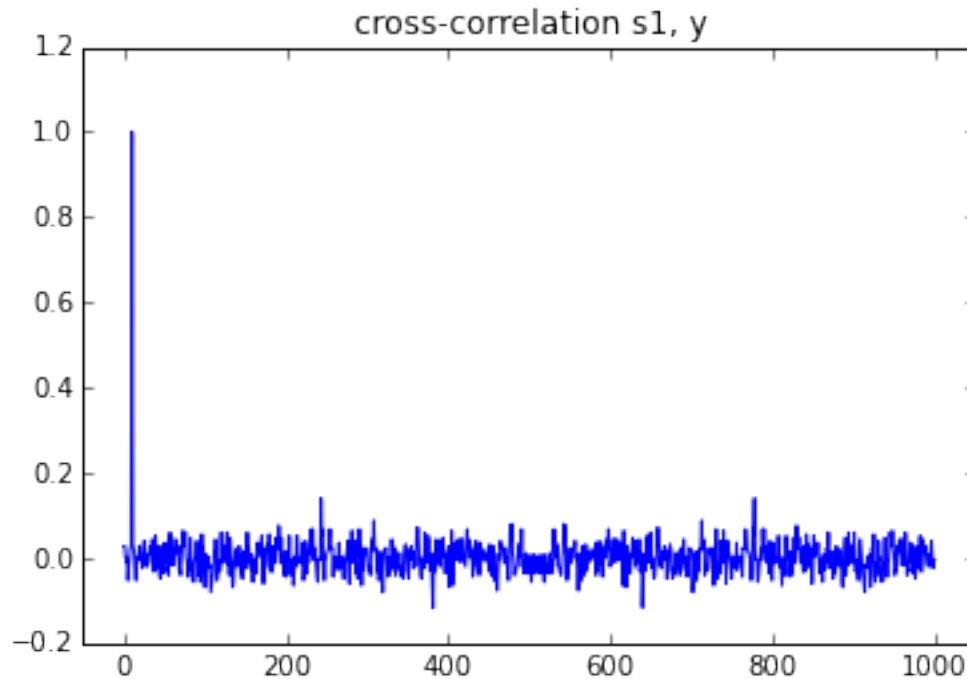
```
In [172]: y = np.roll(s1, 10) # Received y = s1 shifted by 10

# Compute the cross-correlation (all the inner-products of y with shifted versions of s1)
corr = cross_corr(s1, y)

# Plot
plt.title("cross-correlation s1, y")
plt.plot(corr)

x1,x2,y1,y2 = plt.axis()
plt.axis([x1-50,x2+50,y1,y2])
plt.show()

# Find the index of maximum correlation (inner-product)
print(np.argmax(corr))
```



10

6 (c)

```
In [173]: # generate a random normalized vector for s1,
# and a random normalized vector for n
# (running this cell again will generate new random vectors)
s1 = rand_normed_vector(N)
n = rand_normed_vector(N)

print(np.abs(np.dot(s1, n)))
```

0.07

7 (d)

This is the code from part (b), but with the received signal y additionally corrupted by noise

```
In [174]: s1 = rand_normed_vector(N)
n = rand_normed_vector(N)
y = np.roll(s1, 10) + 0.1*n

corr = cross_corr(s1, y)

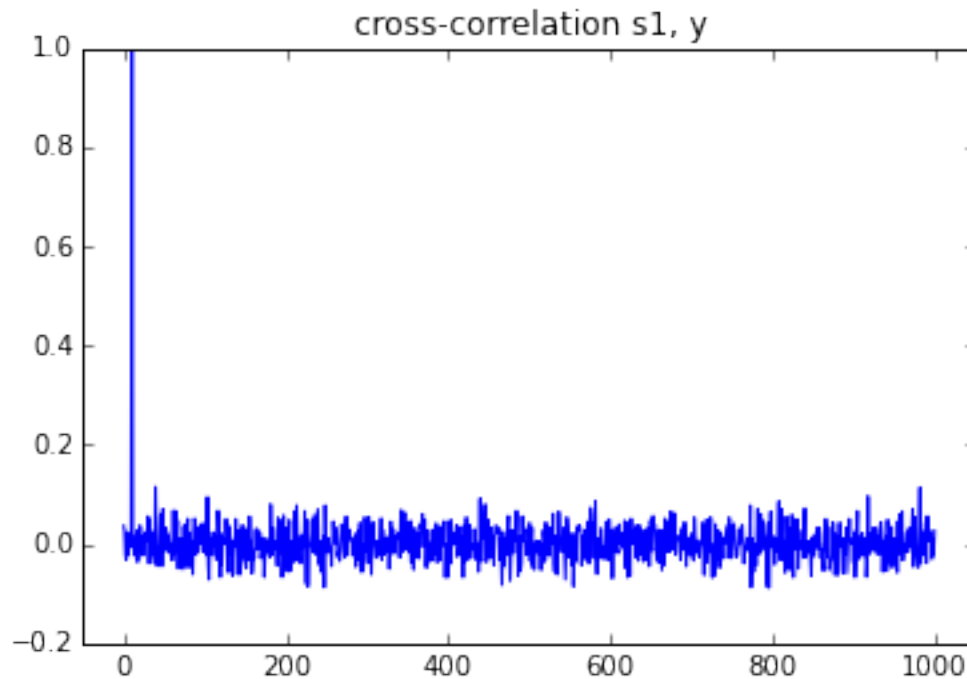
plt.title("cross-correlation s1, y")
plt.plot(corr)
```

```

x1,x2,y1,y2 = plt.axis()
plt.axis([x1-50,x2+50,y1,y2])
plt.show()

# Find the index of maximum correlation (inner-product)
np.argmax(corr)

```



Out[174]: 10

8 (e)

Copy the code provided for part (d), but modify appropriately so the noise is higher. You should generate two cross-correlation plots, one for each noise level in the question. (For example, you can just copy the code from part (d) twice.)

```

In [175]: ## CODE HERE
s1 = rand_normed_vector(N)
n = rand_normed_vector(N)
y = np.roll(s1, 10) + 10*n

corr = cross_corr(s1, y)

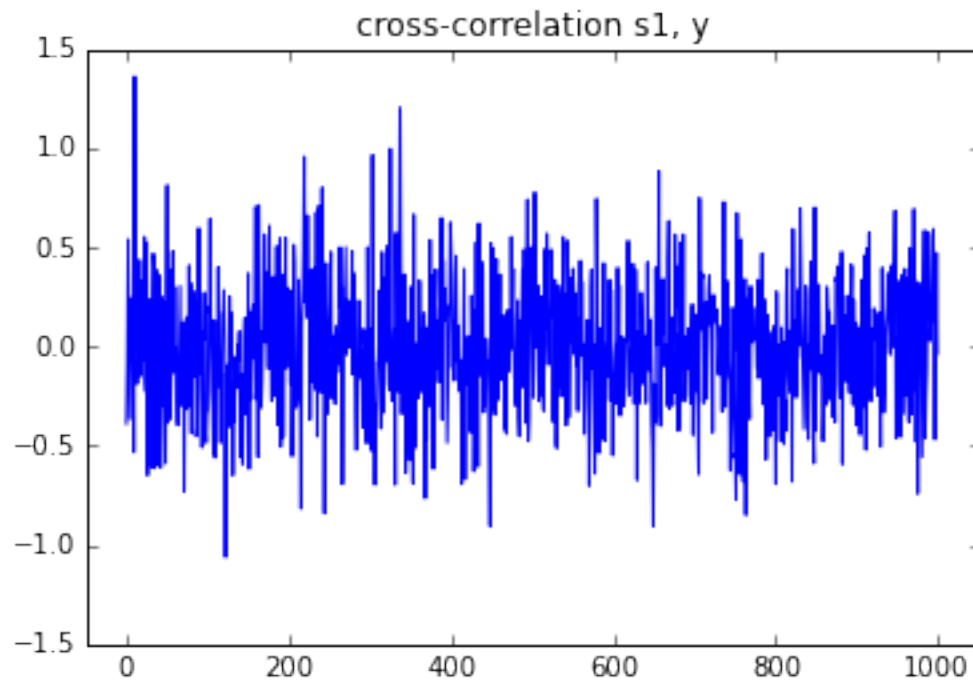
plt.title("cross-correlation s1, y")
plt.plot(corr)

x1,x2,y1,y2 = plt.axis()
plt.axis([x1-50,x2+50,y1,y2])
plt.show()

```



```
# Find the index of maximum correlation (inner-product)
np.argmax(corr)
```



Out[175]: 10

9 (f)

```
In [176]: s1 = rand_normed_vector(N)
          s2 = rand_normed_vector(N)

          y = np.roll(s1, 10) + np.roll(s2, 100)

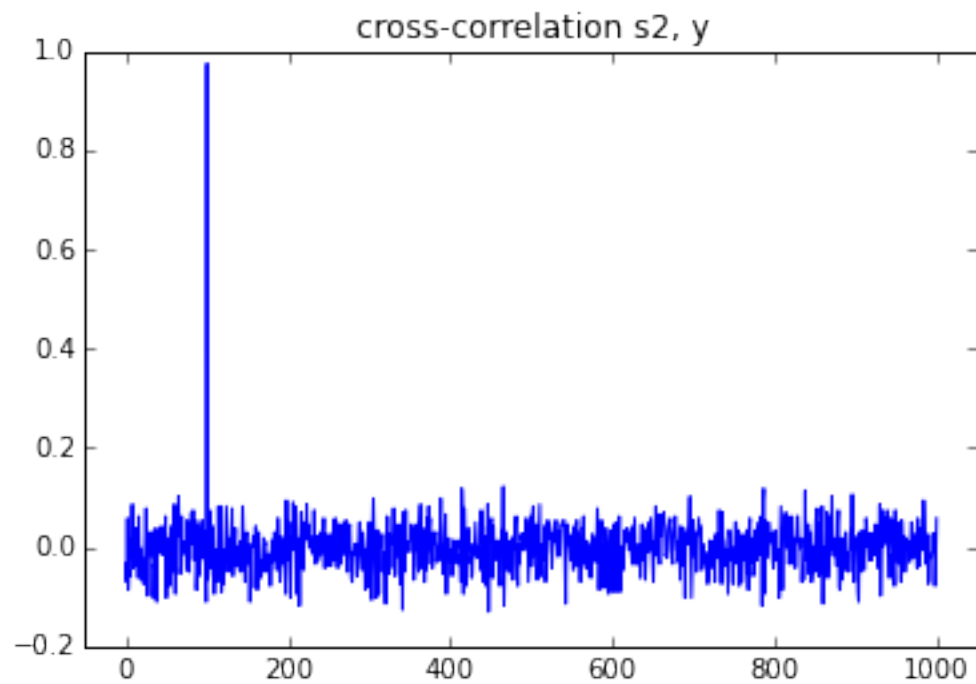
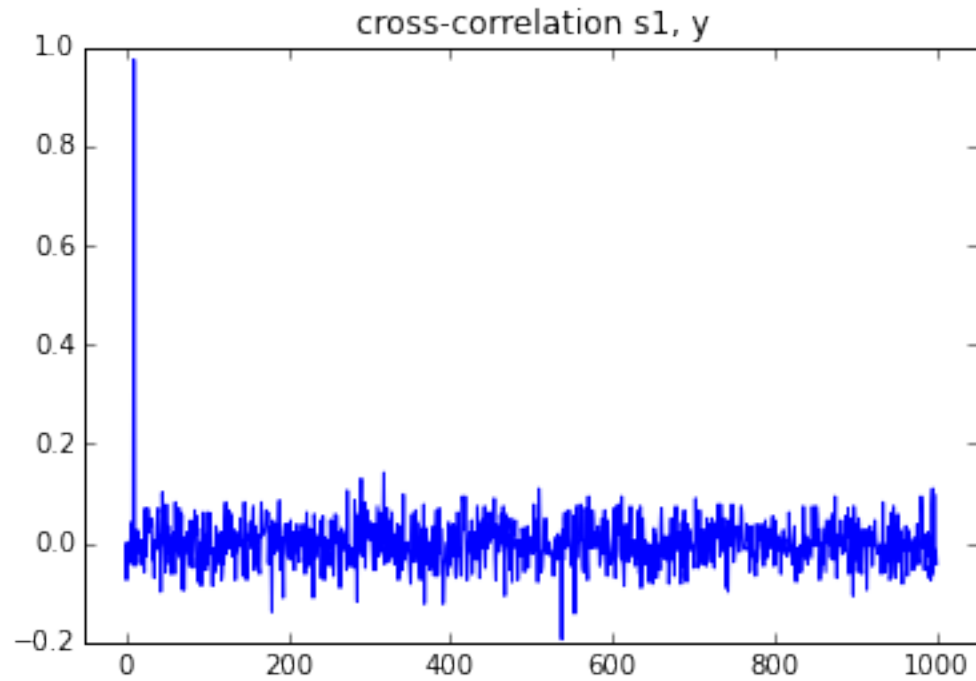
          # Compute cross-correlations:
          corr_s1_y = cross_corr(s1, y)
          corr_s2_y = cross_corr(s2, y)

          # Plot cross-correlations:
          plt.title("cross-correlation s1, y")
          plt.plot(cross_corr(s1, y))
          x1,x2,y1,y2 = plt.axis()
          plt.axis([x1-50,x2+50,y1,y2])
          plt.show()

          plt.title("cross-correlation s2, y")
          plt.plot(cross_corr(s2, y))
          x1,x2,y1,y2 = plt.axis()
          plt.axis([x1-50,x2+50,y1,y2])
```

```
plt.show()

j = np.argmax(corr_s1_y) # find the first signal delay (max index of correlation)
k = np.argmax(corr_s2_y) # find the second signal delay
print(j,k)
```



10 100

10 (g)

This is the same code as part (f), but with slight modification to how the received signal y generated. Run the below cell a few times, to test for different choices of random signals.

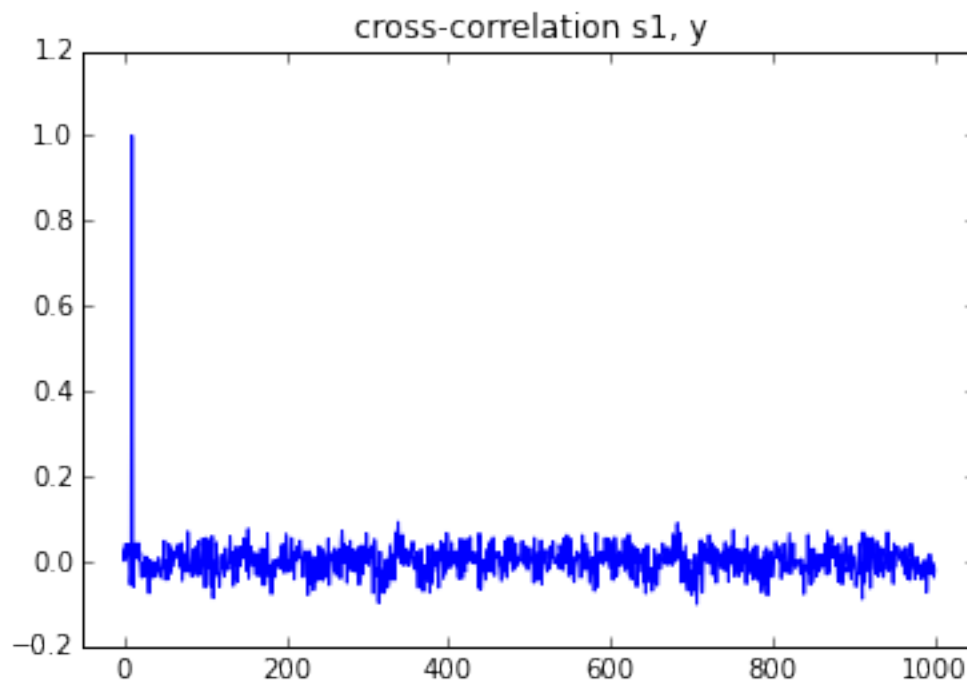
```
In [177]: s1 = rand_normed_vector(N)
          s2 = rand_normed_vector(N)

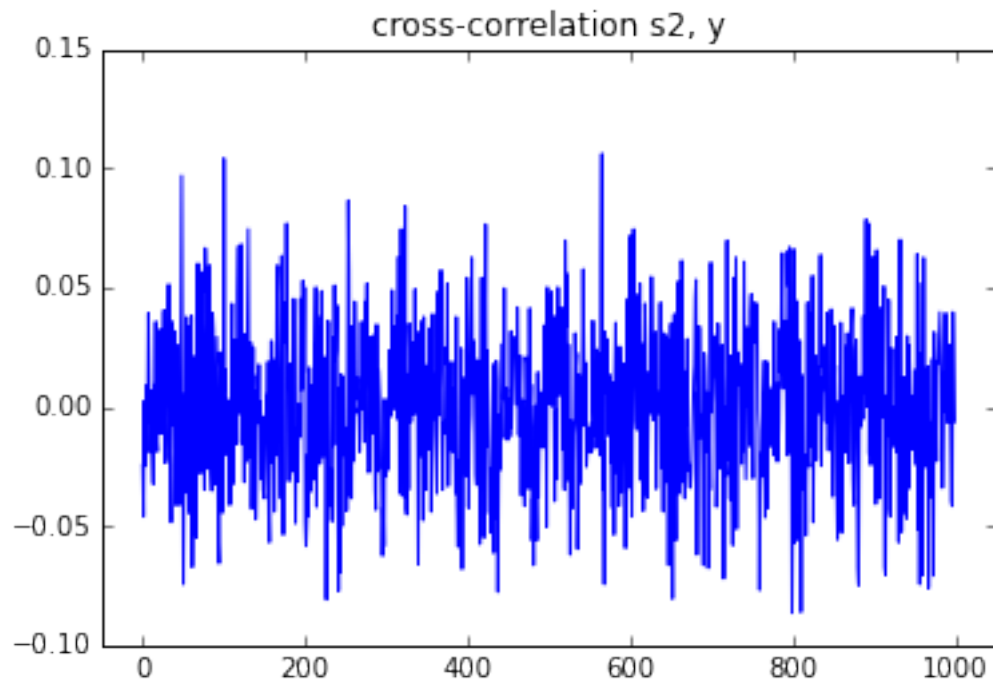
          y = np.roll(s1, 10) + 0.1*np.roll(s2, 100)

          # Compute cross-correlations:
          corr_s1_y = cross_corr(s1, y)
          corr_s2_y = cross_corr(s2, y)

          # Plot cross-correlations:
          plt.title("cross-correlation s1, y")
          plt.plot(cross_corr(s1, y))
          x1,x2,y1,y2 = plt.axis()
          plt.axis([x1-50,x2+50,y1,y2])
          plt.show()

          plt.title("cross-correlation s2, y")
          plt.plot(cross_corr(s2, y))
          x1,x2,y1,y2 = plt.axis()
          plt.axis([x1-50,x2+50,y1,y2])
          plt.show()
```





11 (h)

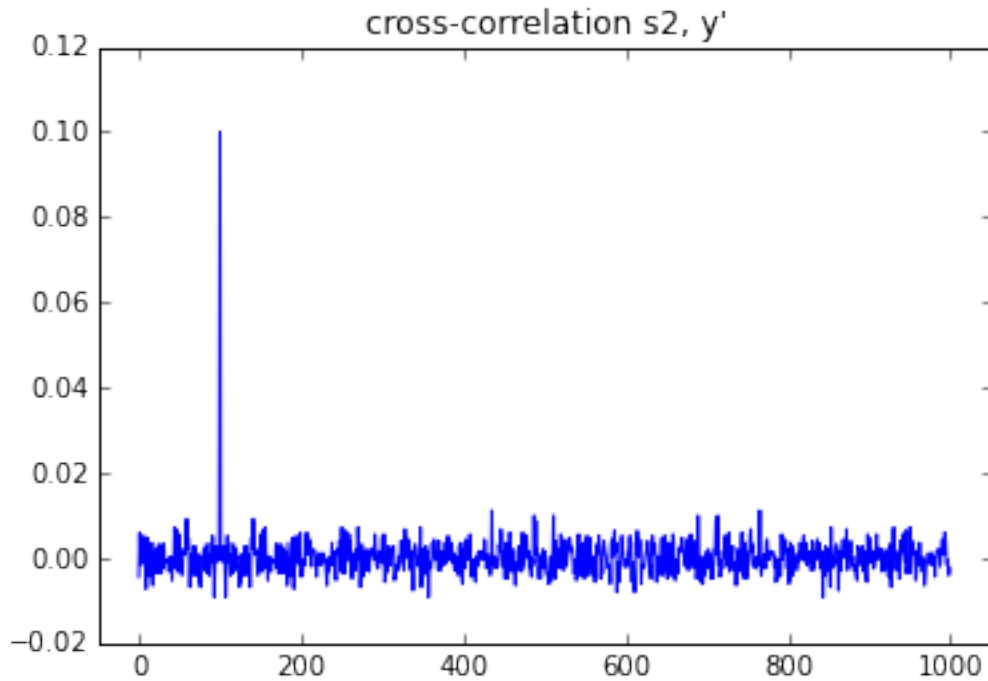
```
In [178]: corr_s1_y = cross_corr(s1, y)
          j = np.argmax(corr_s1_y) # find the first signal delay
          print(j)

          # subtract out the contribution of the first signal
          y_prime = y - np.roll(s1, j)

          # correlate the residual against the second signal
          corr_s2_y = cross_corr(s2, y_prime)

          # Plot
          plt.title("cross-correlation s2, y'")
          plt.plot(corr_s2_y)
          x1,x2,y1,y2 = plt.axis()
          plt.axis([x1-50,x2+50,y1,y2])
          plt.show()

          k = np.argmax(corr_s2_y) # find the second signal delay by looking at the index of max correl
          print(k)
```



100

12 (i)

```
In [179]: s1 = rand_normed_vector(N)
          s2 = rand_normed_vector(N)

          y = 0.7*np.roll(s1, 10) + 0.5*np.roll(s2, 100)

          corr_s1_y = cross_corr(s1, y)
          j = np.argmax(corr_s1_y) # find the first signal delay

          corr_s2_y = cross_corr(s2, y)
          k = np.argmax(corr_s2_y) # find the second signal delay

          print(j, k)

          # Once we have found the shifts, estimate the coefficients as inner-products:
          a1 = np.dot(y, np.roll(s1, j))
          a2 = np.dot(y, np.roll(s2, k))

          print(a1, a2)
```

```
10 100
0.708 0.5112
```

13 (j)

This is the same code as part (i), but with noise added to the received signal y.

```
In [180]: s1 = rand_normed_vector(N)
          s2 = rand_normed_vector(N)
          n = rand_normed_vector(N)

          y = 0.7*np.roll(s1, 10) + 0.5*np.roll(s2, 100) + 0.1*n

          corr_s1_y = cross_corr(s1, y)
          j = np.argmax(corr_s1_y) # find the first signal delay

          corr_s2_y = cross_corr(s2, y)
          k = np.argmax(corr_s2_y) # find the second signal delay

          print(j, k)

          # Once we have found the shifts, estimate the coefficients as inner-products:
          a1 = np.dot(y, np.roll(s1, j))
          a2 = np.dot(y, np.roll(s2, k))

          print(a1, a2)

10 100
0.7014 0.5052
```

14 (k)

```
In [181]: # Given the shifts j, k, setup the matrix A and vector b.

          # Hint: use np.roll(...) to circularly shift vectors.
          # For example, "np.roll(s1, j)" shifts the vector s1 by j indices.
          # A has columns c1, c2 which you should FILL IN BELOW.
          c1 = np.roll(s1, j)
          c2 = np.roll(s2, k)
          A = np.array([c1, c2]).T

          b = y

          # Solve to find the linear least-square solution of Ax ~ b (minimizing error ||Ax - b||)
          xhat = la.inv(A.T.dot(A)).dot(A.T).dot(b)
          print(xhat)

[ 0.69839394  0.50100964]
```

15 (l)

```
In [182]: # Load the signal vectors from file.
          npzfile = np.load("signals.npz")
          y, s1, s2, s3 = [npzfile[f] for f in ['y', 's1', 's2', 's3']]
```

Try to find the delays and coefficients of the three signals s1, s2, s3, from your received signal y. Hint: Make use of the provided code in the previous parts. This should be possible by mostly copy/pasting code. In particular, remember:

- “np.roll(s1, 123)” circularly shifts vector s1 by 123
- “np.argmax(corr)” finds the index of the maximum entry in vector “corr”.

Once you have found candidate delays j, k, l, try running the following function. You should recognize the output.

```
In [183]: # Test your j,k,l by running this function:
def test(j,k,l):
    return [chr(int(i)) for i in (np.array([j,k,l])/20 + 60)]

In [184]: ## TRY TO FIND THE SIGNALS HERE.

print("\n Find delays")
corr_s1_y = cross_corr(s1, y)
j = np.argmax(corr_s1_y) # find the first signal delay
print(j)
plt.plot(corr_s1_y)
plt.show()

corr_s2_y = cross_corr(s2, y)
k = np.argmax(corr_s2_y) # find the second signal delay
print(k)
plt.plot(corr_s2_y)
plt.show()

corr_s3_y = cross_corr(s3, y)
l = np.argmax(corr_s3_y) # find the third signal delay
print(l)
plt.plot(corr_s3_y)
plt.show()

print("Signal 2 is the loudest so remove its contribution.")
c1 = np.roll(s1,j)
c2 = np.roll(s2, k)
c3 = np.roll(s3, l)
A = np.array([c1, c2, c3]).T

b = y

# Solve to find the linear least-square solution of  $Ax \sim b$  (minimizing error  $\|Ax - b\|$ )
xhat = la.inv(A.T.dot(A)).dot(A.T).dot(b)

# subtract out the contribution of second signal
y_prime = y - xhat[1]*c2

print("\n Then repeat process of finding delays")
corr_s1_y = cross_corr(s1, y_prime)
j = np.argmax(corr_s1_y)
print(j)
plt.plot(corr_s1_y)
plt.show()

corr_s3_y = cross_corr(s3, y_prime)
```

```

l = np.argmax(corr_s3_y)
print(l)
plt.plot(corr_s3_y)
plt.show()

print("Signal 1 is loud so subtract its contribution.")
c1 = np.roll(s1,j)
c2 = np.roll(s2, k)
c3 = np.roll(s3, l)
A = np.array([c1, c2, c3]).T

b = y

# Solve to find the linear least-square solution of  $Ax \sim b$  (minimizing error  $\|Ax - b\|$ )
xhat = la.inv(A.T.dot(A)).dot(A.T).dot(b)

# subtract out the contribution of the first signal
y_prime = y - xhat[1]*c2 -xhat[0]*c1

print("Now just solve for signal 3")
corr_s3_y = cross_corr(s3, y_prime)
l = np.argmax(corr_s3_y)
print(l)
plt.plot(corr_s3_y)
plt.show()

c1 = np.roll(s1,j)
c2 = np.roll(s2, k)
c3 = np.roll(s3, l)
A = np.array([c1, c2, c3]).T

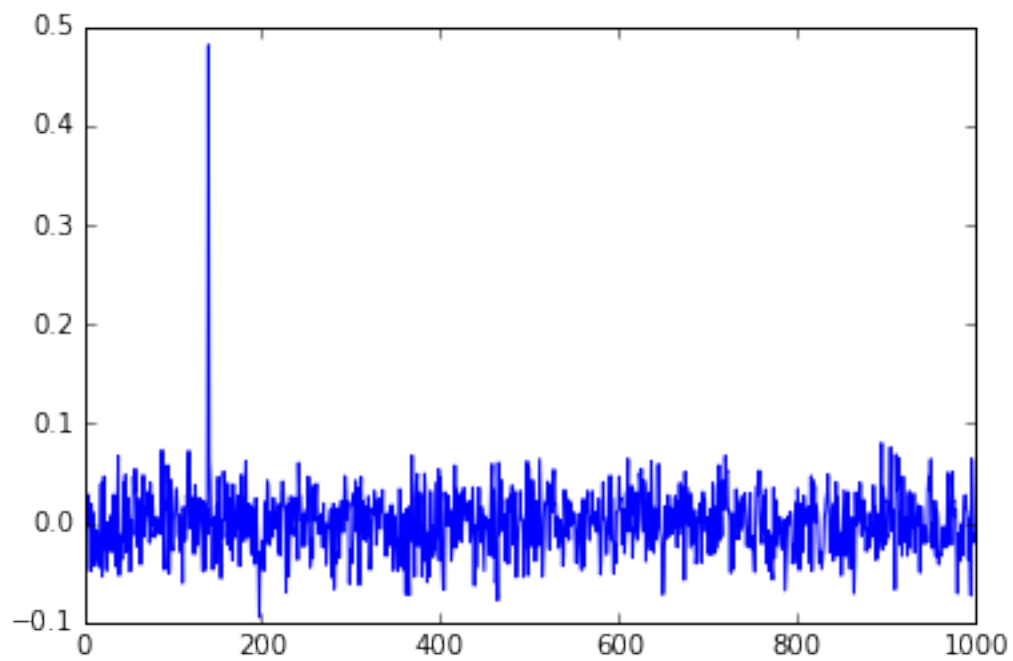
b = y

# Solve to find the linear least-square solution of  $Ax \sim b$  (minimizing error  $\|Ax - b\|$ )
xhat = la.inv(A.T.dot(A)).dot(A.T).dot(b)
print("The new coefficients we get are: ", xhat)

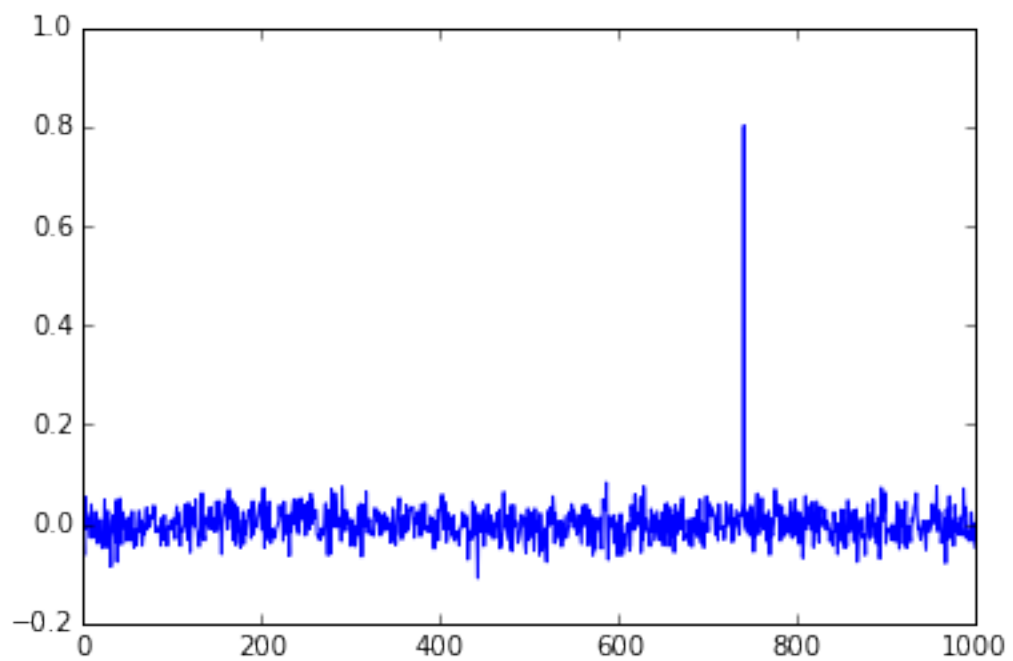
test(j,k,l)

```

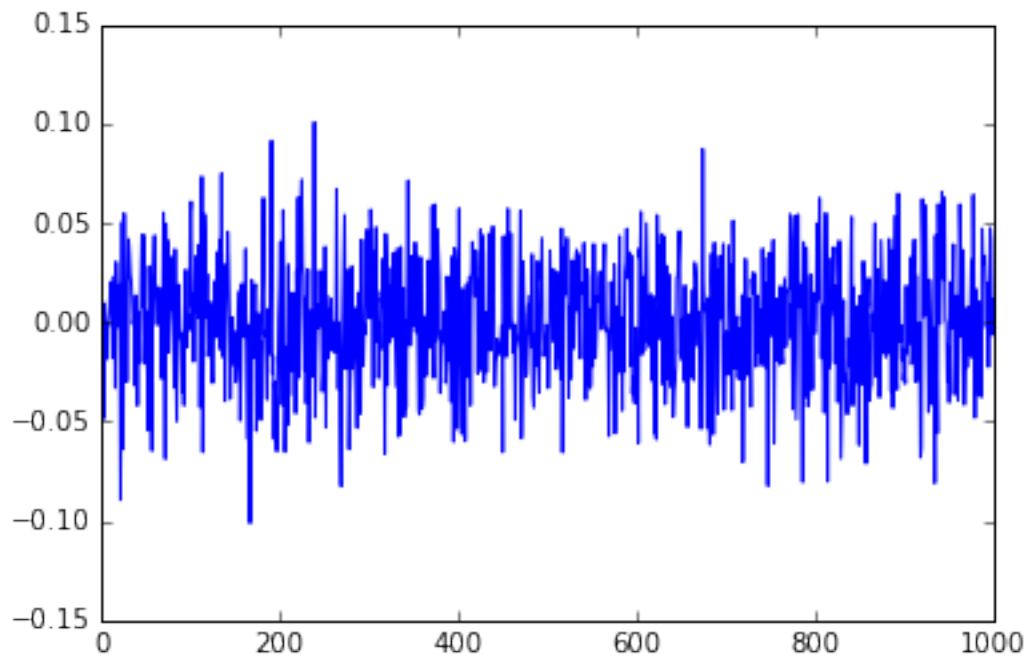
Find delays
140



740

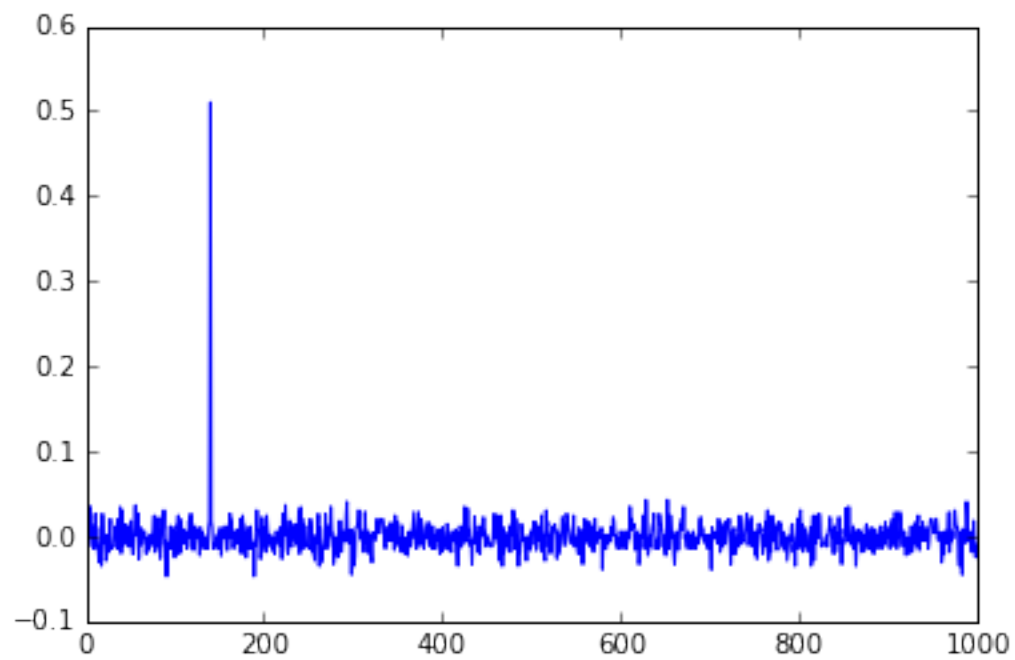


239

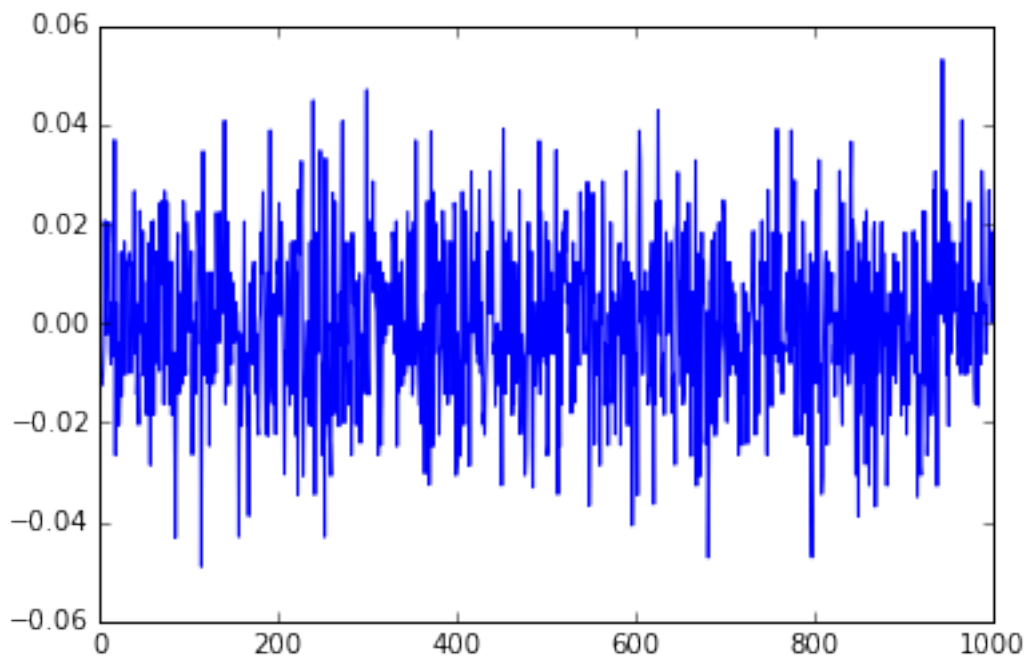


Signal 2 is the loudest so remove its contribution.

Then repeat process of finding delays
140



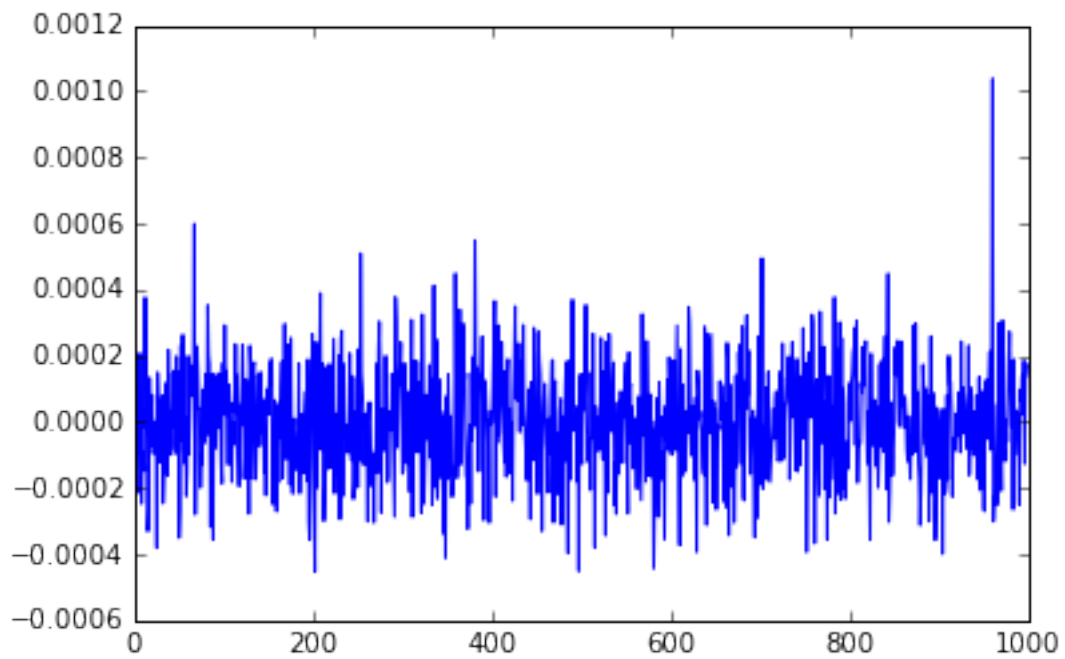
943



Signal 1 is loud so subtract its contribution.

Now just solve for signal 3

960



The new coefficients we get are: [0.51019847 0.81994888 0.00103934]

Out[184]: ['C', 'a', 'l']

16 Deconstructing Trolls

```
In [185]: import numpy as np
import matplotlib.pyplot as plt
import wave as wav
import scipy
from scipy import io
import scipy.io.wavfile
from scipy.io.wavfile import read
from IPython.display import Audio
import warnings
warnings.filterwarnings('ignore')
sound_file_1 = 'm1.wav'
sound_file_2 = 'm2.wav'
rate1, corrupt1 = scipy.io.wavfile.read('m1.wav')
rate2, corrupt2 = scipy.io.wavfile.read('m2.wav')
```

Just as last time, let's listen to the inputs.

```
In [186]: Audio(url='m1.wav', autoplay=False)
```

Out[186]: <IPython.lib.display.Audio object>

```
In [187]: Audio(url='m2.wav', autoplay=False)
```

Out[187]: <IPython.lib.display.Audio object>

In the cell below, complete the function to find the vectors \vec{a} and \vec{b} . Make sure \vec{a} is the original speech and not the troll.

```
In [189]: def remove_troll(m1, m2):
    ##Your code here
```

```
    a =
    b =
    return a, b
```

```
File "<ipython-input-189-531044194be5>", line 5
a =
~
```

SyntaxError: invalid syntax

Run the cell below to test your function

```
In [190]: a, b = remove_troll(corrupt1, corrupt2)
Audio(data=a, rate=rate1)
```

```

-----

NameError                                Traceback (most recent call last)

<ipython-input-190-336f73bed4ad> in <module>()
----> 1 a, b = remove_troll(corrupt1, corrupt2)
      2 Audio(data=a, rate=rate1)

NameError: name 'remove_troll' is not defined

```

Let's now compare our output here to the output from Homework 1. Read through the block of code below and comment on it's output

```

In [191]: ## First let's compute the original vectors representing the speakers using the technique in
a_u = np.sqrt(2)/(1+np.sqrt(3))
a_v = np.sqrt(6)/(1+np.sqrt(3))
b_u = np.sqrt(2)/(1+np.sqrt(3))
b_v = -1*np.sqrt(2)/(1+np.sqrt(3))
s1 = a_u*corrupt1 + a_v*corrupt2
s2 = b_u*corrupt1 + b_v*corrupt2

## Here we will compute various dot products to see which vectors are orthogonal.
## Note that we normalize the vectors before we compare, this is because we want
## to get rid of any scaling.
print("Dot product of the two original speaker outputs ", np.dot(s1/np.linalg.norm(s1), s2/np
print("Dot product of calculated a and b ", np.dot(a/np.linalg.norm(a), b/np.linalg.norm(b)))

Dot product of the two original speaker outputs  -0.00379040459598

```

```

-----

ValueError                                Traceback (most recent call last)

<ipython-input-191-be4cba61913b> in <module>()
    11 ## to get rid of any scaling.
    12 print("Dot product of the two original speaker outputs ", np.dot(s1/np.linalg.norm(s1), s2/np
----> 13 print("Dot product of calculated a and b ", np.dot(a/np.linalg.norm(a), b/np.linalg.norm(b)))

ValueError: shapes (6,) and (1000,) not aligned: 6 (dim 0) != 1000 (dim 0)

```

In []: