Math 110B Project 2: Random or not so random?

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Task 1

In [105]:

In [107]:

```
# generate a vector of random numbers which obeys the given distribution.
# n: length of the vector
# mu: mean value
# sigma: standard deviation.
# dist: choices for the distribution, you need to implement at least normal
        distribution and uniform distribution.
# For normal distribution, you can use ``numpy.random.normal`` to generate.
# For uniform distribution, the interval to sample will be [mu - sigma/sqrt(3),
mu + sigma/sgrt(3)].
from future import division
import numpy as np
import math
def generate_random_numbers(n, mu, sigma, dist="normal"):
    # write your code here.
    if dist == "normal":
        return np.random.normal(mu, sigma, n)
    elif dist == "uniform":
        return np.random.uniform(mu - sigma/math.sqrt(3), mu + sigma/math.sqrt(3)
), n)
    else:
        raise Exception ("The distribution {unknown dist} is not implemented".for
mat(unknown dist=dist))
# test your code:
y test = generate random numbers(5, 0, 0.1, "normal")
In [106]:
y1 = generate random numbers(105, 0.5, 1.0, "normal")
y2 = generate random numbers(105, 0.5, 1.0, "uniform")
x0 = 0
#yk = 1/(k+1) step length
```

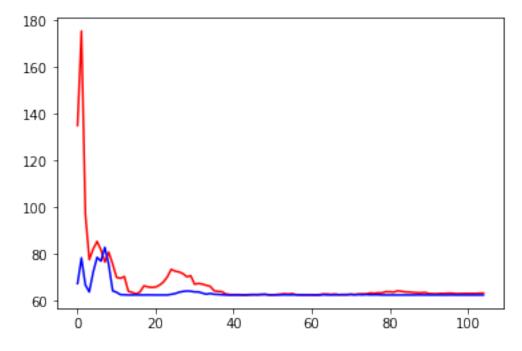
```
def IGD_wr_task1(y):
    n = len(y)
    ordering = np.random.choice(n, n, replace=True)
    #print(ordering)
    x0 = 0
    xk = x0
    Listobj = []
    # implement the algorithm's iteration of IGD. Your result should return the
the final xk
    # at the last iteration and also the history of objective function at each x
k.
    for i in range(n):
        target = y[ordering[i]]
        gama = 1/(i + 1)
        xk = xk - gama*(xk - target)
        objk = 0
        for j in y:
            objk += (xk - j)**2
        objk = objk/2
        Listobj.append(objk)
    return xk, Listobj
xk1 ,Listobj1 = IGD wr task1(y1)
xk2 ,Listobj2 = IGD wr task1(y2)
def IGD_wo_task1(y):
    n = len(y)
    ordering = np.random.choice(n, n, replace=False)
    #print(ordering)
    Listx = []
    Listobj = []
    x0 = 0
    xk = x0
    # implement the algorithm's iteration of IGD. Your result should return the
the final xk
    # at the last iteration and also the history of objective function at each x
k.
    for i in range(n):
        target = y[ordering[i]]
        gama = 1/(i + 1)
        xk = xk - gama*(xk - target)
        Listx.append(xk)
        objk = 0
        for j in y:
            objk += (xk - j)**2
        objk = objk/2
        Listobj.append(objk)
    return xk, Listobj #Listx
xk1_no ,Listobj1_no = IGD_wo_task1(y1)
```

```
xk2_no ,Listobj2_no = IGD_wo_task1(y2)
```

In [108]:

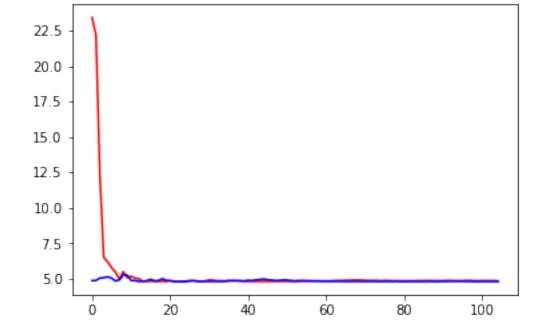
```
import matplotlib.pyplot as plt

plt.plot(np.arange(105), Listobj1, 'r')
plt.plot(np.arange(105), Listobj1_no, 'b')
plt.show()
```



In [109]:

```
plt.plot(np.arange(105), Listobj2, 'r')
plt.plot(np.arange(105), Listobj2_no, 'b')
plt.show()
```



For task 1, ordering without replacement works better.

The proof that IGD_wo_task1 must converge to the true solution is attached in our github repo.

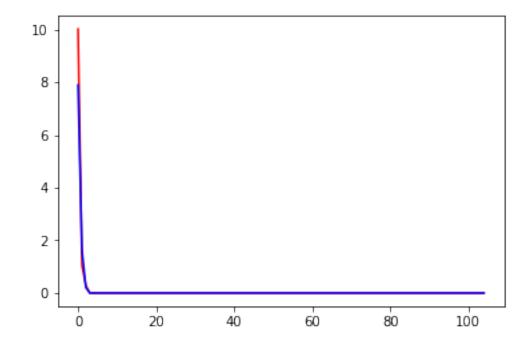
Task 2

```
In [137]:
beta lst = np.random.uniform(1, 2, 105)
\#y = 1
def IGD wr task2(beta lst):
    y=1
    n = len(beta_lst)
    ordering = np.random.choice(n, n, replace=True)
    x0 = 0
    xk = x0
    Listobj = []
    #beta lst = np.random.uniform(1, 2, n)
    gama = 0.95*np.min(1/beta lst)
    for i in range(n):
        #target = y[ordering[i]]
        beta = beta lst[ordering[i]]
        xk = xk - gama*beta*(xk - y)
        objk = 0
        for j in beta_lst:
            objk += j*(xk - y)**2
        objk = objk/2
        Listobj.append(objk)
    return xk, Listobj
xk1 ,Listobj1 = IGD wr task2(beta lst)
xk2 ,Listobj2 = IGD wr task2(beta lst)
def IGD wo task2(beta lst):
    y=1
    n = len(beta lst)
    ordering = np.random.choice(n, n, replace=False)
    x0 = 0
    xk = x0
    Listobj = []
    #beta_lst = np.random.uniform(1, 2, n)
    gama = 0.95*np.min(1/beta lst)
    for i in range(n):
        #target = y[ordering[i]]
        beta = beta_lst[ordering[i]]
        xk = xk - gama*beta*(xk - y)
        objk = 0
        for j in beta 1st:
            objk += j*(xk - y)**2
        objk = objk/2
        Listobj.append(objk)
    return xk. Listobi
```

```
xk1_no ,Listobj1_no = IGD_wo_task2(beta_lst)
xk2_no ,Listobj2_no = IGD_wo_task2(beta_lst)
```

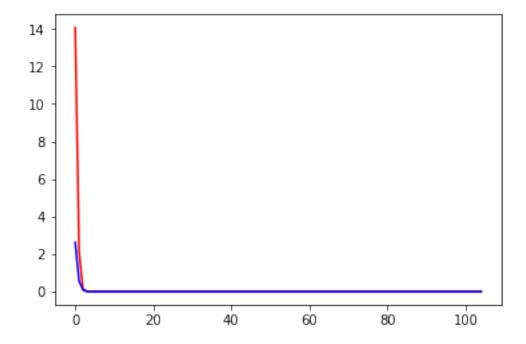
In [138]:

```
plt.plot(np.arange(105), Listobj1, 'r')
plt.plot(np.arange(105), Listobj1_no, 'b')
plt.show()
```



In [139]:

```
plt.plot(np.arange(105), Listobj2, 'r')
plt.plot(np.arange(105), Listobj2_no, 'b')
plt.show()
```



For task 2 1-D dimension example, it seems that the ordering without replacement works better.

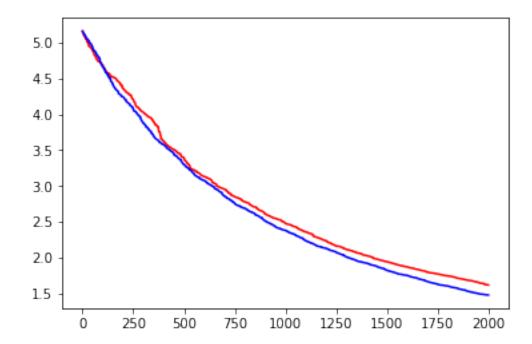
```
Task 3
In [130]:
# generation of exact solution and data y and matrix A.
def generate problem task3(m, n, rho):
    A = np.random.normal(0., 1.0, (m, n))
    x = np.random.random(n) # uniform in (0,1)
    w = np.random.normal(0., rho, m)
    y = np.dot(A,x) + w
    return A, x, y
A1, xstar1, y1 = generate_problem_task3(200, 100, 0.01)
A2, xstar2, y2 = generate problem task3(200, 100, 0.01)
In [131]:
from numpy import linalg as LA
In [132]:
# In these two functions, we could only focus on the first n steps and try to ma
ke comparisons on these data only.
# In practice, it requires more iterations to converge, due to the matrix might
not be easy to deal with.
```

```
# You can put the ordering loop into a naive loop: namely, we simply perform the
IGD code several rounds.
#
# IGD, the ordering is permitted to have replacement.
#
#
round num = 10
def IGD wr task3(y, A, xstar):
    # implement the algorithm's iteration of IGD. Your result should return the
the final xk
    # at the last iteration and also the history of objective function at each x
k.
    n = len(y)
    x0 = np.zeros(A.shape[1])
    xk = x0
    Listobj = []
    Listxk = []
    step size = 10**(-3)
    for each in range(round num):
        ordering = np.random.choice(n, n, replace=True)
        for i in range(n):
            target = y[ordering[i]]
            ak = A[ordering[i],:]
            xk = xk - step size*ak*(np.dot(ak,xk) - target)
            objk = 0
```

```
for j in range(n):
                a = A[j,:]
                objk += (np.dot(a,xk) - y[j])**2
            Listobj.append(objk)
            xkdif = LA.norm(xk-xstar)
            Listxk.append(xkdif)
    return xk, Listobj, Listxk
xk1 ,Listobj1,Listxk1 = IGD_wr_task3(y1,A1, xstar1)
xk2 ,Listobj2,Listxk2 = IGD wr task3(y2,A2, xstar2)
# IGD, the ordering is not permitted to have replacement.
#
#
def IGD wo task3(y, A, xstar):
    # implement the algorithm's iteration of IGD. Your result should return the
the final xk
    # at the last iteration and also the history of objective function at each x
k.
    n = len(y)
    ordering = np.random.choice(n, n, replace=False)
    x0 = np.zeros(A.shape[1])
    xk = x0
    Listobj = []
    Listxk = []
    step size = 10**(-3)
    for each in range(round num):
        for i in range(n):
            target = y[ordering[i]]
            ak = A[ordering[i],:]
            xk = xk - step size*ak*(np.dot(ak,xk) - target)
            objk = 0
            for j in range(n):
                a = A[j,:]
                objk += (np.dot(a,xk) - y[j])**2
            Listobj.append(objk)
            xkdif = LA.norm(xk-xstar)
            Listxk.append(xkdif)
    return xk, Listobj, Listxk
xk1 no ,Listobj1 no, Listxk1_no = IGD_wo_task3(y1,A1, xstar1)
xk2 no ,Listobj2 no, Listxk2 no = IGD wo task3(y2,A2, xstar2)
```

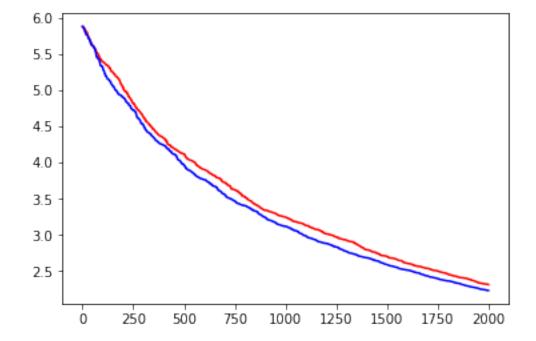
In [133]:

```
plt.plot(np.arange(len(Listobj1)), Listxk1, 'r')
plt.plot(np.arange(len(Listobj1_no)), Listxk1_no, 'b')
plt.show()
```



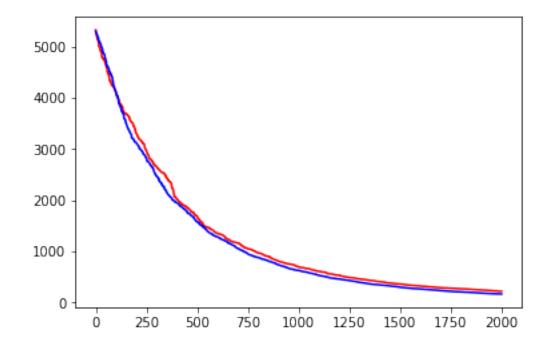
In [134]:

```
plt.plot(np.arange(len(Listobj1)), Listxk2, 'r')
plt.plot(np.arange(len(Listobj1_no)), Listxk2_no, 'b')
plt.show()
```



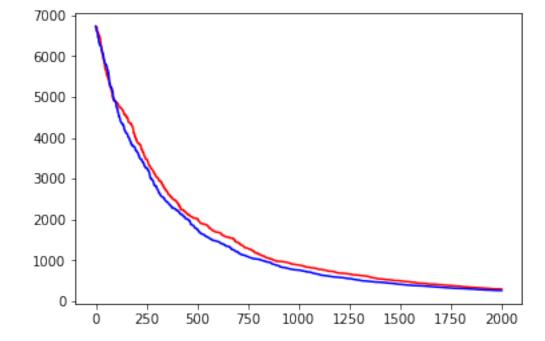
In [135]:

```
plt.plot(np.arange(len(Listobj1)), Listobj1, 'r')
plt.plot(np.arange(len(Listobj1_no)), Listobj1_no, 'b')
plt.show()
```



In [136]:

```
plt.plot(np.arange(len(Listobj2)), Listobj2, 'r')
plt.plot(np.arange(len(Listobj2_no)), Listobj2_no, 'b')
plt.show()
```



For task 3, it seems that the ordering without replacement works better.

In []: