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Homework 3

2 Programming

We are interested in solving the following low rank matrix problem. Given a sparse observation pattern $G \in \{0,1\}^{n \times n}$ and a data matrix $A \in \mathbb{R}^n$ our goal is to recover a low rank matrix pair, $B \in \mathbb{R}^{n \times r}$, $C \in \mathbb{R}^{r \times n}$ by minimizing

$$\min_{B,C} f(B,C) = \sum_{i=1}^{n} \sum_{j=1}^{n} g_{ij} (a_{ij} - e_i^T B(Ce_j))^2 + \frac{\mu}{2} (\|B\|_F^2 + \|C\|_F^2)$$

Note that this is the same as:

$$\min_{B,C} \left\langle G, (A - BC) \circ (A - BC) \right\rangle_F + \frac{\mu}{2} \left\langle B, B \right\rangle_F + \frac{\mu}{2} \left\langle C, C \right\rangle_F$$

Where \circ is the Hadamard product, and $\langle \cdot, \cdot \rangle_F$ is the Frobenius inner product.

Deriving the gradient is an exercise in tensor calculus:

$$g(f) = \left[\nabla_B f, \nabla_C f \right]$$

and $\nabla_{B} f \in \mathbb{R}^{n \times r}$ likewise $\nabla_{C} f \in \mathbb{R}^{r \times n}$

$$\frac{\partial f}{\partial b_{lm}} = \frac{\partial}{\partial b_{lm}} \left(\sum_{i=1}^{n} \sum_{j=1}^{n} g_{ij} (a_{ij} - \sum_{k=1}^{r} b_{ik} c_{kj}) (a_{ij} - \sum_{t=1}^{r} b_{it} c_{tj}) + \frac{\mu}{2} \sum_{i=1}^{n} \sum_{k=1}^{n} (b_{ik} b_{ik} + c_{ik} c_{ik}) \right)$$

I will first focus on the cost or least squares portion:

$$\frac{\partial}{\partial b_{lm}} \sum_{i=1}^{n} \sum_{j=1}^{n} g_{ij} (a_{ij} - \sum_{k=1}^{r} b_{ik} c_{kj}) (a_{ij} - \sum_{t=1}^{r} b_{it} c_{tj})$$

$$= \sum_{i=1}^{n} \sum_{j=1}^{n} g_{ij} (-\sum_{k=1}^{r} \delta_{il} \delta_{km} c_{kj}) (a_{ij} - \sum_{t=1}^{r} b_{it} c_{tj}) + \sum_{i=1}^{n} \sum_{j=1}^{n} g_{ij} (a_{ij} - \sum_{k=1}^{r} b_{ik} c_{kj}) (-\sum_{t=1}^{r} \delta_{il} \delta_{tm} c_{tj})$$

$$= \sum_{i=1}^{n} \sum_{j=1}^{n} g_{ij} (-\delta_{il} c_{mj}) (a_{ij} - \sum_{t=1}^{r} b_{it} c_{tj}) + \sum_{i=1}^{n} \sum_{j=1}^{n} g_{ij} (a_{ij} - \sum_{k=1}^{r} b_{ik} c_{kj}) (-\delta_{il} c_{mj})$$

$$= -2 \sum_{i=1}^{n} \sum_{j=1}^{n} \delta_{il} g_{ij} (A - BC)_{ij} c_{mj}$$

$$= -2 \sum_{i=1}^{n} \sum_{j=1}^{n} \delta_{il} g_{ij} (A - BC)_{lj} c_{mj}$$

$$=-2\sum_{j=1}^{n}\left(G\circ(A-BC)\right)_{lj}c_{mj}$$

$$= -2\bigg(\big(G \circ (A - BC) \big) C^T \bigg)_{lm}$$

A derivative with respect to the regulation yields

$$\frac{\partial}{\partial b_{lm}}(b_{ik}b_{ik}) = 2b_{ik}\delta_{il}\delta_{km} = 2b_{lm}$$

The whole gradient can be represented in matrix form as:

$$\nabla_B f = -2 \left(G \circ (A - BC) \right) C^T + \mu B \in \mathbb{R}^{n \times r}$$

Similarly for the gradient with respect to C, I derived it as:

$$\nabla_C f = -2B^T \left(G \circ (A - BC) \right) + \mu C$$

The Hessian action is needed at the very least to implement a trust region algorithm. The Hessian is a fourth order tensor, its action on a second order matrix results in a second order matrix (as is expected). Heuristically I will decompose the Hessian into a block structure as follows:

$$H = \begin{bmatrix} H_{BB} & H_{BC} \\ H_{CB} & H_{CC} \end{bmatrix}$$

each block has $n^2 r^2$ elements but the shapes vary as B and C have different shapes. Careful attention to indices is warranted.

$$(H_{BB})_{lmpq} = \frac{\partial}{\partial b_{pq}} (\nabla_B f)_{lm} = 2 \sum_{j=1}^n g_{lj} \delta_{lp} c_{qj} c_{mj} + \mu \delta_{lp} \delta_{kq}$$

$$(H_{BC})_{lmpq} = \frac{\partial}{\partial c_{qp}} (\nabla_B f)_{lm} = 2 g_{lp} b_{lq} c_{mp} - 2 (G \circ (A - BC))_{lp} \delta_{mq}$$

$$(H_{CB})_{lmpq} = \frac{\partial}{\partial b_{pq}} (\nabla_C f)_{lm} = 2 g_{pl} b_{pm} c_{ql} - 2 (G \circ (A - BC))_{pl} \delta_{mq}$$

$$(H_{CC})_{lmpq} = \frac{\partial}{\partial c_{qp}} (\nabla_C f)_{lm} = 2 \sum_{j=1}^n b_{im} g_{il} b_{iq} \delta_{lp} + \mu \delta_{mq} \delta_{lp}$$

In order to implement a trust region algorithm we only need the Hessian action (we can also do Newton with just the action if we use a Krylov solver).

The Hessian Action can be defined as follows:

$$\begin{bmatrix} H_{BB} & H_{BC} \\ H_{CB} & H_{CC} \end{bmatrix} \begin{bmatrix} X \\ Y \end{bmatrix} = \begin{bmatrix} H_{BB}X + H_{BC}Y \\ H_{CB}X + H_{CC}Y \end{bmatrix}$$

Each block multiplication is as follows:

$$H_{BB}X = [G \circ (XC)]C^{T} + \mu X$$

$$H_{BC}Y = 2C[G \circ (BY)]^{T} - 2Y[G \circ (A - BC)]^{T}$$

$$H_{CB}X = 2[G \circ (XC)]^{T}B - 2[G \circ (A - BC)]^{T}X$$

$$H_{CC}Y = B^{T}[G \circ (BY)] + \mu Y$$

I will now derive a class for this problem that will encapsulate all the important information.

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

class low_rank_matrix_handler:
    def __init__(self,A,B,C,G):
        self.A = A
        self.B = B
        self.C = C
        self.G = G
        self.mu = 1.0
        self.n,self.r = self.B.shape
```

```
def control(self):
        return np.array([self.B,self.C])
    def objective(self,B=None,C=None):
        if B is None:
            B = self.B
        if C is None:
            C = self.C
        temp = self.A - np.matmul(B,C)
        temp = np.multiply(temp,temp)
        cost = np.sum(np.multiply(self.G,temp))
        reg_B = 0.5*self.mu*np.sum(np.multiply(B,B))
        reg_C = 0.5*self.mu*np.sum(np.multiply(C,C))
        return cost + reg B + reg C
    def gradient(self):
        # Initialize to the regularization portion
        reg B = self.mu*self.B
        reg C = self.mu*self.C
        # Add the least squares (cost) portion
        A BC = self.A - np.matmul(self.B, self.C)
        G had A BC = np.multiply(self.G,A BC)
        cost B = -2*np.matmul(G_had_A_BC,self.C.T)
        cost C = -2*np.matmul(self.B.T,G had A BC)
        grad_B = cost_B + reg_B
        grad C = cost C + reg C
        return grad B, grad C
    def Hessian action(self, X = None, Y=None):
        if X is None:
            X = np.zeros_like(self.B)
        if Y is None:
            Y = np.zeros like(self.C)
        shape B = self.B.shape
        shape C = self.C.shape
        if X.shape != shape B:
            raise ValueError(X)
        if Y.shape != shape C:
            raise ValueError(Y)
        XC = np.matmul(X, self.C)
        G had XC = np.multiply(self.G,XC)
        H BB X = np.matmul(G had XC,self.C.T) + self.mu*X
        A BC = self.A - np.matmul(self.B, self.C)
        BY = np.matmul(self.B,Y)
        G had BY = np.multiply(self.G,BY)
        G had A BC = np.multiply(self.G,A BC)
        H_BC_Y = 2*np.matmul(self.C,G_had_BY.T) - 2*np.matmul(Y,G_had_A_
BC.T)
        H CB X = 2*np.matmul(G had XC.T, self.B) -
2*np.matmul(G had A BC.T,X)
        H CC Y = np.matmul(self.B.T,G had BY) + self.mu*Y
        return [H BB X+H BC Y.T,H CB X+ H CC Y.T]
    def Hessian_inner(self, X = None,Y=None):
        if X is None:
            X = np.zeros_like(self.B)
        if Y is None:
```

```
Y = np.zeros like(self.C)
        H XY = self.Hessian action(X,Y)
        H X = H XY[0]
        H Y = H XY[1]
        X H X = np.multiply(X, H X)
        Y_H_Y = np.multiply(Y.T,H_Y)
        return np.sum(X H X) + np.sum(Y H Y)
    def update_control(self, P_B= None,P_C = None,alpha = 1.):
        if P_B == None:
            P_B = np.zeros_like(self.B)
        if P_C == None:
            P_C = np.zeros_like(self.C)
        self.B += alpha*P_B
        self.C += alpha*P_C
    def assemble_flat_Hessian(self):
        n,r = self.B.shape
        dim = n*r
        H_BB = np.zeros((dim,dim))
        H_BC = np.zeros((dim,dim))
        H_CB = np.zeros((dim,dim))
        H_CC = np.zeros((dim,dim))
        A_BC = self.A - np.matmul(self.B, self.C)
        G_had_A_BC = np.multiply(self.G,A_BC)
        for 1 in range(n):
            for m in range(r):
                for p in range(n):
                    for q in range(r):
                        H_BB[1 + n*m][p + n*q] += self.mu*(l==p)*(m==q)
                        H_BC[1 + n*m][p + n*q] +=
2*self.G[1,p]*self.B[1,q]*self.C[m,p]
                            -2*G_had_A_BC[1,p]*(m == q)
                        H_CB[1 + n*m][p + n*q] +=
2*self.G[p,l]*self.B[p,m]*self.C[q,l]
                            - 2*G_had_A_BC[p,1]*(m == q)
                        H CC[1 + n*m][p + n*q] += self.mu*(1==p)*(m==q)
                        for j in range(n):
                            H BB[1 + n*m][p + n*q] += 2*self.G[1,j]*
(l==p)
                            *self.C[q,j]*self.C[m,j]
                            H_CC[1 + n*m][p + n*q] += 2*self.G[j,l]*
(1==p)
                            *self.B[j,m]*self.B[j,q]
        H = np.bmat([[H BB, H BC], [H CB, H CC]])
        return H
    def assemble flat gradient(self):
        n,r = self.B.shape
        dim = n*r
        flat_grad_B = np.zeros(dim)
        flat_grad_C = np.zeros(dim)
        A BC = self.A - np.matmul(self.B, self.C)
        G_had_A_BC = np.multiply(self.G,A_BC)
        for 1 in range(n):
            for m in range(r):
```

```
flat grad B[l+n*m] += self.mu*self.B[l,m]
                flat_grad_C[l+n*m] += self.mu*self.C[m,1]
                for j in range(n):
                    flat grad B[1+n*m] += -2*G had A BC[1,j]*self.C[m,j]
                    flat grad C[1+n*m] += -2*G had A BC[j,l]*self.B[j,m]
        g = np.concatenate([flat_grad_B,flat_grad_C])
        return g
    def split_and_reshape(self,p):
        temp B, temp C = np.split(p, 2)
        p_B = temp_B.reshape((self.n,self.r))
        p_C = temp_C.reshape((self.n,self.r))
        return p B, p C.T
    def predicted_reduction(self,p):
        g = self.gradient()
        gp = np.sum(np.multiply(g[0],p[0]))
        pHp = self.Hessian\_inner(X = p[0],Y=p[1])
        return -gp -0.5*pHp
    def tensor_norm(self,p):
        p norm 2 = np.sum(np.multiply(p[0],p[0])) + np.sum(
np.multiply(p[1],p[1]))
        return np.sqrt(p_norm_2)
    def tensor_rescale(self,p,alpha):
        return [alpha*p[0],alpha*p[1]]
    def assemble_tensor_Hessian(self):
        n,r = self.B.shape
        dim = n*r
        H = np.zeros((2*n,r,2*n,r))
        A_BC = self.A - np.matmul(self.B,self.C)
        G had A BC = np.multiply(self.G,A BC)
        for 1 in range(n):
            for m in range(r):
                for p in range(n):
                    for q in range(r):
                        #HBB
                        H[1,m,p,q] += self.mu*(l==p)*(m==q)
                        #HBC
                        H[n+1,m,p,q] +=
2*self.G[l,p]*self.B[l,q]*self.C[m,p]
                            -2*G had A BC[1,p]*(m == q)
                        #HCB
                        H[1,m,n+p,q] +=
2*self.G[p,l]*self.B[p,m]*self.C[q,l]
                            - 2*G had A BC[p,1]*(m == q)
                        H[n+1,m,n+p,q] += self.mu*(l==p)*(m==q)
                        for j in range(n):
                            #HBB
                            H[1,m,p,q] += 2*self.G[1,j]*(1==p)
                            *self.C[q,j]*self.C[m,j]
                            #HCC
                            H[n+1,m,n+p,q] += 2*self.G[j,l]*(l==p)
                            *self.B[j,m]*self.B[j,q]
```

```
return H
def assemble tensor gradient(self):
    n,r = self.B.shape
    gradient = np.zeros((2*n,r))
    A BC = self.A - np.matmul(self.B, self.C)
    G had A BC = np.multiply(self.G,A BC)
    for 1 in range(n):
        for m in range(r):
            #G B
            gradient[1,m] += self.mu*self.B[1,m]
            #G C
            gradient[n+1,m] += self.mu*self.C[m,1]
            for j in range(n):
                #G B
                gradient[1,m] += -2*G had A BC[1,j]*self.C[m,j]
                gradient[n+1,m] += -2*G_had_A_BC[j,1]*self.B[j,m]
    return gradient
def split_p(self,p):
    p B, p CT = np.split(p, 2)
    return p B, p CT.T
def Newton_direction_tensor(self):
    g = self.assemble tensor gradient()
    H = self.assemble tensor Hessian()
    p = np.linalg.tensorsolve(H,-g)
    return self.split p(p)
```

Now I need to generate the data matrix A, sparse observational pattern G and initialize the control variables B and C

Because it doesn't really matter I will take $A \sim \mathcal{N}(0, \sigma^2)^{n \times n}$

I take B and C to be a pair of low rank recovery of identity (i.e. they are both identity padded with extra zeros in the strictly rectuangular rows or columns see numpy.matlib.eye for implementation). And $B*C=I_r\in\mathbb{R}^{n\times n}$, identity up to the r^{th} diagonal and then all zeros.

```
In [3]: def alternating minimization(low rank matrix handler, tolerance = 1e-8, ma
        x_{iters} = 10000,
                              c_armijo = 1e-4,rho_armijo =0.9,verbose = False,
                              back_track = False,min_length = 1e-3):
            lagr = low_rank_matrix_handler
            grad norm0_sq = la.norm(lagr.gradient()[0])**2 + la.norm(lagr.gradie
        nt()[1])**2
            grad_norm0 = np.sqrt(grad_norm0_sq)
            grad_norm = grad_norm0
            iters = 0
            while iters <= max iters:</pre>
                 for i in range(2):
                     partial_grad_norm0 = la.norm(lagr.gradient()[i])
                     partial_grad_norm = partial_grad_norm0
                     while partial grad norm > tolerance*partial grad norm0:
                         cost = lagr.objective()
                         P = lagr.gradient()
                         alpha = 0.1
                         if back track:
                             cost old = cost
                             alpha = 1.
                             while alpha > min_length:
                                 B_test = lagr.B -alpha*P[0]
                                 C_test = lagr.C -alpha*P[1]
                                 if i ==0:
                                      cost = lagr.objective(B_test)
                                 elif i == 1:
                                      cost = lagr.objective(C test)
                                  if cost < cost old +c armijo*alpha*grad norm :</pre>
                                     break
                                 else:
                                      alpha *= rho armijo
                         if i == 0:
                             lagr.B -= alpha*P[0]
                         elif i == 1:
                             lagr.C -= alpha*P[1]
                         partial_grad_norm = la.norm(P[i])
                         grad_norm = lagr.tensor_norm(P)
                         if (grad_norm < tolerance*grad_norm0):</pre>
```

```
return iters
                 if verbose:
                     alt_string = ['B','C']
                     if iters == 0:
                         print "\n{0:5} {1:5} {2:15} {3:15} {4:20} {5:1
5}".format(
                               "Iter", 'Alt' , "cost", "||g||L2", "partial
 \|g\|L2", "alpha")
                     else:
                         print "\n{0:<5d} {1:<5} {2:<15e} {3:<15e} {4:<20</pre>
e} {5:<15e}".format(</pre>
                               iters, alt_string[i], cost, grad_norm, par
tial_grad_norm, alpha )
                converged = (grad_norm < tolerance*grad_norm0)</pre>
                 if converged:
                     return converged, iters
                 iters += 1
    return converged, iters
```

```
In [4]: B = le-l*np.ones((n,r))
C = le-l*np.ones((r,n))
lagr = low_rank_matrix_handler(A,B,C,G)

g = lagr.gradient()
alternating_minimization(lagr,back_track = False,verbose=True)
```

Iter Alt	cost	g L2	partial \ g\ L2	alpha
1 В 00e-01	1.219645e+00	8.072305e-01	5.401241e-01	1.0000
2 B 00e-01	1.192064e+00	7.729692e-01	4.811701e-01	1.0000
3 B 00e-01	1.170175e+00	7.467592e-01	4.286594e-01	1.0000
4 B 00e-01	1.152802e+00	7.271402e-01	3.818867e-01	1.0000
5 B 00e-01	1.139014e+00	7.128260e-01	3.402240e-01	1.0000
6 B 00e-01	1.128070e+00	7.027104e-01	3.031125e-01	1.0000
7 В 00e-01	1.119384e+00	6.958636e-01	2.700541e-01	1.0000
8 B 00e-01	1.112488e+00	6.915209e-01	2.406056e-01	1.0000
9 В 00e-01	1.107015e+00	6.890646e-01	2.143723e-01	1.0000
10 B 00e-01	1.102670e+00	6.880039e-01	1.910026e-01	1.0000
11 B 00e-01	1.099220e+00	6.879545e-01	1.701836e-01	1.0000
12 B 00e-01	1.096482e+00	6.886191e-01	1.516365e-01	1.0000
13 B 00e-01	1.094308e+00	6.897708e-01	1.351131e-01	1.0000
14 B 00e-01	1.092582e+00	6.912378e-01	1.203922e-01	1.0000
15 B 00e-01	1.091211e+00	6.928922e-01	1.072770e-01	1.0000
16 B 00e-01	1.090123e+00	6.946397e-01	9.559218e-02	1.0000
17 B 00e-01	1.089259e+00	6.964121e-01	8.518145e-02	1.0000
18 B 00e-01	1.088573e+00	6.981609e-01	7.590576e-02	1.0000

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19 B 00e-01	1.088028e+00	6.998529e-01	6.764121e-02	1.0000
20 B 00e-01	1.087596e+00	7.014661e-01	6.027744e-02	1.0000
21 B 00e-01	1.087252e+00	7.029867e-01	5.371616e-02	1.0000
22 B 00e-01	1.086979e+00	7.044076e-01	4.786981e-02	1.0000
23 B 00e-01	1.086763e+00	7.057257e-01	4.266041e-02	1.0000
24 B 00e-01	1.086590e+00	7.069415e-01	3.801849e-02	1.0000
25 B 00e-01	1.086454e+00	7.080576e-01	3.388215e-02	1.0000
26 B 00e-01	1.086345e+00	7.090781e-01	3.019627e-02	1.0000
27 В 00e-01	1.086259e+00	7.100081e-01	2.691175e-02	1.0000
28 B 00e-01	1.086191e+00	7.108532e-01	2.398483e-02	1.0000
29 B 00e-01	1.086136e+00	7.116193e-01	2.137653e-02	1.0000
30 B 00e-01	1.086093e+00	7.123124e-01	1.905214e-02	1.0000
31 B 00e-01	1.086059e+00	7.129384e-01	1.698072e-02	1.0000
32 B 00e-01	1.086031e+00	7.135028e-01	1.513472e-02	1.0000
33 B 00e-01	1.086010e+00	7.140112e-01	1.348957e-02	1.0000
34 B 00e-01	1.085992e+00	7.144684e-01	1.202340e-02	1.0000
35 B 00e-01	1.085979e+00	7.148793e-01	1.071673e-02	1.0000
36 B 00e-01	1.085968e+00	7.152482e-01	9.552178e-03	1.0000
37 B 00e-01	1.085959e+00	7.155792e-01	8.514280e-03	1.0000

38 B 00e-01	1.085952e+00	7.158760e-01	7.589248e-03	1.0000
39 B 00e-01	1.085947e+00	7.161419e-01	6.764796e-03	1.0000
40 B 00e-01	1.085943e+00	7.163800e-01	6.029978e-03	1.0000
41 B 00e-01	1.085939e+00	7.165932e-01	5.375041e-03	1.0000
42 B 00e-01	1.085936e+00	7.167839e-01	4.791293e-03	1.0000
43 B 00e-01	1.085934e+00	7.169545e-01	4.270991e-03	1.0000
44 B 00e-01	1.085933e+00	7.171071e-01	3.807231e-03	1.0000
45 B 00e-01	1.085931e+00	7.172435e-01	3.393865e-03	1.0000
46 B 00e-01	1.085930e+00	7.173654e-01	3.025412e-03	1.0000
47 B 00e-01	1.085929e+00	7.174743e-01	2.696989e-03	1.0000
48 B 00e-01	1.085929e+00	7.175716e-01	2.404242e-03	1.0000
49 B 00e-01	1.085928e+00	7.176585e-01	2.143294e-03	1.0000
50 B 00e-01	1.085928e+00	7.177361e-01	1.910687e-03	1.0000
51 B 00e-01	1.085927e+00	7.178054e-01	1.703341e-03	1.0000
52 B 00e-01	1.085927e+00	7.178673e-01	1.518511e-03	1.0000
53 B 00e-01	1.085927e+00	7.179226e-01	1.353750e-03	1.0000
54 B 00e-01	1.085927e+00	7.179719e-01	1.206877e-03	1.0000
55 B 00e-01	1.085926e+00	7.180159e-01	1.075949e-03	1.0000
56 B 00e-01	1.085926e+00	7.180552e-01	9.592337e-04	1.0000

57 B 00e-01	1.085926e+00	7.180903e-01	8.551867e-04	1.0000
58 B 00e-01	1.085926e+00	7.181216e-01	7.624324e-04	1.0000
59 B 00e-01	1.085926e+00	7.181495e-01	6.797441e-04	1.0000
60 B 00e-01	1.085926e+00	7.181744e-01	6.060289e-04	1.0000
61 B 00e-01	1.085926e+00	7.181967e-01	5.403123e-04	1.0000
62 B 00e-01	1.085926e+00	7.182165e-01	4.817259e-04	1.0000
63 B 00e-01	1.085926e+00	7.182342e-01	4.294956e-04	1.0000
64 B 00e-01	1.085926e+00	7.182500e-01	3.829313e-04	1.0000
65 B 00e-01	1.085926e+00	7.182641e-01	3.414181e-04	1.0000
66 B 00e-01	1.085926e+00	7.182767e-01	3.044076e-04	1.0000
67 B 00e-01	1.085926e+00	7.182879e-01	2.714112e-04	1.0000
68 B 00e-01	1.085926e+00	7.182979e-01	2.419933e-04	1.0000
69 B 00e-01	1.085926e+00	7.183068e-01	2.157656e-04	1.0000
70 B 00e-01	1.085926e+00	7.183148e-01	1.923819e-04	1.0000
71 B 00e-01	1.085926e+00	7.183219e-01	1.715336e-04	1.0000
72 B 00e-01	1.085926e+00	7.183282e-01	1.529458e-04	1.0000
73 B 00e-01	1.085926e+00	7.183339e-01	1.363731e-04	1.0000
74 B 00e-01	1.085926e+00	7.183389e-01	1.215970e-04	1.0000
75 B 00e-01	1.085926e+00	7.183434e-01	1.084226e-04	1.0000

76 B	1.085926e+00	7.183474e-01	9.667623e-05	1.0000
00e-01				
77 B 00e-01	1.085926e+00	7.183510e-01	8.620301e-05	1.0000
78 B 00e-01	1.085926e+00	7.183542e-01	7.686488e-05	1.0000
79 B 00e-01	1.085926e+00	7.183571e-01	6.853876e-05	1.0000
80 B 00e-01	1.085926e+00	7.183596e-01	6.111491e-05	1.0000
81 B 00e-01	1.085926e+00	7.183619e-01	5.449551e-05	1.0000
82 B 00e-01	1.085926e+00	7.183639e-01	4.859336e-05	1.0000
83 B 00e-01	1.085926e+00	7.183657e-01	4.333070e-05	1.0000
84 B 00e-01	1.085926e+00	7.183673e-01	3.863820e-05	1.0000
85 B 00e-01	1.085926e+00	7.183687e-01	3.445408e-05	1.0000
86 B 00e-01	1.085926e+00	7.183700e-01	3.072323e-05	1.0000
87 B 00e-01	1.085926e+00	7.183712e-01	2.739652e-05	1.0000
88 B 00e-01	1.085926e+00	7.183722e-01	2.443016e-05	1.0000
89 B 00e-01	1.085926e+00	7.183731e-01	2.178509e-05	1.0000
90 B 00e-01	1.085926e+00	7.183739e-01	1.942652e-05	1.0000
91 B 00e-01	1.085926e+00	7.183746e-01	1.732338e-05	1.0000
92 B 00e-01	1.085926e+00	7.183753e-01	1.544801e-05	1.0000
93 B 00e-01	1.085926e+00	7.183758e-01	1.377573e-05	1.0000
94 B 00e-01	1.085926e+00	7.183764e-01	1.228454e-05	1.0000

		Homework 3		
95 E 00e-01	3 1.085926e+00	7.183768e-01	1.095482e-05	1.0000
96 E 00e-01	1.085926e+00	7.183772e-01	9.769076e-06	1.0000
97 E 00e-01	3 1.085926e+00	7.183776e-01	8.711718e-06	1.0000
98 E 00e-01	3 1.085926e+00	7.183779e-01	7.768840e-06	1.0000
99 E 00e-01	3 1.085926e+00	7.183782e-01	6.928041e-06	1.0000
100 E 00e-01	3 1.085926e+00	7.183785e-01	6.178267e-06	1.0000
101 F 00e-01	1.085926e+00	7.183787e-01	5.509660e-06	1.0000
102 F 00e-01	1.085926e+00	7.183789e-01	4.913430e-06	1.0000
103 E 00e-01	1.085926e+00	7.183791e-01	4.381739e-06	1.0000
104 F 00e-01	1.085926e+00	7.183792e-01	3.907600e-06	1.0000
105 E 00e-01	1.085926e+00	7.183794e-01	3.484781e-06	1.0000
106 E 00e-01	1.085926e+00	7.183795e-01	3.107725e-06	1.0000
107 F 00e-01	1.085926e+00	7.183796e-01	2.771477e-06	1.0000
108 F 00e-01	1.085926e+00	7.183797e-01	2.471620e-06	1.0000
109 E 00e-01	1.085926e+00	7.183798e-01	2.204215e-06	1.0000
110 F 00e-01	1.085926e+00	7.183799e-01	1.965747e-06	1.0000
111 F 00e-01	1.085926e+00	7.183800e-01	1.753085e-06	1.0000
112 F 00e-01	1.085926e+00	7.183801e-01	1.563435e-06	1.0000
113 F 00e-01	3 1.085926e+00	7.183801e-01	1.394306e-06	1.0000

		Homework 3		
114 B 00e-01	1.085926e+00	7.183802e-01	1.243478e-06	1.0000
115 В 00e-01	1.085926e+00	7.183802e-01	1.108969e-06	1.0000
116 B 00e-01	1.085926e+00	7.183803e-01	9.890137e-07	1.0000
117 В 00e-01	1.085926e+00	7.183803e-01	8.820366e-07	1.0000
118 B 00e-01	1.085926e+00	7.183803e-01	7.866333e-07	1.0000
119 В 00e-01	1.085926e+00	7.183804e-01	7.015513e-07	1.0000
120 в 00e-01	1.085926e+00	7.183804e-01	6.256736e-07	1.0000
121 В 00e-01	1.085926e+00	7.183804e-01	5.580044e-07	1.0000
122 В 00e-01	1.085926e+00	7.183804e-01	4.976554e-07	1.0000
123 В 00e-01	1.085926e+00	7.183804e-01	4.438346e-07	1.0000
124 В 00e-01	1.085926e+00	7.183805e-01	3.958356e-07	1.0000
125 В 00e-01	1.085926e+00	7.183805e-01	3.530285e-07	1.0000
126 В 00e-01	1.085926e+00	7.183805e-01	3.148516e-07	1.0000
127 В 00e-01	1.085926e+00	7.183805e-01	2.808039e-07	1.0000
128 В 00e-01	1.085926e+00	7.183805e-01	2.504389e-07	1.0000
129 В 00e-01	1.085926e+00	7.183805e-01	2.233580e-07	1.0000
130 В 00e-01	1.085926e+00	7.183805e-01	1.992059e-07	1.0000
131 B 00e-01	1.085926e+00	7.183805e-01	1.776660e-07	1.0000
132 B 00e-01	1.085926e+00	7.183805e-01	1.584555e-07	1.0000

		Homework 3		
133 B 00e-01	1.085926e+00	7.183805e-01	1.413225e-07	1.0000
134 В 00e-01	1.085926e+00	7.183806e-01	1.260424e-07	1.0000
135 B 00e-01	1.085926e+00	7.183806e-01	1.124147e-07	1.0000
136 B 00e-01	1.085926e+00	7.183806e-01	1.002606e-07	1.0000
137 B 00e-01	1.085926e+00	7.183806e-01	8.942080e-08	1.0000
138 B 00e-01	1.085926e+00	7.183806e-01	7.975316e-08	1.0000
139 В 00e-01	1.085926e+00	7.183806e-01	7.113088e-08	1.0000
140 В 00e-01	1.085926e+00	7.183806e-01	6.344091e-08	1.0000
141 B 00e-01	1.085926e+00	7.183806e-01	5.658243e-08	1.0000
142 B 00e-01	1.085926e+00	7.183806e-01	5.046552e-08	1.0000
143 B 00e-01	1.085926e+00	7.183806e-01	4.500997e-08	1.0000
144 B 00e-01	1.085926e+00	7.183806e-01	4.014428e-08	1.0000
145 B 00e-01	1.085926e+00	7.183806e-01	3.580466e-08	1.0000
146 B 00e-01	1.085926e+00	7.183806e-01	3.193421e-08	1.0000
147 B 00e-01	1.085926e+00	7.183806e-01	2.848221e-08	1.0000
148 B 00e-01	1.085926e+00	7.183806e-01	2.540341e-08	1.0000
149 B 00e-01	1.085926e+00	7.183806e-01	2.265746e-08	1.0000
150 B 00e-01	1.085926e+00	7.183806e-01	2.020837e-08	1.0000
151 B 00e-01	1.085926e+00	7.183806e-01	1.802403e-08	1.0000

		Homework 3		
152 B 00e-01	1.085926e+00	7.183806e-01	1.607584e-08	1.0000
153 В 00e-01	1.085926e+00	7.183806e-01	1.433824e-08	1.0000
154 В 00e-01	1.085926e+00	7.183806e-01	1.278848e-08	1.0000
155 B 00e-01	1.085926e+00	7.183806e-01	1.140625e-08	1.0000
156 В 00e-01	1.085926e+00	7.183806e-01	1.017343e-08	1.0000
157 В 00e-01	1.085926e+00	7.183806e-01	9.073870e-09	1.0000
158 В 00e-01	1.085926e+00	7.183806e-01	8.093167e-09	1.0000
159 В 00e-01	1.085926e+00	7.183806e-01	7.218470e-09	1.0000
160 В 00e-01	1.085926e+00	7.183806e-01	6.438319e-09	1.0000
161 B 00e-01	1.085926e+00	7.183806e-01	5.742493e-09	1.0000
162 C 00e-01	1.085926e+00	7.183806e-01	7.183806e-01	1.0000
163 C 00e-01	1.037223e+00	6.427308e-01	6.376026e-01	1.0000
164 C 00e-01	9.988486e-01	5.854215e-01	5.661753e-01	1.0000
165 C 00e-01	9.685836e-01	5.434972e-01	5.029853e-01	1.0000
166 C 00e-01	9.446921e-01	5.141746e-01	4.470554e-01	1.0000
167 C 00e-01	9.258143e-01	4.948568e-01	3.975270e-01	1.0000
168 C 00e-01	9.108846e-01	4.832055e-01	3.536457e-01	1.0000
169 C 00e-01	8.990664e-01	4.772149e-01	3.147485e-01	1.0000
170 C 00e-01	8.897031e-01	4.752488e-01	2.802523e-01	1.0000

		Homework 3		
171 C 00e-01	8.822783e-01	4.760308e-01	2.496441e-01	1.0000
172 C 00e-01	8.763855e-01	4.786020e-01	2.224724e-01	1.0000
173 C 00e-01	8.717048e-01	4.822644e-01	1.983398e-01	1.0000
174 C 00e-01	8.679837e-01	4.865230e-01	1.768960e-01	1.0000
175 C 00e-01	8.650233e-01	4.910354e-01	1.578324e-01	1.0000
176 C 00e-01	8.626661e-01	4.955710e-01	1.408769e-01	1.0000
177 C 00e-01	8.607878e-01	4.999797e-01	1.257895e-01	1.0000
178 C 00e-01	8.592900e-01	5.041679e-01	1.123582e-01	1.0000
179 C 00e-01	8.580948e-01	5.080821e-01	1.003960e-01	1.0000
180 C 00e-01	8.571404e-01	5.116960e-01	8.973757e-02	1.0000
181 C 00e-01	8.563778e-01	5.150017e-01	8.023677e-02	1.0000
182 C 00e-01	8.557680e-01	5.180037e-01	7.176438e-02	1.0000
183 C 00e-01	8.552802e-01	5.207142e-01	6.420603e-02	1.0000
184 C 00e-01	8.548896e-01	5.231500e-01	5.746046e-02	1.0000
185 C 00e-01	8.545767e-01	5.253307e-01	5.143799e-02	1.0000
186 C 00e-01	8.543260e-01	5.272767e-01	4.605912e-02	1.0000
187 C 00e-01	8.541249e-01	5.290088e-01	4.125336e-02	1.0000
188 C 00e-01	8.539636e-01	5.305468e-01	3.695816e-02	1.0000
189 C 00e-01	8.538341e-01	5.319100e-01	3.311800e-02	1.0000

		Homework 3		
190 C 00e-01	8.537301e-01	5.331163e-01	2.968357e-02	1.0000
191 C 00e-01	8.536466e-01	5.341820e-01	2.661106e-02	1.0000
192 C 00e-01	8.535794e-01	5.351225e-01	2.386151e-02	1.0000
193 C 00e-01	8.535254e-01	5.359516e-01	2.140026e-02	1.0000
194 C 00e-01	8.534820e-01	5.366816e-01	1.919649e-02	1.0000
195 C 00e-01	8.534470e-01	5.373240e-01	1.722274e-02	1.0000
196 C 00e-01	8.534189e-01	5.378888e-01	1.545456e-02	1.0000
197 C 00e-01	8.533962e-01	5.383850e-01	1.387016e-02	1.0000
198 C 00e-01	8.533780e-01	5.388208e-01	1.245010e-02	1.0000
199 C 00e-01	8.533633e-01	5.392032e-01	1.117707e-02	1.0000
200 C 00e-01	8.533514e-01	5.395386e-01	1.003560e-02	1.0000
201 C 00e-01	8.533418e-01	5.398327e-01	9.011892e-03	1.0000
202 C 00e-01	8.533341e-01	5.400904e-01	8.093620e-03	1.0000
203 C 00e-01	8.533279e-01	5.403162e-01	7.269777e-03	1.0000
204 C 00e-01	8.533229e-01	5.405140e-01	6.530526e-03	1.0000
205 C 00e-01	8.533189e-01	5.406872e-01	5.867071e-03	1.0000
206 C 00e-01	8.533156e-01	5.408388e-01	5.271548e-03	1.0000
207 C 00e-01	8.533130e-01	5.409714e-01	4.736924e-03	1.0000
208 C 00e-01	8.533108e-01	5.410874e-01	4.256903e-03	1.0000

		Homework 3		
209 C 00e-01	8.533091e-01	5.411890e-01	3.825851e-03	1.0000
210 C 00e-01	8.533077e-01	5.412777e-01	3.438724e-03	1.0000
211 C 00e-01	8.533066e-01	5.413553e-01	3.091005e-03	1.0000
212 C 00e-01	8.533057e-01	5.414232e-01	2.778646e-03	1.0000
213 C 00e-01	8.533050e-01	5.414825e-01	2.498021e-03	1.0000
214 C 00e-01	8.533044e-01	5.415344e-01	2.245882e-03	1.0000
215 C 00e-01	8.533039e-01	5.415797e-01	2.019315e-03	1.0000
216 C 00e-01	8.533035e-01	5.416193e-01	1.815707e-03	1.0000
217 C 00e-01	8.533032e-01	5.416539e-01	1.632718e-03	1.0000
218 C 00e-01	8.533029e-01	5.416841e-01	1.468244e-03	1.0000
219 C 00e-01	8.533027e-01	5.417105e-01	1.320403e-03	1.0000
220 C 00e-01	8.533026e-01	5.417335e-01	1.187501e-03	1.0000
221 C 00e-01	8.533024e-01	5.417537e-01	1.068022e-03	1.0000
222 C 00e-01	8.533023e-01	5.417713e-01	9.606028e-04	1.0000
223 C 00e-01	8.533022e-01	5.417866e-01	8.640201e-04	1.0000
224 C 00e-01	8.533022e-01	5.418000e-01	7.771759e-04	1.0000
225 C 00e-01	8.533021e-01	5.418117e-01	6.990840e-04	1.0000
226 C 00e-01	8.533021e-01	5.418220e-01	6.288588e-04	1.0000
227 C 00e-01	8.533020e-01	5.418309e-01	5.657048e-04	1.0000

		Homework 3		
228 C 00e-01	8.533020e-01	5.418387e-01	5.089074e-04	1.0000
229 C 00e-01	8.533020e-01	5.418455e-01	4.578246e-04	1.0000
230 C 00e-01	8.533019e-01	5.418514e-01	4.118796e-04	1.0000
231 C 00e-01	8.533019e-01	5.418566e-01	3.705541e-04	1.0000
232 C 00e-01	8.533019e-01	5.418611e-01	3.333823e-04	1.0000
233 C 00e-01	8.533019e-01	5.418651e-01	2.999455e-04	1.0000
234 C 00e-01	8.533019e-01	5.418685e-01	2.698676e-04	1.0000
235 C 00e-01	8.533019e-01	5.418716e-01	2.428103e-04	1.0000
236 C 00e-01	8.533019e-01	5.418742e-01	2.184696e-04	1.0000
237 C 00e-01	8.533019e-01	5.418765e-01	1.965721e-04	1.0000
238 C 00e-01	8.533019e-01	5.418785e-01	1.768721e-04	1.0000
239 C 00e-01	8.533019e-01	5.418802e-01	1.591488e-04	1.0000
240 C 00e-01	8.533019e-01	5.418817e-01	1.432033e-04	1.0000
241 C 00e-01	8.533019e-01	5.418831e-01	1.288571e-04	1.0000
242 C 00e-01	8.533019e-01	5.418842e-01	1.159494e-04	1.0000
243 C 00e-01	8.533019e-01	5.418853e-01	1.043360e-04	1.0000
244 C 00e-01	8.533019e-01	5.418861e-01	9.388667e-05	1.0000
245 C 00e-01	8.533019e-01	5.418869e-01	8.448473e-05	1.0000
246 C 00e-01	8.533019e-01	5.418876e-01	7.602503e-05	1.0000

		Homework 3		
247 C 00e-01	8.533019e-01	5.418882e-01	6.841302e-05	1.0000
248 C 00e-01	8.533019e-01	5.418887e-01	6.156368e-05	1.0000
249 C 00e-01	8.533019e-01	5.418891e-01	5.540051e-05	1.0000
250 C 00e-01	8.533019e-01	5.418895e-01	4.985471e-05	1.0000
251 C 00e-01	8.533019e-01	5.418899e-01	4.486437e-05	1.0000
252 C 00e-01	8.533019e-01	5.418902e-01	4.037381e-05	1.0000
253 C 00e-01	8.533019e-01	5.418904e-01	3.633294e-05	1.0000
254 C 00e-01	8.533019e-01	5.418906e-01	3.269670e-05	1.0000
255 C 00e-01	8.533019e-01	5.418908e-01	2.942454e-05	1.0000
256 C 00e-01	8.533019e-01	5.418910e-01	2.647997e-05	1.0000
257 C 00e-01	8.533019e-01	5.418912e-01	2.383019e-05	1.0000
258 C 00e-01	8.533019e-01	5.418913e-01	2.144566e-05	1.0000
259 C 00e-01	8.533019e-01	5.418914e-01	1.929982e-05	1.0000
260 C 00e-01	8.533019e-01	5.418915e-01	1.736875e-05	1.0000
261 C 00e-01	8.533019e-01	5.418916e-01	1.563096e-05	1.0000
262 C 00e-01	8.533019e-01	5.418917e-01	1.406709e-05	1.0000
263 C 00e-01	8.533019e-01	5.418917e-01	1.265973e-05	1.0000
264 C 00e-01	8.533019e-01	5.418918e-01	1.139320e-05	1.0000
265 C 00e-01	8.533019e-01	5.418919e-01	1.025341e-05	1.0000

		Homework 3		
266 C 00e-01	8.533019e-01	5.418919e-01	9.227676e-06	1.0000
267 C 00e-01	8.533019e-01	5.418919e-01	8.304573e-06	1.0000
268 C 00e-01	8.533019e-01	5.418920e-01	7.473832e-06	1.0000
269 C 00e-01	8.533019e-01	5.418920e-01	6.726208e-06	1.0000
270 C 00e-01	8.533019e-01	5.418920e-01	6.053385e-06	1.0000
271 C 00e-01	8.533019e-01	5.418920e-01	5.447874e-06	1.0000
272 C 00e-01	8.533019e-01	5.418921e-01	4.902942e-06	1.0000
273 C 00e-01	8.533019e-01	5.418921e-01	4.412524e-06	1.0000
274 C 00e-01	8.533019e-01	5.418921e-01	3.971168e-06	1.0000
275 C 00e-01	8.533019e-01	5.418921e-01	3.573963e-06	1.0000
276 C 00e-01	8.533019e-01	5.418921e-01	3.216492e-06	1.0000
277 C 00e-01	8.533019e-01	5.418921e-01	2.894780e-06	1.0000
278 C 00e-01	8.533019e-01	5.418921e-01	2.605249e-06	1.0000
279 C 00e-01	8.533019e-01	5.418921e-01	2.344679e-06	1.0000
280 C 00e-01	8.533019e-01	5.418922e-01	2.110173e-06	1.0000
281 C 00e-01	8.533019e-01	5.418922e-01	1.899123e-06	1.0000
282 C 00e-01	8.533019e-01	5.418922e-01	1.709184e-06	1.0000
283 C 00e-01	8.533019e-01	5.418922e-01	1.538242e-06	1.0000
284 C 00e-01	8.533019e-01	5.418922e-01	1.384398e-06	1.0000

		Homework 3		
285 C 00e-01	8.533019e-01	5.418922e-01	1.245942e-06	1.0000
286 C 00e-01	8.533019e-01	5.418922e-01	1.121334e-06	1.0000
287 C 00e-01	8.533019e-01	5.418922e-01	1.009189e-06	1.0000
288 C 00e-01	8.533019e-01	5.418922e-01	9.082598e-07	1.0000
289 C 00e-01	8.533019e-01	5.418922e-01	8.174254e-07	1.0000
290 C 00e-01	8.533019e-01	5.418922e-01	7.356757e-07	1.0000
291 C 00e-01	8.533019e-01	5.418922e-01	6.621021e-07	1.0000
292 C 00e-01	8.533019e-01	5.418922e-01	5.958867e-07	1.0000
293 C 00e-01	8.533019e-01	5.418922e-01	5.362937e-07	1.0000
294 C 00e-01	8.533019e-01	5.418922e-01	4.826607e-07	1.0000
295 C 00e-01	8.533019e-01	5.418922e-01	4.343915e-07	1.0000
296 C 00e-01	8.533019e-01	5.418922e-01	3.909497e-07	1.0000
297 C 00e-01	8.533019e-01	5.418922e-01	3.518525e-07	1.0000
298 C 00e-01	8.533019e-01	5.418922e-01	3.166654e-07	1.0000
299 C 00e-01	8.533019e-01	5.418922e-01	2.849973e-07	1.0000
300 C 00e-01	8.533019e-01	5.418922e-01	2.564962e-07	1.0000
301 C 00e-01	8.533019e-01	5.418922e-01	2.308455e-07	1.0000
302 C 00e-01	8.533019e-01	5.418922e-01	2.077599e-07	1.0000
303 C 00e-01	8.533019e-01	5.418922e-01	1.869831e-07	1.0000

		Homework 3		
304 C 00e-01	8.533019e-01	5.418922e-01	1.682841e-07	1.0000
305 C 00e-01	8.533019e-01	5.418922e-01	1.514551e-07	1.0000
306 C 00e-01	8.533019e-01	5.418922e-01	1.363091e-07	1.0000
307 C 00e-01	8.533019e-01	5.418922e-01	1.226778e-07	1.0000
308 C 00e-01	8.533019e-01	5.418922e-01	1.104097e-07	1.0000
309 C 00e-01	8.533019e-01	5.418922e-01	9.936840e-08	1.0000
310 C 00e-01	8.533019e-01	5.418922e-01	8.943131e-08	1.0000
311 C 00e-01	8.533019e-01	5.418922e-01	8.048797e-08	1.0000
312 C 00e-01	8.533019e-01	5.418922e-01	7.243899e-08	1.0000
313 C 00e-01	8.533019e-01	5.418922e-01	6.519494e-08	1.0000
314 C 00e-01	8.533019e-01	5.418922e-01	5.867531e-08	1.0000
315 C 00e-01	8.533019e-01	5.418922e-01	5.280767e-08	1.0000
316 C 00e-01	8.533019e-01	5.418922e-01	4.752681e-08	1.0000
317 C 00e-01	8.533019e-01	5.418922e-01	4.277405e-08	1.0000
318 C 00e-01	8.533019e-01	5.418922e-01	3.849658e-08	1.0000
319 C 00e-01	8.533019e-01	5.418922e-01	3.464687e-08	1.0000
320 C 00e-01	8.533019e-01	5.418922e-01	3.118213e-08	1.0000
321 C 00e-01	8.533019e-01	5.418922e-01	2.806388e-08	1.0000
322 C 00e-01	8.533019e-01	5.418922e-01	2.525746e-08	1.0000

		Homework 3		
323 C 00e-01	8.533019e-01	5.418922e-01	2.273168e-08	1.0000
324 C 00e-01	8.533019e-01	5.418922e-01	2.045849e-08	1.0000
325 C 00e-01	8.533019e-01	5.418922e-01	1.841262e-08	1.0000
326 C 00e-01	8.533019e-01	5.418922e-01	1.657134e-08	1.0000
327 C 00e-01	8.533019e-01	5.418922e-01	1.491419e-08	1.0000
328 C 00e-01	8.533019e-01	5.418922e-01	1.342276e-08	1.0000
329 C 00e-01	8.533019e-01	5.418922e-01	1.208047e-08	1.0000
330 C 00e-01	8.533019e-01	5.418922e-01	1.087242e-08	1.0000
331 C 00e-01	8.533019e-01	5.418922e-01	9.785169e-09	1.0000
332 C 00e-01	8.533019e-01	5.418922e-01	8.806646e-09	1.0000
333 C 00e-01	8.533019e-01	5.418922e-01	7.925976e-09	1.0000
334 C 00e-01	8.533019e-01	5.418922e-01	7.133374e-09	1.0000
335 B 00e-01	8.533019e-01	5.418922e-01	5.418922e-01	1.0000
336 B 00e-01	8.263975e-01	4.548365e-01	4.510846e-01	1.0000
337 В 00e-01	8.077547e-01	3.890027e-01	3.754941e-01	1.0000
338 B 00e-01	7.948365e-01	3.401139e-01	3.125707e-01	1.0000
339 B 00e-01	7.858850e-01	3.047162e-01	2.601916e-01	1.0000
340 B 00e-01	7.796823e-01	2.798856e-01	2.165900e-01	1.0000
341 B 00e-01	7.753842e-01	2.631022e-01	1.802950e-01	1.0000

		Homework 3		
342 B 00e-01	7.724060e-01	2.522341e-01	1.500820e-01	1.0000
343 B 00e-01	7.703422e-01	2.455534e-01	1.249320e-01	1.0000
344 B 00e-01	7.689122e-01	2.417296e-01	1.039965e-01	1.0000
345 B 00e-01	7.679213e-01	2.397870e-01	8.656932e-02	1.0000
346 B 00e-01	7.672347e-01	2.390386e-01	7.206247e-02	1.0000
347 B 00e-01	7.667589e-01	2.390164e-01	5.998660e-02	1.0000
348 B 00e-01	7.664292e-01	2.394107e-01	4.993435e-02	1.0000
349 B 00e-01	7.662007e-01	2.400214e-01	4.156660e-02	1.0000
350 B 00e-01	7.660424e-01	2.407227e-01	3.460107e-02	1.0000
351 B 00e-01	7.659328e-01	2.414384e-01	2.880280e-02	1.0000
352 B 00e-01	7.658567e-01	2.421244e-01	2.397617e-02	1.0000
353 B 00e-01	7.658041e-01	2.427571e-01	1.995836e-02	1.0000
354 B 00e-01	7.657676e-01	2.433261e-01	1.661384e-02	1.0000
355 B 00e-01	7.657423e-01	2.438288e-01	1.382978e-02	1.0000
356 B 00e-01	7.657248e-01	2.442672e-01	1.151225e-02	1.0000
357 B 00e-01	7.657126e-01	2.446459e-01	9.583086e-03	1.0000
358 B 00e-01	7.657042e-01	2.449706e-01	7.977200e-03	1.0000
359 B 00e-01	7.656984e-01	2.452474e-01	6.640421e-03	1.0000
360 B 00e-01	7.656943e-01	2.454823e-01	5.527652e-03	1.0000

		Homework 3		
361 B 00e-01	7.656915e-01	2.456810e-01	4.601356e-03	1.0000
362 B 00e-01	7.656896e-01	2.458485e-01	3.830284e-03	1.0000
363 B 00e-01	7.656883e-01	2.459894e-01	3.188424e-03	1.0000
364 B 00e-01	7.656873e-01	2.461077e-01	2.654124e-03	1.0000
365 B 00e-01	7.656867e-01	2.462069e-01	2.209359e-03	1.0000
366 B 00e-01	7.656862e-01	2.462900e-01	1.839126e-03	1.0000
367 B 00e-01	7.656859e-01	2.463595e-01	1.530934e-03	1.0000
368 B 00e-01	7.656857e-01	2.464176e-01	1.274388e-03	1.0000
369 B 00e-01	7.656856e-01	2.464661e-01	1.060832e-03	1.0000
370 B 00e-01	7.656855e-01	2.465066e-01	8.830633e-04	1.0000
371 B 00e-01	7.656854e-01	2.465404e-01	7.350840e-04	1.0000
372 B 00e-01	7.656853e-01	2.465686e-01	6.119023e-04	1.0000
373 B 00e-01	7.656853e-01	2.465921e-01	5.093627e-04	1.0000
374 B 00e-01	7.656853e-01	2.466116e-01	4.240063e-04	1.0000
375 B 00e-01	7.656853e-01	2.466280e-01	3.529534e-04	1.0000
376 B 00e-01	7.656852e-01	2.466416e-01	2.938072e-04	1.0000
377 B 00e-01	7.656852e-01	2.466529e-01	2.445725e-04	1.0000
378 B 00e-01	7.656852e-01	2.466623e-01	2.035883e-04	1.0000
379 B 00e-01	7.656852e-01	2.466702e-01	1.694720e-04	1.0000

		Homework 3		
380 B 00e-01	7.656852e-01	2.466767e-01	1.410727e-04	1.0000
381 B 00e-01	7.656852e-01	2.466822e-01	1.174324e-04	1.0000
382 B 00e-01	7.656852e-01	2.466867e-01	9.775369e-05	1.0000
383 B 00e-01	7.656852e-01	2.466905e-01	8.137262e-05	1.0000
384 B 00e-01	7.656852e-01	2.466936e-01	6.773660e-05	1.0000
385 B 00e-01	7.656852e-01	2.466962e-01	5.638564e-05	1.0000
386 B 00e-01	7.656852e-01	2.466984e-01	4.693682e-05	1.0000
387 B 00e-01	7.656852e-01	2.467002e-01	3.907138e-05	1.0000
388 B 00e-01	7.656852e-01	2.467018e-01	3.252399e-05	1.0000
389 B 00e-01	7.656852e-01	2.467030e-01	2.707378e-05	1.0000
390 B 00e-01	7.656852e-01	2.467041e-01	2.253689e-05	1.0000
391 B 00e-01	7.656852e-01	2.467049e-01	1.876027e-05	1.0000
392 B 00e-01	7.656852e-01	2.467057e-01	1.561652e-05	1.0000
393 B 00e-01	7.656852e-01	2.467063e-01	1.299958e-05	1.0000
394 B 00e-01	7.656852e-01	2.467068e-01	1.082118e-05	1.0000
395 B 00e-01	7.656852e-01	2.467072e-01	9.007819e-06	1.0000
396 B 00e-01	7.656852e-01	2.467075e-01	7.498333e-06	1.0000
397 B 00e-01	7.656852e-01	2.467078e-01	6.241800e-06	1.0000
398 B 00e-01	7.656852e-01	2.467081e-01	5.195830e-06	1.0000

		Homework 3		
399 B 00e-01	7.656852e-01	2.467083e-01	4.325139e-06	1.0000
400 B 00e-01	7.656852e-01	2.467084e-01	3.600354e-06	1.0000
401 B 00e-01	7.656852e-01	2.467086e-01	2.997025e-06	1.0000
402 B 00e-01	7.656852e-01	2.467087e-01	2.494798e-06	1.0000
403 B 00e-01	7.656852e-01	2.467088e-01	2.076732e-06	1.0000
404 B 00e-01	7.656852e-01	2.467089e-01	1.728724e-06	1.0000
405 B 00e-01	7.656852e-01	2.467089e-01	1.439033e-06	1.0000
406 B 00e-01	7.656852e-01	2.467090e-01	1.197887e-06	1.0000
407 B 00e-01	7.656852e-01	2.467090e-01	9.971511e-07	1.0000
408 B 00e-01	7.656852e-01	2.467091e-01	8.300535e-07	1.0000
409 B 00e-01	7.656852e-01	2.467091e-01	6.909573e-07	1.0000
410 B 00e-01	7.656852e-01	2.467091e-01	5.751701e-07	1.0000
411 B 00e-01	7.656852e-01	2.467092e-01	4.787860e-07	1.0000
412 B 00e-01	7.656852e-01	2.467092e-01	3.985534e-07	1.0000
413 B 00e-01	7.656852e-01	2.467092e-01	3.317658e-07	1.0000
414 B 00e-01	7.656852e-01	2.467092e-01	2.761702e-07	1.0000
415 B 00e-01	7.656852e-01	2.467092e-01	2.298910e-07	1.0000
416 B 00e-01	7.656852e-01	2.467092e-01	1.913670e-07	1.0000
417 B 00e-01	7.656852e-01	2.467092e-01	1.592987e-07	1.0000

		Homework 3		
418 B 00e-01	7.656852e-01	2.467092e-01	1.326042e-07	1.0000
419 B 00e-01	7.656852e-01	2.467092e-01	1.103830e-07	1.0000
420 B 00e-01	7.656852e-01	2.467092e-01	9.188560e-08	1.0000
421 B 00e-01	7.656852e-01	2.467092e-01	7.648787e-08	1.0000
422 B 00e-01	7.656852e-01	2.467092e-01	6.367041e-08	1.0000
423 B 00e-01	7.656852e-01	2.467093e-01	5.300084e-08	1.0000
424 B 00e-01	7.656852e-01	2.467093e-01	4.411923e-08	1.0000
425 B 00e-01	7.656852e-01	2.467093e-01	3.672595e-08	1.0000
426 B 00e-01	7.656852e-01	2.467093e-01	3.057160e-08	1.0000
427 В 00e-01	7.656852e-01	2.467093e-01	2.544856e-08	1.0000
428 B 00e-01	7.656852e-01	2.467093e-01	2.118402e-08	1.0000
429 B 00e-01	7.656852e-01	2.467093e-01	1.763411e-08	1.0000
430 в 00e-01	7.656852e-01	2.467093e-01	1.467907e-08	1.0000
431 B 00e-01	7.656852e-01	2.467093e-01	1.221923e-08	1.0000
432 B 00e-01	7.656852e-01	2.467093e-01	1.017159e-08	1.0000
433 B 00e-01	7.656852e-01	2.467093e-01	8.467085e-09	1.0000
434 B 00e-01	7.656852e-01	2.467093e-01	7.048213e-09	1.0000
435 B 00e-01	7.656852e-01	2.467093e-01	5.867109e-09	1.0000
436 B 00e-01	7.656852e-01	2.467093e-01	4.883928e-09	1.0000

		Homework 3		
437 C 00e-01	7.656852e-01	2.467093e-01	2.467093e-01	1.0000
438 C 00e-01	7.601959e-01	1.984820e-01	1.982911e-01	1.0000
439 C 00e-01	7.566498e-01	1.600311e-01	1.593753e-01	1.0000
440 C 00e-01	7.543590e-01	1.294004e-01	1.280969e-01	1.0000
441 C 00e-01	7.528792e-01	1.050574e-01	1.029572e-01	1.0000
442 C 00e-01	7.519232e-01	8.579409e-02	8.275120e-02	1.0000
443 C 00e-01	7.513056e-01	7.065277e-02	6.651079e-02	1.0000
444 C 00e-01	7.509066e-01	5.886857e-02	5.345765e-02	1.0000
445 C 00e-01	7.506489e-01	4.982205e-02	4.296627e-02	1.0000
446 C 00e-01	7.504824e-01	4.300034e-02	3.453388e-02	1.0000
447 C 00e-01	7.503748e-01	3.796687e-02	2.775640e-02	1.0000
448 C 00e-01	7.503054e-01	3.434207e-02	2.230904e-02	1.0000
449 C 00e-01	7.502605e-01	3.179587e-02	1.793076e-02	1.0000
450 C 00e-01	7.502315e-01	3.004865e-02	1.441174e-02	1.0000
451 C 00e-01	7.502127e-01	2.887387e-02	1.158335e-02	1.0000
452 C 00e-01	7.502006e-01	2.809715e-02	9.310050e-03	1.0000
453 C 00e-01	7.501928e-01	2.759054e-02	7.482897e-03	1.0000
454 C 00e-01	7.501878e-01	2.726378e-02	6.014334e-03	1.0000
455 C 00e-01	7.501845e-01	2.705504e-02	4.833985e-03	1.0000

456 C ODDE-OI 7.501824e-01 2.692293e-02 3.885287e-03 1.0000 457 C ODDE-OI 7.501811e-01 2.684014e-02 3.122776e-03 1.0000 458 C ODDE-OI 7.501802e-01 2.678886e-02 2.509913e-03 1.0000 459 C ODDE-OI 7.501796e-01 2.675755e-02 2.017327e-03 1.0000 460 C ODDE-OI 7.501792e-01 2.672793e-02 1.303203e-03 1.0000 461 C ODDE-OI 7.501780e-01 2.672186e-02 1.047441e-03 1.0000 462 C ODDE-OI 7.501787e-01 2.671873e-02 8.418747e-04 1.0000 463 C ODDE-OI 7.501787e-01 2.671873e-02 8.418747e-04 1.0000 465 C ODDE-OI 7.501786e-01 2.671694e-02 5.438549e-04 1.0000 466 C ODDE-OI 7.501786e-01 2.671710e-02 4.371201e-04 1.0000 467 C ODDE-OI 7.501786e-01 2.671806e-02 2.823816e-04 1.0000 468 C ODDE-OI 7.501786e-01 2.671806e-02 2.269625e-04 1.0000 469 C ODDE-OI 7.501786e-01			Homework 3		
00e-01 458		7.501824e-01	2.692293e-02	3.885287e-03	1.0000
00e-01 459		7.501811e-01	2.684014e-02	3.122776e-03	1.0000
00e-01 460		7.501802e-01	2.678886e-02	2.509913e-03	1.0000
00e-01 461		7.501796e-01	2.675756e-02	2.017327e-03	1.0000
00e-01 462		7.501792e-01	2.673882e-02	1.621415e-03	1.0000
00e-01 463		7.501790e-01	2.672793e-02	1.303203e-03	1.0000
00e-01 464		7.501788e-01	2.672186e-02	1.047441e-03	1.0000
00e-01 465		7.501787e-01	2.671873e-02	8.418747e-04	1.0000
00e-01 466		7.501787e-01	2.671734e-02	6.766518e-04	1.0000
00e-01 467		7.501786e-01	2.671694e-02	5.438549e-04	1.0000
00e-01 468		7.501786e-01	2.671710e-02	4.371201e-04	1.0000
00e-01 469 C 00e-01 7.501786e-01 2.671862e-02 2.269625e-04 1.0000 470 C 00e-01 7.501786e-01 2.671915e-02 1.824198e-04 1.0000 471 C 00e-01 7.501786e-01 2.671963e-02 1.466188e-04 1.0000 472 C 00e-01 7.501786e-01 2.672005e-02 1.178440e-04 1.0000 473 C 00e-01 7.501786e-01 2.672041e-02 9.471640e-05 1.0000 474 C 7.501786e-01 2.672071e-02 7.612774e-05 1.0000		7.501786e-01	2.671752e-02	3.513327e-04	1.0000
00e-01 470		7.501786e-01	2.671806e-02	2.823816e-04	1.0000
00e-01 471		7.501786e-01	2.671862e-02	2.269625e-04	1.0000
00e-01 472		7.501786e-01	2.671915e-02	1.824198e-04	1.0000
00e-01 473 C 7.501786e-01 2.672041e-02 9.471640e-05 1.0000 00e-01 474 C 7.501786e-01 2.672071e-02 7.612774e-05 1.0000		7.501786e-01	2.671963e-02	1.466188e-04	1.0000
00e-01 474 C 7.501786e-01 2.672071e-02 7.612774e-05 1.0000		7.501786e-01	2.672005e-02	1.178440e-04	1.0000
		7.501786e-01	2.672041e-02	9.471640e-05	1.0000
	-	7.501786e-01	2.672071e-02	7.612774e-05	1.0000

		Homework 3		
475 C 00e-01	7.501786e-01	2.672097e-02	6.118722e-05	1.0000
476 C 00e-01	7.501786e-01	2.672117e-02	4.917886e-05	1.0000
477 C 00e-01	7.501786e-01	2.672135e-02	3.952722e-05	1.0000
478 C 00e-01	7.501786e-01	2.672149e-02	3.176977e-05	1.0000
479 C 00e-01	7.501786e-01	2.672160e-02	2.553476e-05	1.0000
480 C 00e-01	7.501786e-01	2.672170e-02	2.052341e-05	1.0000
481 C 00e-01	7.501786e-01	2.672177e-02	1.649557e-05	1.0000
482 C 00e-01	7.501786e-01	2.672183e-02	1.325822e-05	1.0000
483 C 00e-01	7.501786e-01	2.672188e-02	1.065621e-05	1.0000
484 C 00e-01	7.501786e-01	2.672192e-02	8.564867e-06	1.0000
485 C 00e-01	7.501786e-01	2.672195e-02	6.883961e-06	1.0000
486 C 00e-01	7.501786e-01	2.672198e-02	5.532943e-06	1.0000
487 C 00e-01	7.501786e-01	2.672200e-02	4.447070e-06	1.0000
488 C 00e-01	7.501786e-01	2.672202e-02	3.574306e-06	1.0000
489 C 00e-01	7.501786e-01	2.672203e-02	2.872827e-06	1.0000
490 C 00e-01	7.501786e-01	2.672204e-02	2.309018e-06	1.0000
491 C 00e-01	7.501786e-01	2.672205e-02	1.855859e-06	1.0000
492 C 00e-01	7.501786e-01	2.672206e-02	1.491636e-06	1.0000
493 C 00e-01	7.501786e-01	2.672206e-02	1.198893e-06	1.0000

		Homework 3		
494 C 00e-01	7.501786e-01	2.672207e-02	9.636034e-07	1.0000
495 C 00e-01	7.501786e-01	2.672207e-02	7.744905e-07	1.0000
496 C 00e-01	7.501786e-01	2.672207e-02	6.224922e-07	1.0000
497 C 00e-01	7.501786e-01	2.672208e-02	5.003244e-07	1.0000
498 C 00e-01	7.501786e-01	2.672208e-02	4.021327e-07	1.0000
499 C 00e-01	7.501786e-01	2.672208e-02	3.232118e-07	1.0000
500 C 00e-01	7.501786e-01	2.672208e-02	2.597796e-07	1.0000
501 C 00e-01	7.501786e-01	2.672208e-02	2.087963e-07	1.0000
502 C 00e-01	7.501786e-01	2.672208e-02	1.678188e-07	1.0000
503 C 00e-01	7.501786e-01	2.672208e-02	1.348833e-07	1.0000
504 C 00e-01	7.501786e-01	2.672208e-02	1.084117e-07	1.0000
505 C 00e-01	7.501786e-01	2.672208e-02	8.713524e-08	1.0000
506 C 00e-01	7.501786e-01	2.672208e-02	7.003443e-08	1.0000
507 C 00e-01	7.501786e-01	2.672209e-02	5.628975e-08	1.0000
508 C 00e-01	7.501786e-01	2.672209e-02	4.524256e-08	1.0000
509 C 00e-01	7.501786e-01	2.672209e-02	3.636343e-08	1.0000
510 C 00e-01	7.501786e-01	2.672209e-02	2.922689e-08	1.0000
511 C 00e-01	7.501786e-01	2.672209e-02	2.349094e-08	1.0000
512 C 00e-01	7.501786e-01	2.672209e-02	1.888071e-08	1.0000

		Homework 3		
513 C 00e-01	7.501786e-01	2.672209e-02	1.517525e-08	1.0000
514 C 00e-01	7.501786e-01	2.672209e-02	1.219702e-08	1.0000
515 C 00e-01	7.501786e-01	2.672209e-02	9.803282e-09	1.0000
516 C 00e-01	7.501786e-01	2.672209e-02	7.879330e-09	1.0000
517 C 00e-01	7.501786e-01	2.672209e-02	6.332965e-09	1.0000
518 C 00e-01	7.501786e-01	2.672209e-02	5.090083e-09	1.0000
519 C 00e-01	7.501786e-01	2.672209e-02	4.091124e-09	1.0000
520 C 00e-01	7.501786e-01	2.672209e-02	3.288216e-09	1.0000
521 C 00e-01	7.501786e-01	2.672209e-02	2.642884e-09	1.0000
522 C 00e-01	7.501786e-01	2.672209e-02	2.124203e-09	1.0000
523 В 00e-01	7.501786e-01	2.672209e-02	2.672209e-02	1.0000
524 B 00e-01	7.501143e-01	2.137884e-02	2.137864e-02	1.0000
525 В 00e-01	7.500732e-01	1.710436e-02	1.710369e-02	1.0000
526 В 00e-01	7.500468e-01	1.368488e-02	1.368358e-02	1.0000
527 В 00e-01	7.500300e-01	1.094943e-02	1.094736e-02	1.0000
528 В 00e-01	7.500192e-01	8.761269e-03	8.758285e-03	1.0000
529 В 00e-01	7.500123e-01	7.011013e-03	7.006947e-03	1.0000
530 в 00e-01	7.500079e-01	5.611174e-03	5.605813e-03	1.0000
531 В 00e-01	7.500050e-01	4.491786e-03	4.484854e-03	1.0000

		Homework 3		
532 В 00e-01	7.500032e-01	3.596897e-03	3.588046e-03	1.0000
533 B 00e-01	7.500021e-01	2.881778e-03	2.870568e-03	1.0000
534 B 00e-01	7.500013e-01	2.310682e-03	2.296559e-03	1.0000
535 B 00e-01	7.500009e-01	1.855058e-03	1.837331e-03	1.0000
536 В 00e-01	7.500006e-01	1.492116e-03	1.469931e-03	1.0000
537 В 00e-01	7.500004e-01	1.203683e-03	1.175999e-03	1.0000
538 B 00e-01	7.500002e-01	9.752822e-04	9.408416e-04	1.0000
539 В 00e-01	7.500002e-01	7.953845e-04	7.527075e-04	1.0000
540 B 00e-01	7.500001e-01	6.547946e-04	6.021934e-04	1.0000
541 B 00e-01	7.500001e-01	5.461344e-04	4.817767e-04	1.0000
542 B 00e-01	7.500001e-01	4.634019e-04	3.854389e-04	1.0000
543 B 00e-01	7.500000e-01	4.016021e-04	3.083651e-04	1.0000
544 B 00e-01	7.500000e-01	3.564684e-04	2.467033e-04	1.0000
545 B 00e-01	7.500000e-01	3.243013e-04	1.973716e-04	1.0000
546 B 00e-01	7.500000e-01	3.019203e-04	1.579045e-04	1.0000
547 B 00e-01	7.500000e-01	2.866805e-04	1.263293e-04	1.0000
548 B 00e-01	7.500000e-01	2.764865e-04	1.010681e-04	1.0000
549 В 00e-01	7.500000e-01	2.697603e-04	8.085813e-05	1.0000
550 В 00e-01	7.500000e-01	2.653663e-04	6.468945e-05	1.0000

		Homework 3		
551 B 00e-01	7.500000e-01	2.625157e-04	5.175391e-05	1.0000
552 B 00e-01	7.500000e-01	2.606752e-04	4.140501e-05	1.0000
553 B 00e-01	7.500000e-01	2.594906e-04	3.312552e-05	1.0000
554 В 00е-01	7.500000e-01	2.587298e-04	2.650162e-05	1.0000
555 В 00e-01	7.500000e-01	2.582418e-04	2.120226e-05	1.0000
556 В 00e-01	7.500000e-01	2.579292e-04	1.696258e-05	1.0000
557 В 00e-01	7.500000e-01	2.577290e-04	1.357068e-05	1.0000
558 В 00e-01	7.500000e-01	2.576009e-04	1.085704e-05	1.0000
559 В 00e-01	7.500000e-01	2.575189e-04	8.686027e-06	1.0000
560 в 00e-01	7.500000e-01	2.574665e-04	6.949137e-06	1.0000
561 В 00e-01	7.500000e-01	2.574330e-04	5.559563e-06	1.0000
562 В 00e-01	7.500000e-01	2.574116e-04	4.447853e-06	1.0000
563 В 00e-01	7.500000e-01	2.573979e-04	3.558444e-06	1.0000
564 В 00e-01	7.500000e-01	2.573892e-04	2.846885e-06	1.0000
565 В 00e-01	7.500000e-01	2.573837e-04	2.277611e-06	1.0000
566 В 00e-01	7.500000e-01	2.573801e-04	1.822172e-06	1.0000
567 В 00e-01	7.500000e-01	2.573779e-04	1.457804e-06	1.0000
568 В 00e-01	7.500000e-01	2.573764e-04	1.166296e-06	1.0000
569 В 00e-01	7.500000e-01	2.573755e-04	9.330794e-07	1.0000

		Homework 3		
570 В 00e-01	7.500000e-01	2.573749e-04	7.464975e-07	1.0000
571 В 00e-01	7.500000e-01	2.573746e-04	5.972251e-07	1.0000
572 В 00e-01	7.500000e-01	2.573743e-04	4.778018e-07	1.0000
573 В 00e-01	7.500000e-01	2.573742e-04	3.822589e-07	1.0000
574 B 00e-01	7.500000e-01	2.573741e-04	3.058210e-07	1.0000
575 В 00e-01	7.500000e-01	2.573741e-04	2.446679e-07	1.0000
576 В 00e-01	7.500000e-01	2.573740e-04	1.957433e-07	1.0000
577 В 00e-01	7.500000e-01	2.573740e-04	1.566017e-07	1.0000
578 В 00e-01	7.500000e-01	2.573740e-04	1.252871e-07	1.0000
579 В 00e-01	7.500000e-01	2.573740e-04	1.002342e-07	1.0000
580 В 00e-01	7.500000e-01	2.573740e-04	8.019103e-08	1.0000
581 B 00e-01	7.500000e-01	2.573740e-04	6.415574e-08	1.0000
582 В 00e-01	7.500000e-01	2.573740e-04	5.132693e-08	1.0000
583 B 00e-01	7.500000e-01	2.573740e-04	4.106341e-08	1.0000
584 B 00e-01	7.500000e-01	2.573740e-04	3.285222e-08	1.0000
585 В 00e-01	7.500000e-01	2.573740e-04	2.628297e-08	1.0000
586 В 00e-01	7.500000e-01	2.573740e-04	2.102734e-08	1.0000
587 В 00e-01	7.500000e-01	2.573740e-04	1.682263e-08	1.0000
588 В 00e-01	7.500000e-01	2.573740e-04	1.345872e-08	1.0000

		Homework 3		
589 В 00e-01	7.500000e-01	2.573740e-04	1.076747e-08	1.0000
590 В 00e-01	7.500000e-01	2.573740e-04	8.614364e-09	1.0000
591 B 00e-01	7.500000e-01	2.573740e-04	6.891805e-09	1.0000
592 В 00e-01	7.500000e-01	2.573740e-04	5.513695e-09	1.0000
593 В 00e-01	7.500000e-01	2.573740e-04	4.411156e-09	1.0000
594 В 00e-01	7.500000e-01	2.573740e-04	3.529086e-09	1.0000
595 В 00e-01	7.500000e-01	2.573740e-04	2.823397e-09	1.0000
596 В 00e-01	7.500000e-01	2.573740e-04	2.258820e-09	1.0000
597 В 00e-01	7.500000e-01	2.573740e-04	1.807139e-09	1.0000
598 В 00e-01	7.500000e-01	2.573740e-04	1.445777e-09	1.0000
599 В 00e-01	7.500000e-01	2.573740e-04	1.156674e-09	1.0000
600 в 00e-01	7.500000e-01	2.573740e-04	9.253811e-10	1.0000
601 B 00e-01	7.500000e-01	2.573740e-04	7.403387e-10	1.0000
602 В 00e-01	7.500000e-01	2.573740e-04	5.922979e-10	1.0000
603 В 00e-01	7.500000e-01	2.573740e-04	4.738601e-10	1.0000
604 В 00e-01	7.500000e-01	2.573740e-04	3.791052e-10	1.0000
605 В 00e-01	7.500000e-01	2.573740e-04	3.032978e-10	1.0000
606 В 00e-01	7.500000e-01	2.573740e-04	2.426496e-10	1.0000
607 C 00e-01	7.500000e-01	2.573740e-04	2.573740e-04	1.0000

		Homework 3		
608 C 00e-01	7.500000e-01	2.058992e-04	2.058992e-04	1.0000
609 C 00e-01	7.500000e-01	1.647194e-04	1.647194e-04	1.0000
610 C 00e-01	7.500000e-01	1.317755e-04	1.317755e-04	1.0000
611 C 00e-01	7.500000e-01	1.054204e-04	1.054204e-04	1.0000
612 C 00e-01	7.500000e-01	8.433631e-05	8.433631e-05	1.0000
613 C 00e-01	7.500000e-01	6.746905e-05	6.746905e-05	1.0000
614 C 00e-01	7.500000e-01	5.397524e-05	5.397524e-05	1.0000
615 C 00e-01	7.500000e-01	4.318020e-05	4.318019e-05	1.0000
616 C 00e-01	7.500000e-01	3.454416e-05	3.454415e-05	1.0000
617 C 00e-01	7.500000e-01	2.763533e-05	2.763532e-05	1.0000
618 C 00e-01	7.500000e-01	2.210827e-05	2.210826e-05	1.0000
619 C 00e-01	7.500000e-01	1.768662e-05	1.768661e-05	1.0000
620 C 00e-01	7.500000e-01	1.414931e-05	1.414929e-05	1.0000
621 C 00e-01	7.500000e-01	1.131945e-05	1.131943e-05	1.0000
622 C 00e-01	7.500000e-01	9.055574e-06	9.055543e-06	1.0000
623 C 00e-01	7.500000e-01	7.244473e-06	7.244434e-06	1.0000
624 C 00e-01	7.500000e-01	5.795596e-06	5.795547e-06	1.0000
625 C 00e-01	7.500000e-01	4.636498e-06	4.636438e-06	1.0000
626 C 00e-01	7.500000e-01	3.709226e-06	3.709150e-06	1.0000

		Homework 3		
627 C 00e-01	7.500000e-01	2.967414e-06	2.967320e-06	1.0000
628 C 00e-01	7.500000e-01	2.373974e-06	2.373856e-06	1.0000
629 C 00e-01	7.500000e-01	1.899232e-06	1.899085e-06	1.0000
630 C 00e-01	7.500000e-01	1.519452e-06	1.519268e-06	1.0000
631 C 00e-01	7.500000e-01	1.215644e-06	1.215414e-06	1.0000
632 C 00e-01	7.500000e-01	9.726183e-07	9.723315e-07	1.0000
633 C 00e-01	7.500000e-01	7.782237e-07	7.778652e-07	1.0000
634 C 00e-01	7.500000e-01	6.227402e-07	6.222922e-07	1.0000
635 C 00e-01	7.500000e-01	4.983937e-07	4.978338e-07	1.0000
636 C 00e-01	7.500000e-01	3.989667e-07	3.982670e-07	1.0000
637 C 00e-01	7.500000e-01	3.194878e-07	3.186136e-07	1.0000
638 C 00e-01	7.500000e-01	2.559828e-07	2.548909e-07	1.0000
639 C 00e-01	7.500000e-01	2.052760e-07	2.039127e-07	1.0000
640 C 00e-01	7.500000e-01	1.648310e-07	1.631302e-07	1.0000
641 C 00e-01	7.500000e-01	1.326241e-07	1.305041e-07	1.0000
642 C 00e-01	7.500000e-01	1.070415e-07	1.044033e-07	1.0000
643 C 00e-01	7.500000e-01	8.679779e-08	8.352265e-08	1.0000
644 C 00e-01	7.500000e-01	7.086949e-08	6.681812e-08	1.0000
645 C 00e-01	7.500000e-01	5.843977e-08	5.345449e-08	1.0000

		Homework 3		
646 C 00e-01	7.500000e-01	4.885231e-08	4.276360e-08	1.0000
647 C 00e-01	7.500000e-01	4.157172e-08	3.421088e-08	1.0000
648 C 00e-01	7.500000e-01	3.615065e-08	2.736870e-08	1.0000
649 C 00e-01	7.500000e-01	3.220579e-08	2.189496e-08	1.0000
650 C 00e-01	7.500000e-01	2.940464e-08	1.751597e-08	1.0000
651 C 00e-01	7.500000e-01	2.746236e-08	1.401278e-08	1.0000
652 C 00e-01	7.500000e-01	2.614369e-08	1.121022e-08	1.0000
653 C 00e-01	7.500000e-01	2.526364e-08	8.968176e-09	1.0000
654 C 00e-01	7.500000e-01	2.468395e-08	7.174541e-09	1.0000
655 C 00e-01	7.500000e-01	2.430570e-08	5.739633e-09	1.0000
656 C 00e-01	7.500000e-01	2.406049e-08	4.591706e-09	1.0000
657 C 00e-01	7.500000e-01	2.390224e-08	3.673365e-09	1.0000
658 C 00e-01	7.500000e-01	2.380041e-08	2.938692e-09	1.0000
659 C 00e-01	7.500000e-01	2.373501e-08	2.350954e-09	1.0000
660 C 00e-01	7.500000e-01	2.369305e-08	1.880763e-09	1.0000
661 C 00e-01	7.500000e-01	2.366617e-08	1.504610e-09	1.0000
662 C 00e-01	7.500000e-01	2.364894e-08	1.203688e-09	1.0000
663 C 00e-01	7.500000e-01	2.363791e-08	9.629508e-10	1.0000
664 C 00e-01	7.500000e-01	2.363085e-08	7.703607e-10	1.0000

00e-01		2.362633e-08	6.162884e-10	1.0000
	0000e-01			
		2.362343e-08	4.930309e-10	1.0000
667 C 7.500	0000e-01	2.362158e-08	3.944247e-10	1.0000
668 C 7.500	0000e-01	2.362040e-08	3.155397e-10	1.0000
669 C 7.500	0000e-01	2.361964e-08	2.524319e-10	1.0000
670 C 7.500	0000e-01	2.361915e-08	2.019453e-10	1.0000
671 C 7.500	0000e-01	2.361884e-08	1.615563e-10	1.0000
672 C 7.500	0000e-01	2.361864e-08	1.292451e-10	1.0000
673 C 7.500	0000e-01	2.361851e-08	1.033962e-10	1.0000
674 C 7.500)000e-01	2.361843e-08	8.271683e-11	1.0000
675 C 7.500	0000e-01	2.361838e-08	6.617362e-11	1.0000
676 C 7.500	0000e-01	2.361835e-08	5.293854e-11	1.0000
677 C 7.500	0000e-01	2.361833e-08	4.235101e-11	1.0000
678 C 7.500	0000e-01	2.361831e-08	3.388079e-11	1.0000
679 C 7.500	0000e-01	2.361830e-08	2.710476e-11	1.0000
680 C 7.500	0000e-01	2.361830e-08	2.168365e-11	1.0000
681 C 7.500	0000e-01	2.361829e-08	1.734701e-11	1.0000
682 C 7.500	0000e-01	2.361829e-08	1.387757e-11	1.0000
683 C 7.500	0000e-01	2.361829e-08	1.110190e-11	1.0000

		Homework 3		
684 C 00e-01	7.500000e-01	2.361829e-08	8.881562e-12	1.0000
685 C 00e-01	7.500000e-01	2.361829e-08	7.105094e-12	1.0000
686 C 00e-01	7.500000e-01	2.361829e-08	5.684231e-12	1.0000
687 C 00e-01	7.500000e-01	2.361829e-08	4.547029e-12	1.0000
688 C 00e-01	7.500000e-01	2.361829e-08	3.637646e-12	1.0000
689 C 00e-01	7.500000e-01	2.361829e-08	2.910228e-12	1.0000
690 C 00e-01	7.500000e-01	2.361829e-08	2.328138e-12	1.0000
691 B 00e-01	7.500000e-01	2.361829e-08	2.361829e-08	1.0000
692 B 00e-01	7.500000e-01	1.889463e-08	1.889463e-08	1.0000
693 B 00e-01	7.500000e-01	1.511570e-08	1.511570e-08	1.0000
694 B 00e-01	7.500000e-01	1.209256e-08	1.209256e-08	1.0000
695 B 00e-01	7.500000e-01	9.674051e-09	9.674051e-09	1.0000

Out[4]: 696

```
In [5]: def Newton(low rank matrix handler, tolerance = 1e-8, max iters = 10000,
                              c armijo = 1e-4, rho armijo =0.9, verbose = False,
                              back_track = False,min_length = 1e-3):
            lagr = low_rank_matrix_handler
            grad norm0_sq = la.norm(lagr.gradient()[0])**2 + la.norm(lagr.gradie
        nt()[1])**2
            grad norm0 = np.sqrt(grad_norm0_sq)
            grad norm = grad norm0
            iters = 0
            while grad norm > tolerance*grad norm0 and iters <= max iters:</pre>
                 cost = lagr.objective()
                 g = lagr.assemble flat gradient()
                 H = lagr.assemble flat Hessian()
                 P = lagr.Newton_direction_tensor()
                 alpha = 1.0
                 if back track:
                     cost_old = cost
                     alpha = 1.
                     while alpha > min_length:
                         B_test = lagr.B +alpha*P[0]
                         C_test = lagr.C +alpha*P[1]
                         cost = lagr.objective(B test,C test)
                         if cost < cost_old +c_armijo*alpha*grad_norm :</pre>
                             break
                         else:
                             alpha *= rho armijo
                 lagr.B += alpha*P[0]
                 lagr.C += alpha*P[1]
                 grad norm = lagr.tensor norm(P)
                 if verbose:
                     if iters == 0:
                         print "\n{0:10} {1:15} {2:15} {3:15}".format(
                               "Iteration", "cost", "||g||L2", "alpha")
                     else:
                         print "\n{0:<10d} {1:<15e} {2:<15e} {3:<15e}".format(</pre>
                               iters, cost, grad norm, alpha)
                 converged = (grad norm < tolerance*grad norm0)</pre>
                 if converged:
                     return converged, iters
                 iters += 1
            return converged, iters
```

```
In [6]: B = np.ones((n,r))
    C = np.ones((r,n))

lagr = low_rank_matrix_handler(A,B,C,G)

Newton(lagr,back_track = False,verbose=True)
```

Iteration	cost	g L2	alpha
1	2.728845e+01	5.305739e+00	1.000000e+00
2	3.351323e+00	1.003696e+01	1.000000e+00
3	1.151980e+03	9.298092e+00	1.000000e+00
4	9.234720e+00	7.417011e+00	1.000000e+00
5	4.058713e+01	2.748416e+00	1.000000e+00
6	1.070959e+01	1.622058e+00	1.000000e+00
7	3.297881e+00	2.959304e+00	1.000000e+00
8	1.710127e+00	7.327686e-01	1.000000e+00
9	1.135522e+00	8.775151e-01	1.000000e+00
10	7.506281e-01	3.544399e-02	1.000000e+00
11	7.500000e-01	3.188247e-06	1.000000e+00
12	7.500000e-01	2.320439e-18	1.000000e+00

Out[6]: (True, 12)

```
In [7]: def trust_region(low_rank_matrix_handler,tolerance = 1e-3,max_iters =100
        00,
                              delta hat = 1.,delta 0 rel = 1.0, eta = 0.25,verbos
        e = False:
            lagr = low_rank_matrix_handler
            grad_norm0 = lagr.tensor_norm(lagr.gradient())
            grad norm = grad norm0
            iters = 0
            delta_0 = delta_0_rel*delta_hat
            delta = delta_0
            while grad norm > tolerance*grad norm0 and iters <= max iters:</pre>
                 cost 0 = lagr.objective()
                 P = lagr.Newton direction tensor()
                 P norm = lagr.tensor norm(P)
                 if P norm > delta:
                     scale = delta/P_norm
                     P = lagr.tensor rescale(P,scale)
                     P norm = delta
                 predicted_reduction = lagr.predicted_reduction(P)
                 B \text{ test} = lagr.B + P[0]
                 C test = lagr.C +P[1]
                 cost_p = lagr.objective(B=B_test,C=C_test)
                 actual_reduction = cost_0 - cost_p
                 rho = actual reduction/predicted reduction
                 if rho < 0.25:
                     delta *= 0.25
                 else:
                     if (rho > 0.75) and (P norm == delta):
                         delta = min(2*delta,delta hat)
                 if rho > eta:
                     lagr.B += delta*P[0]
                     lagr.C += delta*P[1]
                 grad norm = lagr.tensor norm(lagr.gradient())
                 if verbose:
                     if iters == 0:
                         print "\n{0:4} {1:9} {2:9} {3:9} {4:9}".format(
                               "Iteration", "cost", "||g||L2", "rho", "delta" )
                     else:
                         print "\n{0:<9d} {1:<9.1e} {2:<9.1e} {3:<9.1e} {4:<9.1</pre>
        e}".format(
                               iters, cost_0, grad_norm, rho, delta )
                 converged = (grad norm < tolerance*grad norm0)</pre>
                 if converged:
                     return converged, iters
                 if (rho ==np.nan):
                     return False, iters
                 iters += 1
            return iters
```