

# TD #4

## Large-scale Mathematical Programming

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# Universal Isometric Embedding

- ▶ Given metric space  $X$  with  $|X| = n$  and distance matrix (DM)  $D$
- ▶ UIE: finds embedding in  $\ell_\infty^n$
- ▶ Define  $x_{ik} = D_{ik}$  for all  $i, k \leq n$
- ▶ **Thm.:** the DM of  $x$  is  $D$   
*proof seen in lecture*
- ▶ Every graph  $G = (V, E)$  gives rise to a metric space  
*take  $X = V$  and  $d(u, v) = \text{length of shortest path } u \rightarrow v$*

# Universal Isometric Embedding

Exercises (use AMPL and Python):

1. Generate random weighted biconnected graph  $G$  with  $|V| = 50$  output to AMPL .dat
2. Verify its connectedness using Floyd-Warshall's all-shortest-paths algorithm
3. Construct the DM  $\bar{G}$  of the metric space induced by  $G$
4. Find the UIE  $x$  of  $\bar{G}$  in  $\ell_\infty$
5. Verify the DM of  $x$  in  $\ell_\infty$  is  $\bar{G}$
6. Reduce the dimensionality of  $x$  to  $K \in \{2, 3\}$  and draw the realization *see below for PCA details*

# Principal Component Analysis

- ▶ PCA involves finding eigenvalues and eigenvectors
- ▶ AMPL can do it, but it's painful and inefficient
- ▶ Let's use Python instead

## PCA: dist2Gram

```
## convert a distance matrix to a Gram matrix
def dist2Gram(D):
    n = D.shape[0]
    J = np.identity(n) - (1.0/n)*np.ones((n,n))
    G = -0.5 * np.dot(J,np.dot(np.square(D), J))
    return G
```

## PCA: factor

```
## factor a square matrix
def factor(A):
    n = A.shape[0]
    (evals, evecs) = np.linalg.eigh(A)
    evals[evals < 0] = 0 # closest SDP matrix
    X = evecs
    sqrootdiag = np.eye(n)
    for i in range(n):
        sqrootdiag[i,i] = math.sqrt(evals[i])
    X = X.dot(sqrootdiag)
    # because default eig order is small->large
    return np.fliplr(X)
```

# PCA: pca

```
## principal component analysis
def PCA(B,K):
    x = factor(B)
    # only first K columns
    x = x[:,0:K]
    return x
```

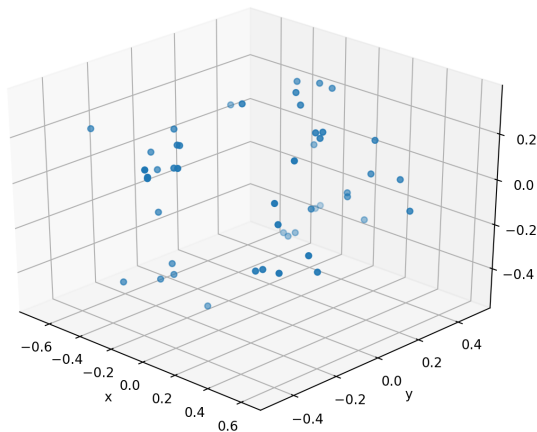
# PCA: main

```
import sys
import numpy as np
import math
import types
from matplotlib import pyplot as plt
from mpl_toolkits.mplot3d import Axes3D
myZero = 1e-9
K = 3 # can be 2 or 3
f = sys.argv[1] # read input filename from command line
lines = [line.rstrip('\n').split()[2:] for line in open(f) if line[0] == 'x']
n = len(lines)
# turn into float array
X = np.array([[float(lines[i][j]) for j in range(n)] for i in range(n)])
G = dist2Gram(X) # if X produced by UIE, X = its own dist matrix
x = PCA(G,K)
if K == 2:
    plt.scatter(x[:,0], x[:,1])
elif K == 3:
    fig = plt.figure()
    ax = Axes3D(fig)
    ax.scatter(x[:,0], x[:,1], x[:,2])
plt.show()
```



# PCA: exercise

Use Python code to display UIE of 50-vtx rnd graph in 3D



# Distance Geometry Problem

Use AMPL to implement 4 DGP MP formulations

1. System of quadratic equations (sqp):

$$\forall \{u, v\} \in E \quad \|x_u - x_v\|_2^2 = d_{uv}^2$$

2. Slack/surplus variables (ssv):

$$\min \left\{ \sum_{\{u,v\} \in E} s_{uv}^2 \mid \forall \{u, v\} \in E \quad \|x_u - x_v\|_2^2 = d_{uv}^2 + s_{uv} \right\}$$

3. Unconstrained quartic polynomial (uqp):

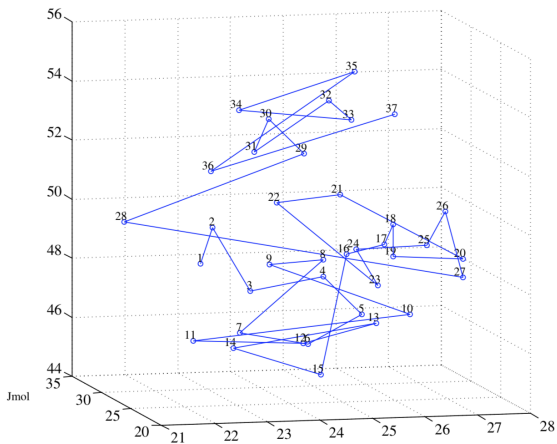
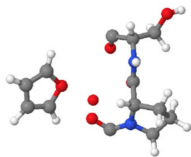
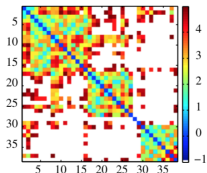
$$\min \sum_{\{u,v\} \in E} (\|x_u - x_v\|_2^2 - d_{uv}^2)^2$$

4. Pull-and-push (p&p):

$$\max \left\{ \sum_{\{u,v\} \in E} \|x_u - x_v\|_2^2 \mid \forall \{u, v\} \in E \quad \|x_u - x_v\|_2^2 \leq d_{uv}^2 \right\}$$

and test them with the protein graph [tiny\\_graph.dat](#)

# DGP: the tiny\_gph instance



# Distance Geometry Problem

- ▶ Use Python to draw the 4 realizations in 3D

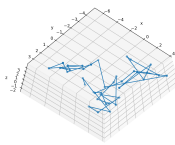
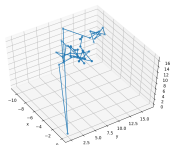
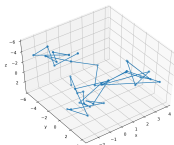
sqp

ssv

uqp

p&p

none  
found



*are they similar?*

- ▶ Compute the UIE of `tiny_gph.dat`

*are there high values in UIE? Why?*

- ▶ Use PCA to display it in 3D

*with high values (left) / replace high values by -1.00 (right)*

