# **Euclidean Distance Completion**

Eucidean Distance Matrices (EDM) are a special class of symmetric matrices that are generated from distance tables. We say  $E \in \mathbb{R}^{n \times n}$  is an EDM if there is a latent variable  $X \in \mathbb{R}^{n \times d}$  such that  $E_{ij} = \|X_i - X_j\|_2^2$  where  $X_i$  is the ith row of the matrix X. We will restrict to the type of EDM where d is fixed, and in particular, d = 2, as these are the EDM that are generated from squared distances in the plane.

### More information

- Liberti, L., Lavor, C., Maculan, N., & Mucherino, A. (2014). Euclidean distance geometry and applications. *Siam Review*, *56*(1), 3-69
- Dokmanic, I., Parhizkar, R., Ranieri, J., & Vetterli, M. (2015). Euclidean distance matrices: essential theory, algorithms, and applications. *IEEE Signal Processing Magazine*, 32(6), 12-30.
- Boyd, S. and Vandenberghe, L (2006). Convex Optimization. Cambridge University Press (see section 8.3.3, and exercise 2.36)

#### Our formulation

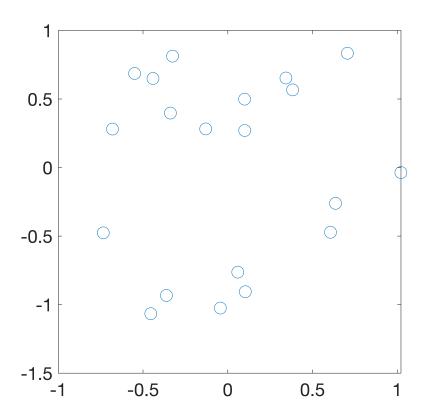
Writing  $K(Y) = 1 \operatorname{diag}(Y)^T + \operatorname{diag}(Y) 1^T - 2Y$ , we have that E is an EDM if  $E = K(XX^T)$  for some X as above. Suppose we only observe  $E_{i,j}$  for some  $(i,j) \in \Omega$ , and we wish to recover E (or more accurately, we wish to recover the underlying X -- note that this is only determined up to rotations).

We can solve  $\min_{XX} \sum_{(i,j) \in \Omega} (K(XX)_{i,j} - E_{i,j})^2$  subject to  $XX \geqslant 0$  and  $\operatorname{rank}(XX) = 2$  (for the d = 2 case).

This is intractable, but if we drop the rank constraint, we have a SDP, which we can solve via CVX.

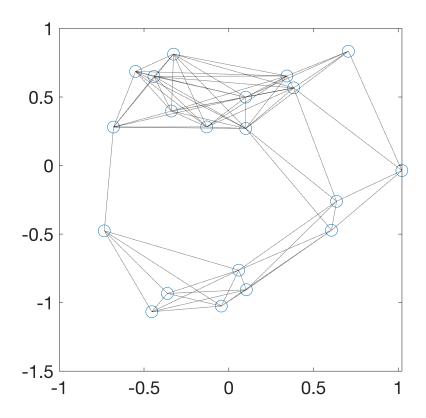
### Setup a problem

```
= 20; % number of points
n
  = 2; % dimension
d
  = ones(n,1);
0
    = @(X) diag(X)*o' + o*diag(X)' - 2*X;
% Make some sample data in [-1,1]x[-1,1]
rng(1);
X0 = 2*(rand(n,d)-.5);
% Put center-of-mass at zero
X0 = X0 - repmat( ones(1,n)/n*X0, n, 1 );
% Plot it:
figure(1); clf;
h=plot( X0(:,1), X0(:,2) , 'o', 'MarkerSize',12);
% h.MarkerFaceColor = h.Color;
set(gca, 'fontsize',18)
hold all
axis square
```



## Observe only some pairwise distances

```
figure(1); clf;
h=plot( X0(:,1), X0(:,2) , 'o', 'MarkerSize',12);
% h.MarkerFaceColor = h.Color;
set(gca,'fontsize',18)
hold all
axis square
% Only observe some of these
   = K(X0*X0');
Е
min_dist
            = 0.86; % this should give perfect recovery
% min_dist
              = 0.85; % this should not
E_thresh
            = ( E < min_dist );
E_thresh
            = E_thresh - tril(E_thresh);
for i = 1:n
    for j = find( E_thresh(i,:) )
        if ~isempty(j)
            hold all
            plot( [X0(i,1),X0(j,1)], [X0(i,2),X0(j,2) ], 'k:' );
        end
    end
end
```



```
omega = find( E_thresh(:) );
obs = @(y) y(omega);
fprintf('Using %d pairwise observations\n',length(omega))
```

Using 74 pairwise observations

# Solve the SDP using CVX

Approximate rank of solution is 2

```
s = svd(XX);
disp(s(1:5))
```

```
8.6551
4.6465
0.0000
0.0000
disp(s(5))
```

1.3488e-09

### Recreate data from XX

```
% We want to find the *right* X such that X*X' = XX
[V,D] = eig(XX);
[eigv,ind] = sort( diag(D), 'descend' );
        = V(:,ind(1:2))*diag( sqrt(eigv(1:2)) );
% Solve Procrustes problem to find best rotation
[u,\sim,v] = svd(X0'*X, 'econ');
R
       = u*v';
Χ
       = X*R';
figure(1); clf;
h=plot( X0(:,1), X0(:,2) , 'o', 'MarkerSize',12);
% h.MarkerFaceColor = h.Color;
set(gca, 'fontsize',18)
hold all
axis square
hold all
hh = plot(X(:,1), X(:,2), '*');
legend([h,hh],'Actual locations','Estimated','location','southeast');
```

