

CS164: Voronoi/Delaunay Diagrams, Distance Functions



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Distance Functions

Object Matching Queries

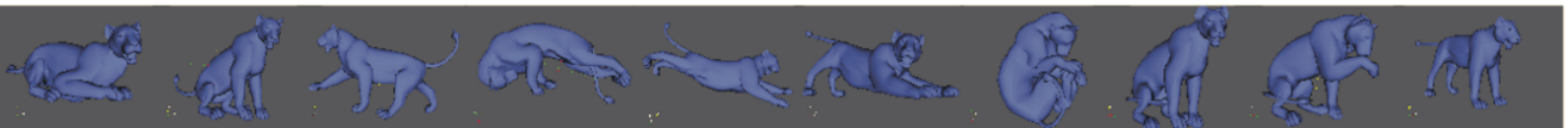
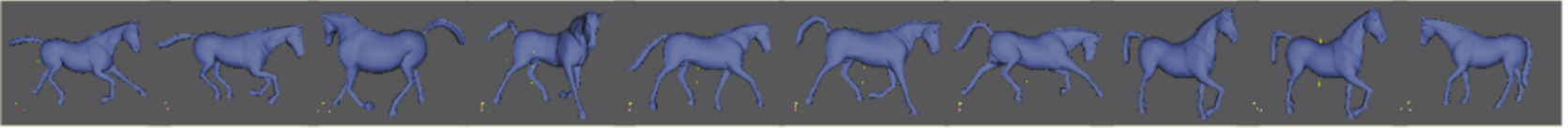
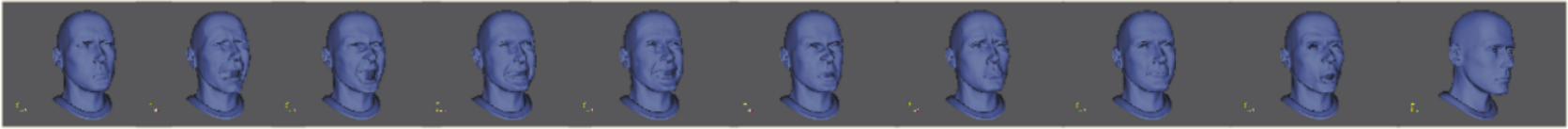
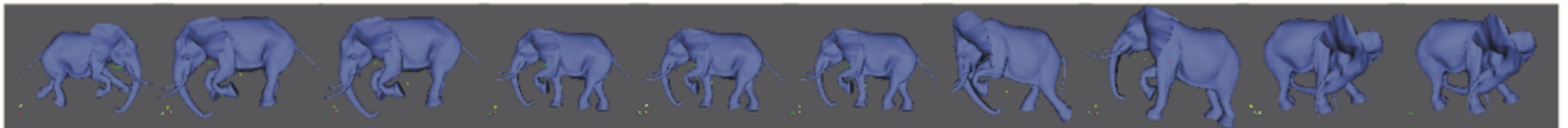
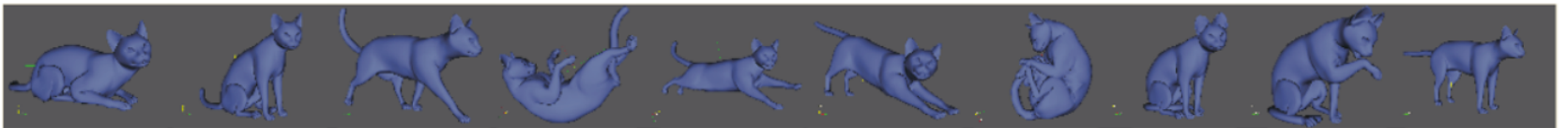
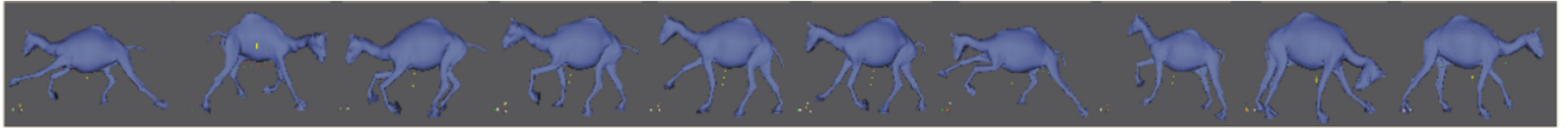
- assume you have database \mathcal{D} of objects.
- assume \mathcal{D} is composed by several objects, and that each of these objects belongs to one of n classes C_1, \dots, C_n .
- imagine you are given a new object o , not in your database, and you are asked to determine whether o belongs to one of the classes. If yes, you also need to point to the class.
- One simple procedure is to say that you will assign object o the class of the *closest* object in \mathcal{D} :

$$\text{class}(o) = \text{class}(z)$$

where $z \in \mathcal{D}$ minimizes **dist**(o, z)

- in order to do this, one first needs to define a notion **dist** of *distance* or *dis-similarity between objects*.

Useful for Object/Shape Classification

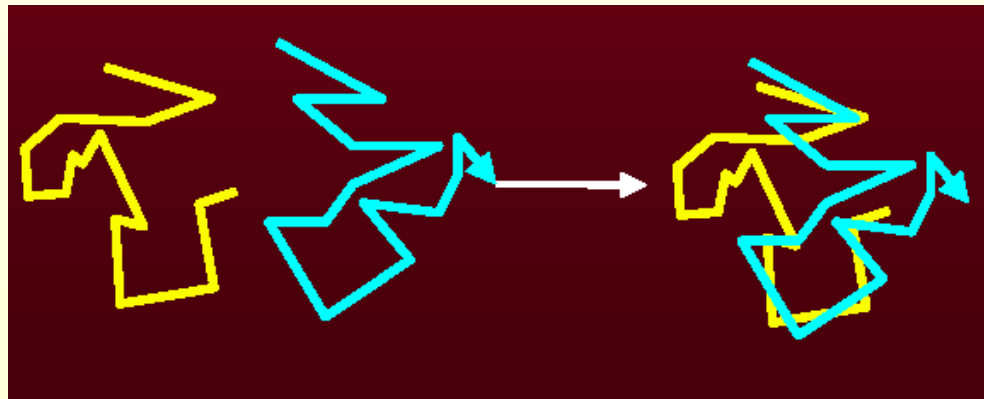


More Shape Similarity Methods

- Hausdorff distance
- Fréchet distance
- Morphing metrics

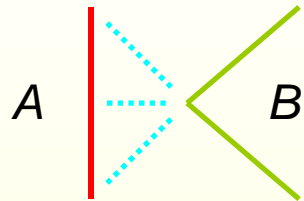
Proteins are defined as having a common fold if they have the same major secondary structures in the same arrangement and with the same topological connections

(SCOP)



Hausdorff Distance

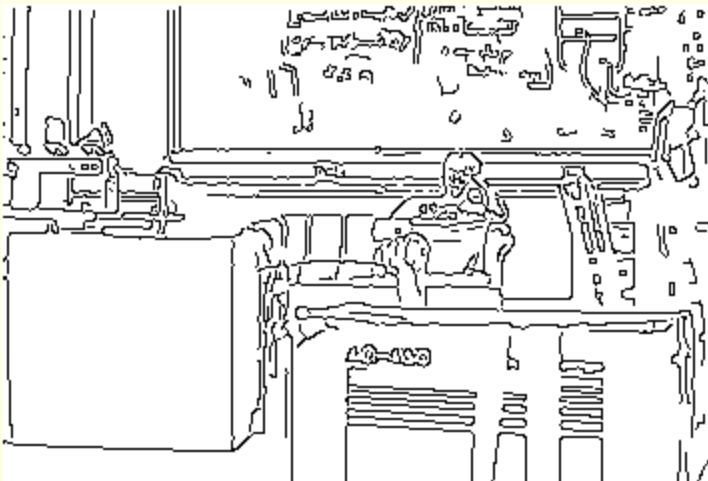
Many-to-many correspondences can be simpler ...



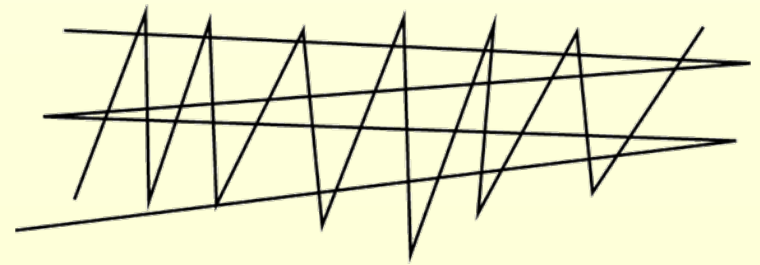
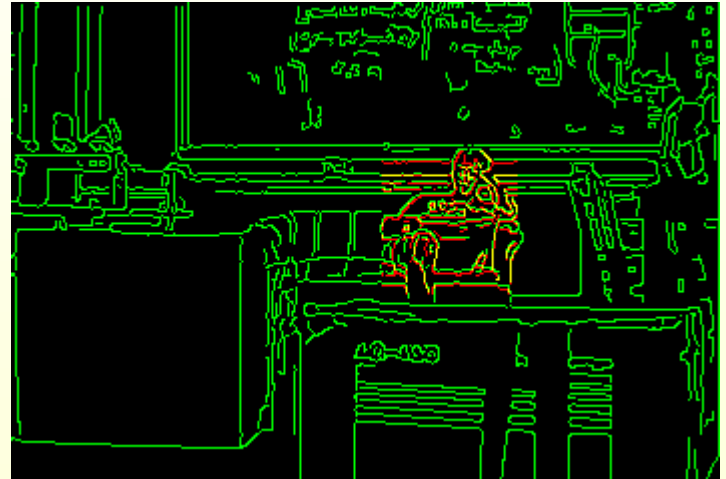
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[Huttenlocher *et. al.*, 93]



Hausdorff Definition

We are two point sets $A = \{a_1, a_2, \dots, a_n\}$ and $B = \{b_1, b_2, \dots, b_m\}$ in E^2 . The one-sided Hausdorff distance from A to B is defined as:

$$\tilde{\delta}_H(A, B) = \max_{a \in A} \min_{b \in B} \|a - b\|$$

The (bidirectional) Hausdorff distance between A and B is then defined as:

$$\delta_H(A, B) = \max \left(\tilde{\delta}_H(A, B), \tilde{\delta}_H(B, A) \right)$$

For fixed A and B , it can easily be computed in time $O((n+m) \log(n+m))$

Hausdorff Variations

- Order statistics – use percentile max (say the 90% largest distance from A to B) to avoid undue impact of outliers (**fractional** Hausdorff)
- Typically, one of the sets (say B) may be moved by a transformation group G

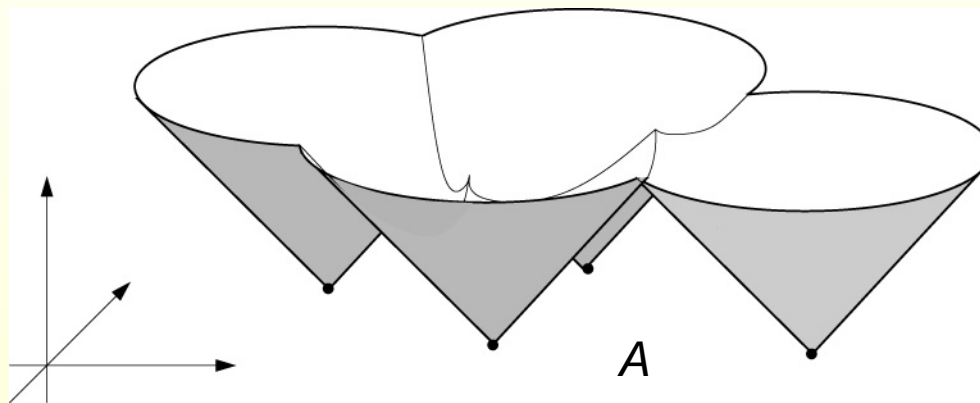
$$\tilde{\delta}_{H,\mathcal{G}}(A, B) = \min_{T \in \mathcal{G}} \max_{a \in A} \min_{b \in B} \|a - T(b)\|$$

- Both “vector” and “raster” methods can be used

Computing Hausdorff

In the plane, vector form, under translation ...

The Voronoi surface of A , a piecewise conical surface



A lower envelope surface

$$d(x) = \min_{a \in A} \|x - a\|$$

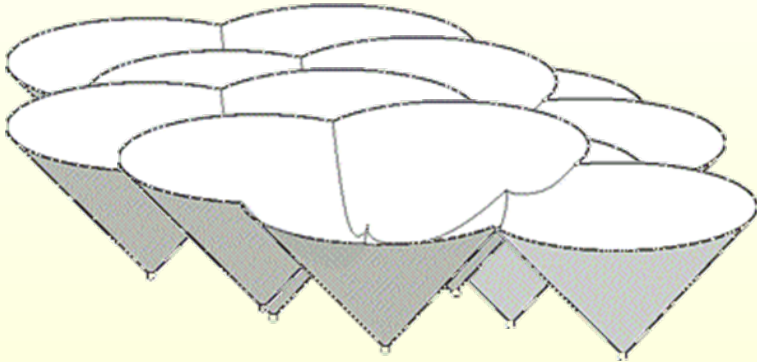
Translate B by t

$$\delta_b(t) = \min_{a \in A} \|a - (b+t)\| = \min_{a \in A} \|(a-b) - t\| = d_{-b}(t)$$

Computing Hausdorff, II

$$f(t) = \tilde{\delta}_H(B + t, A) = \max_{b \in B} \delta_b(t)$$

Upper envelope of m Voronoi surfaces
 $A-b_1, A-b_2, \dots, A-b_m$

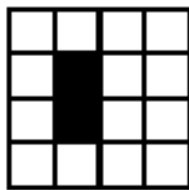


Can be done in time
 $O(nm(n+m) \text{ polylog}(n+m))$

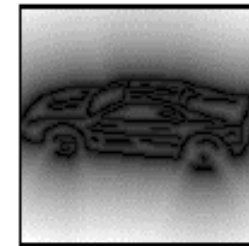
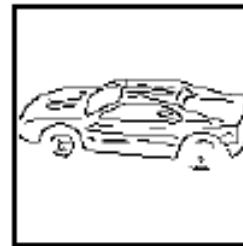
The amount of computation gets out of hand when we allow rotations and go to 3-D.

Raster Hausdorff

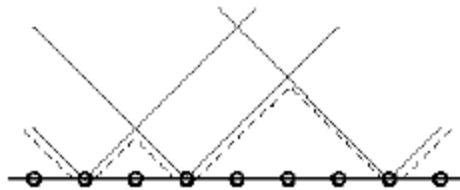
Distance transforms computed on a grid



2	1	2	3
1	0	1	2
1	0	1	2
2	1	2	3



(fast marching, level sets, ...)



∞	0	∞	0	∞	∞	∞	0	∞
∞	0	1	0	1	2	3	0	1
1	0	1	0	1	2	1	0	1

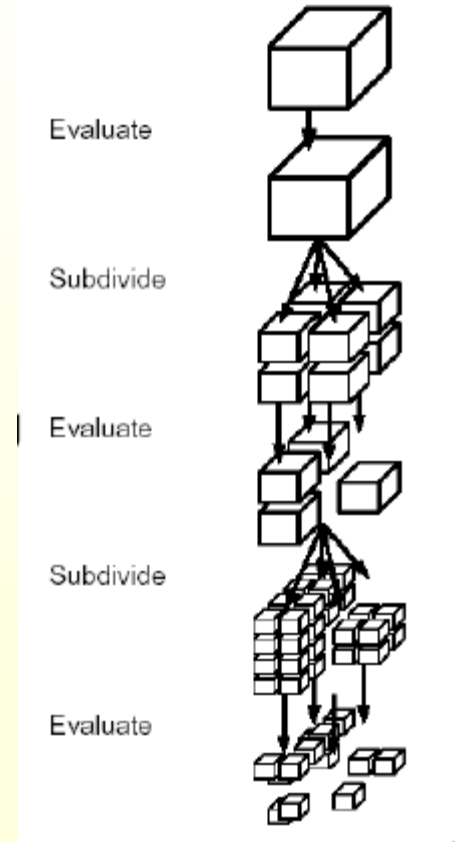
A 1-d example

Fast Hausdorff Search

- Branch and bound hierarchical search of transformation space
- Consider 2-D transformation space of translation in x and y
 - (Fractional) Hausdorff distance cannot change faster than linearly with translation
 - Similar constraints for other transformations
 - “Quad-tree” decomposition, compute distance for transform at center of each cell
 - If larger than cell half-width, rule out cell
 - Otherwise subdivide cell and consider children

Fast Hausdorff Search, II

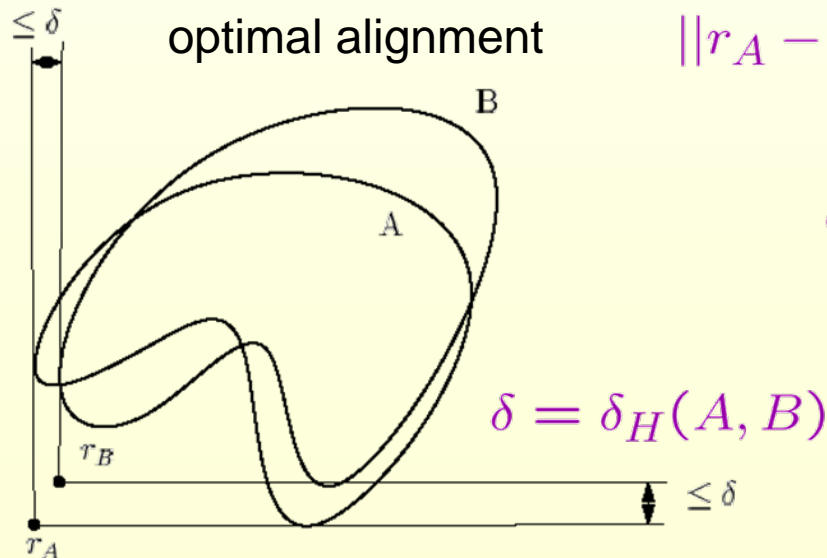
- Guaranteed (or admissible) search heuristic
 - Bound on how good answer could be in unexplored region
 - Cannot miss an answer
 - In worst case won't rule anything out.
 - In practice rule out vast majority of transformations
- In practice rule out vast majority of transformations
 - Can use even simpler tests than computing distance at each cell center



Reference Points

We match shapes by aligning certain well-chosen **reference points**. Such schemes can give constant-factor approximations to the Hausdorff distance [Alt *et. al.*, 91].

Example: approximate Hausdorff in 2-D under translations by matching lower left corner of bounding box



$$\|r_A - r_B\| \leq \sqrt{2}\delta$$

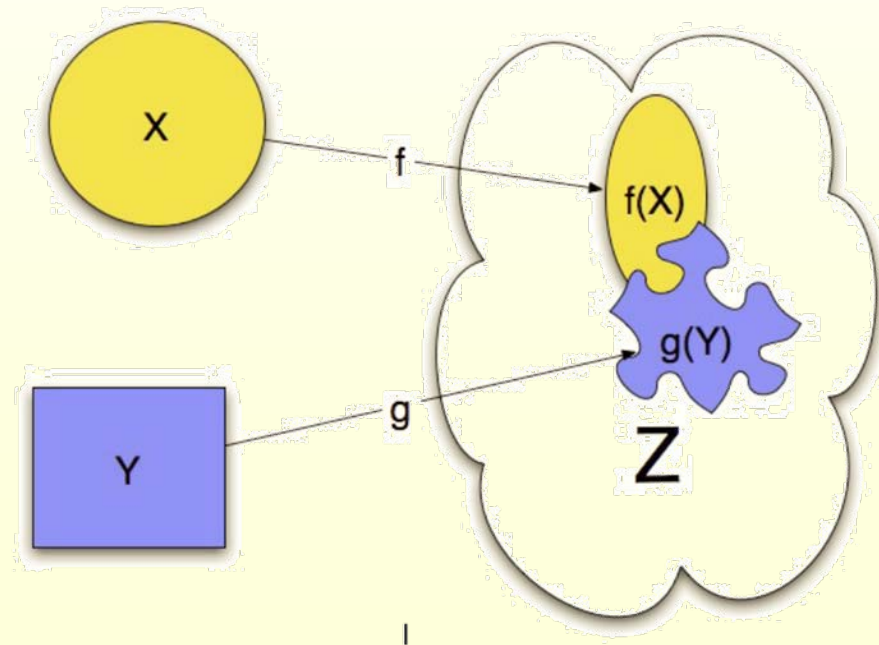
$$\begin{aligned}\delta_H(A, B') &\leq \delta_H(A, B) + \delta_H(B, B') \\ &\leq (\sqrt{2} + 1)\delta\end{aligned}$$

Can be improved by local resampling

Possible for rigid motions, etc.

Gromov-Hausdorff Distance

$$d_{\mathcal{GH}}(X, Y) = \inf_{Z, f, g} d_{\mathcal{H}}^Z(f(X), g(Y))$$



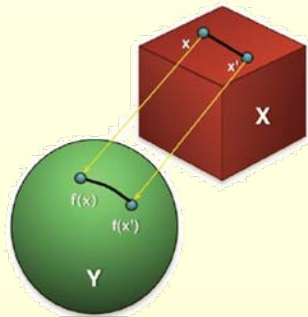
Gromov-Hausdorff Alternate Form

For compact spaces (X, d_X) and (Y, d_Y) let

$$d_{\mathcal{GH}}^{(2)}(X, Y) = \frac{1}{2} \inf_R \max_{(x,y), (x',y') \in R} |d_X(x, x') - d_Y(y, y')|$$

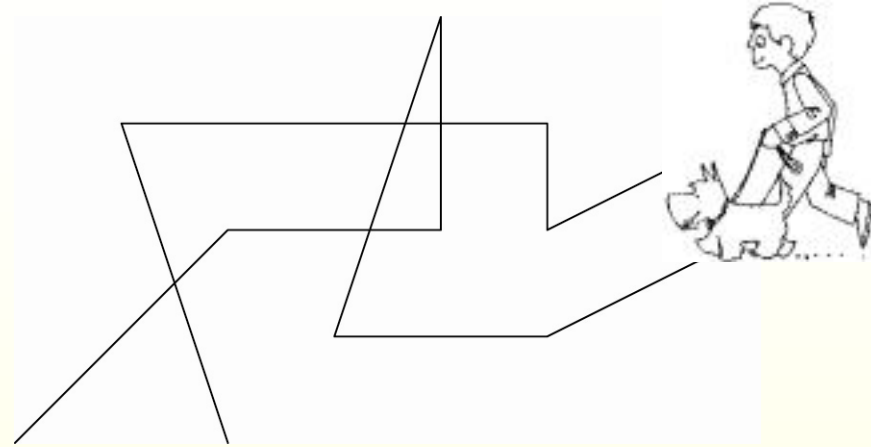
We write, compactly,

$$d_{\mathcal{GH}}^{(2)}(X, Y) = \frac{1}{2} \inf_R \|d_X - d_Y\|_{L^\infty(R \times R)}$$



Hard to compute ...

Fréchet Distance

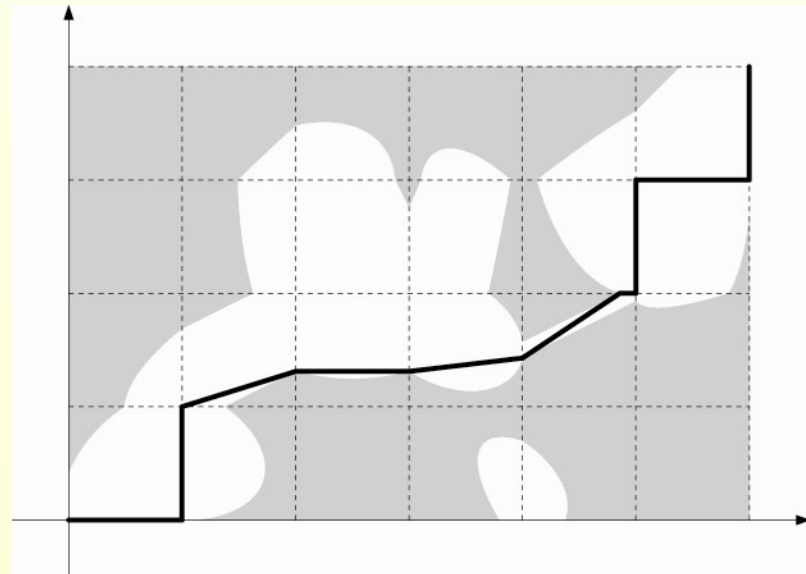
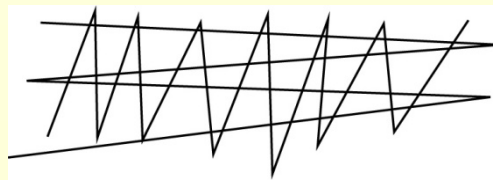


$$\delta_F(f, g) = \inf_{\alpha, \beta} \max_{t \in [0, 1]} \|f(\alpha(t)) - g(\beta(t))\|$$

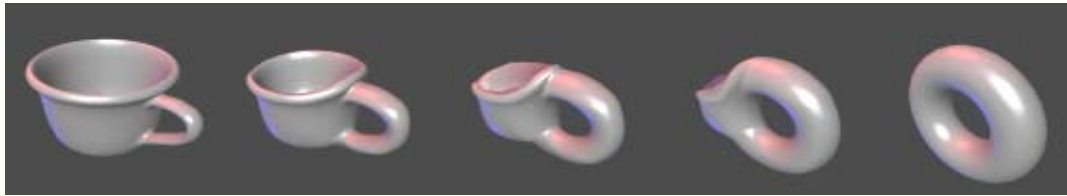
From a decision problem to
an optimization problem.

Guess and verify ...

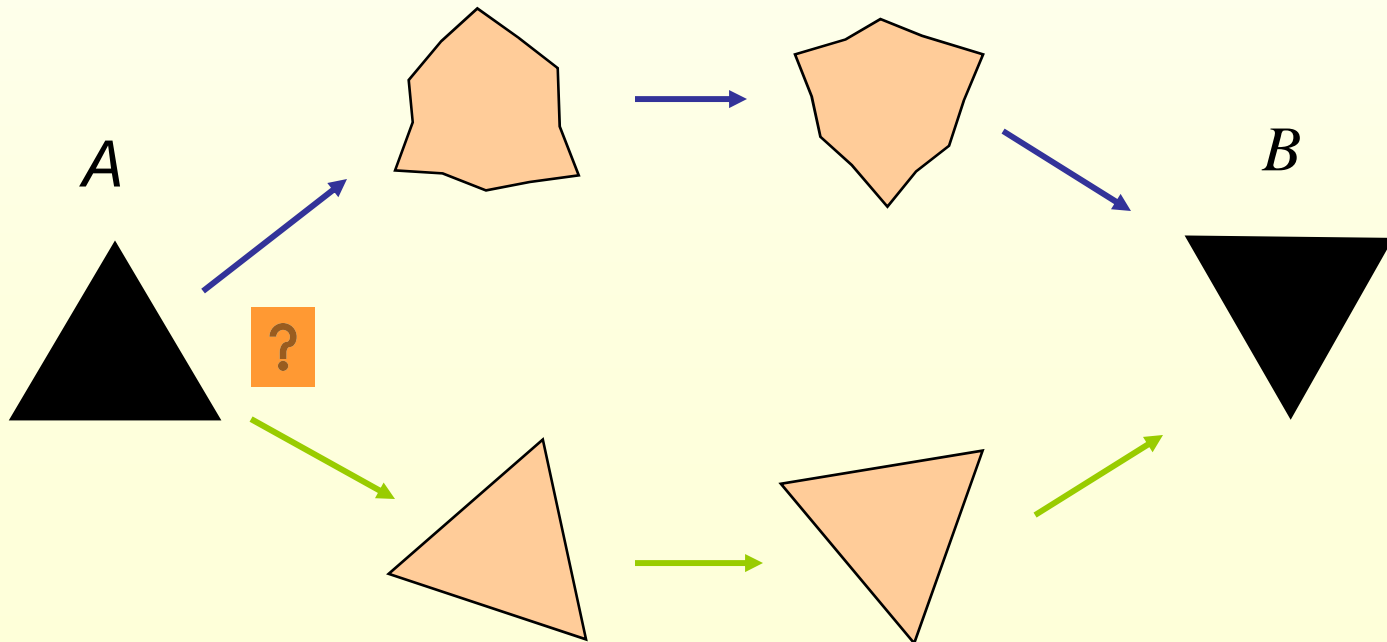
$O(mn \log(mn))$



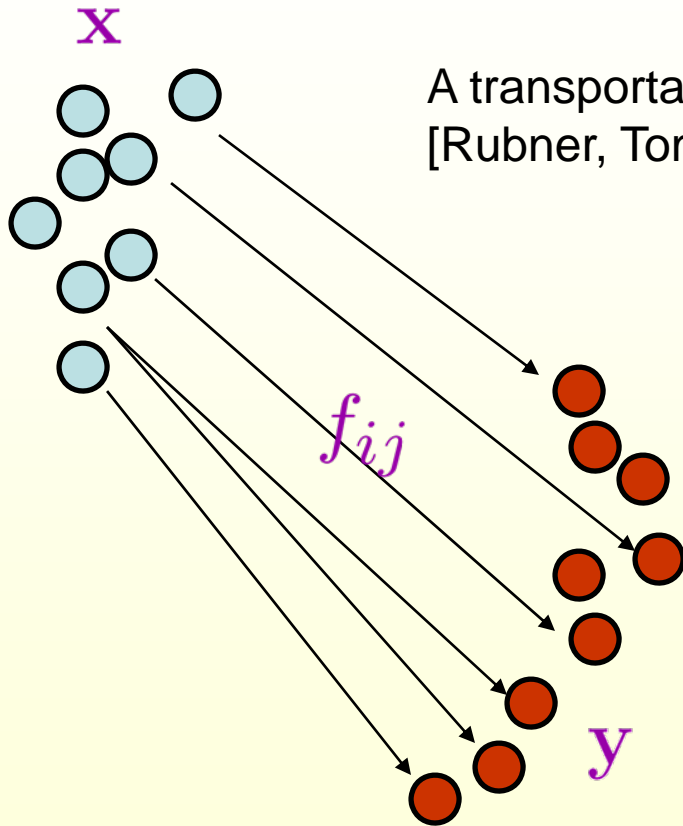
Morphing Distance



minimum cost transformation
from A to B



The Earth Mover's Distance



A transportation metric via linear programming
[Rubner, Tomasi, G., 98]

$$\min \sum_{i \in \mathcal{I}} \sum_{j \in \mathcal{J}} c_{ij} f_{ij}$$

$$f_{ij} \geq 0 \quad i \in \mathcal{I}, j \in \mathcal{J}$$

$$\sum_{i \in \mathcal{I}} f_{ij} = y_j \quad j \in \mathcal{J}$$

$$\sum_{j \in \mathcal{J}} f_{ij} \leq x_i \quad i \in \mathcal{I},$$

$$\text{EMD}(\mathbf{x}, \mathbf{y}) = \frac{\sum_{i \in \mathcal{I}} \sum_{j \in \mathcal{J}} c_{ij} f_{ij}}{\sum_{j \in \mathcal{J}} y_j}$$

Also Wasserstein distance,
Kantorovich-Rubinstein distance ...