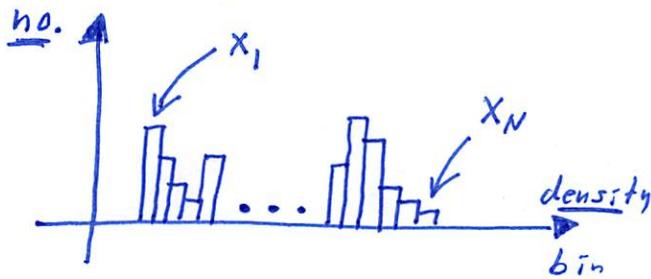


■ On Solving Under-determined linear systems
& Material/Object Classification and Detection

① A material is understood as its associated "characteristic histogram" - e.g., a discrete density histogram ('binned' and using N bins):



→ Histogram interpreted as an N -dim positional vector - that is normalized with only non-negative components/coordinates:

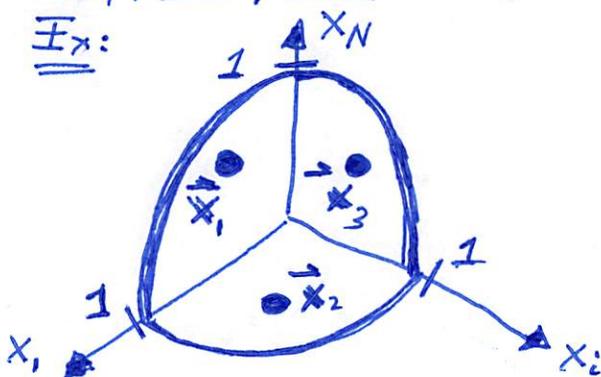
$$\vec{x} = \begin{pmatrix} x_1 \\ \vdots \\ x_N \end{pmatrix}$$

$$\Rightarrow \|\vec{x}\| = \sqrt{\vec{x} \cdot \vec{x}} = \sqrt{\sum_{i=1}^N x_i^2} = l$$

$$\Rightarrow \text{normalized vector: } \vec{x}^{\text{norm}} = \frac{1}{l} \cdot \vec{x}$$

② Each "pure", proto-typical material has an associated (positional) normalized vector defining a point on that part of a unit hyper-sphere that has only non-negative coordinates:

Ex:



→ $\vec{x}_1, \vec{x}_2, \vec{x}_3$ are 3 points on the N -dim. hypersphere representing 3 distinct, pure proto-type materials.

Solving over- and under-determined sets of equations

Suppose

$$\mathbf{y} = M\mathbf{x}$$

where M is a $n \times m$ matrix, \mathbf{y} is a known n -vector and \mathbf{x} is an unknown m -vector.

①

First, assume $n > m$. In this case there are more constraints than unknowns, and the system is overdetermined, with no solutions (except for degenerate cases). We can find a least-squares solution that minimizes the error $(\mathbf{y} - M\mathbf{x})$. We want to find \mathbf{x} that minimizes

OVER-
DETERMINED

$$\|\mathbf{y} - M\mathbf{x}\|^2$$

or

$$(\mathbf{y} - M\mathbf{x})^T (\mathbf{y} - M\mathbf{x})$$

or

$$\mathbf{y}^T \mathbf{y} - \mathbf{y}^T M\mathbf{x} - \mathbf{x}^T M^T \mathbf{y} + \mathbf{x}^T M^T M\mathbf{x}$$

Differentiating w.r.t. \mathbf{x} and setting the result equal to zero yields

$$-(\mathbf{y}^T M)^T - (M^T \mathbf{y}) + 2M^T M\mathbf{x} = 0$$

so

$$\mathbf{x} = (M^T M)^{-1} M^T \mathbf{y} \Leftrightarrow M^T M\mathbf{x} = M^T \mathbf{y}$$

where $(M^T M)^{-1} M^T$ (a $m \times n$ matrix) is called a pseudo-inverse.

②

Second, assume $n < m$. In this case there are fewer constraints than unknowns, and the system is underdetermined, with an infinite number of solutions. We can pick one of these solutions by finding the smallest one. That is, we will minimize \mathbf{x} subject to the constraint $\mathbf{y} = M\mathbf{x}$. The method of Lagrange multipliers has us add a term to the quantity to be minimized:

UNDER-
DETERMINED !

$$\|\mathbf{x}\|^2 + \lambda^T (\mathbf{y} - M\mathbf{x})$$

Differentiating w.r.t \mathbf{x} and setting the result equal to zero yields

$$2\mathbf{x} - M^T \lambda = 0$$

We can't just solve for λ since M is not a square matrix, but we can premultiply by M to obtain

$$2M\mathbf{x} - MM^T \lambda = 0$$

and using $\mathbf{y} = M\mathbf{x}$ gives us

$$2\mathbf{y} = MM^T \lambda$$

so

$$\lambda = 2(MM^T)^{-1} \mathbf{y}$$

and hence

$$\mathbf{x} = M^T (MM^T)^{-1} \mathbf{y}$$

where $M^T (MM^T)^{-1}$ (a $m \times n$ matrix) is called a pseudo-inverse.

* This is just ONE possibility to define a unique solution! QUESTION!

"What is the 'best way' to define a unique solution for our purpose = classification ???

* More general:

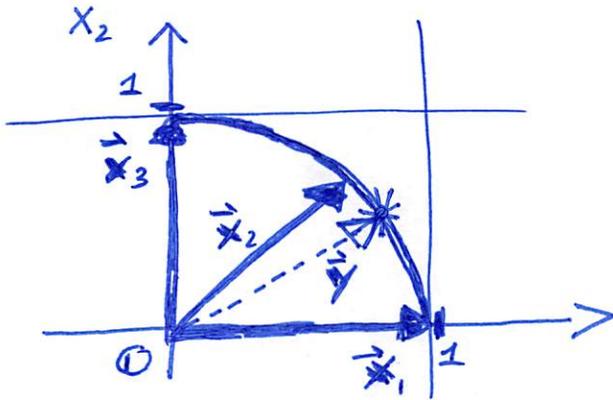
FIND \mathbf{x} such that $\|\mathbf{x} - \mathbf{p}\|_2$ is minimal.

"Solution \mathbf{x} closest to a specified \mathbf{p} "

$$\mathbf{x} = \mathbf{p} + M^T (MM^T)^{-1} (\mathbf{y} - M\mathbf{p})$$

REF: Boehm & Prantztch, Numerical Methods, p. 41.

③ Example: Under-determined problem in the plane -
Using 3 2D vectors



→ 3 normalized vectors
with only positive x_1 and
 x_2 coordinates:

$$\vec{x}_1 = \begin{pmatrix} 1 \\ 0 \end{pmatrix}, \vec{x}_2 = \begin{pmatrix} \sqrt{2}/2 \\ \sqrt{2}/2 \end{pmatrix}, \vec{x}_3 = \begin{pmatrix} 0 \\ 1 \end{pmatrix}$$

→ $\vec{y} = \begin{pmatrix} x_1 \\ x_2 \end{pmatrix}$ to be represented
by "the best" linear
combination of $\vec{x}_1, \vec{x}_2, \vec{x}_3$:

→ Determining the solution that minimizes

$$\vec{x} = \begin{pmatrix} \alpha \\ \beta \\ \gamma \end{pmatrix} \text{ in } M\vec{x} = \vec{y} \text{ leads to the system}$$

$$\alpha \vec{x}_1 + \beta \vec{x}_2 + \gamma \vec{x}_3 = \vec{y}$$

$$\alpha \begin{pmatrix} 1 \\ 0 \end{pmatrix} + \beta \begin{pmatrix} \sqrt{2}/2 \\ \sqrt{2}/2 \end{pmatrix} + \gamma \begin{pmatrix} 0 \\ 1 \end{pmatrix} = \begin{pmatrix} x_1 \\ x_2 \end{pmatrix}$$

$$\begin{pmatrix} 1 & \sqrt{2}/2 & 0 \\ 0 & \sqrt{2}/2 & 1 \end{pmatrix} \begin{pmatrix} \alpha \\ \beta \\ \gamma \end{pmatrix} = \begin{pmatrix} x_1 \\ x_2 \end{pmatrix} \Leftrightarrow M\vec{\alpha} = \begin{pmatrix} x_1 \\ x_2 \end{pmatrix}$$

$$\vec{\alpha} = M^T (M M^T)^{-1} \begin{pmatrix} x_1 \\ x_2 \end{pmatrix}$$

$$= \begin{pmatrix} 1 & 0 \\ \sqrt{2}/2 & \sqrt{2}/2 \\ 0 & 1 \end{pmatrix} \begin{pmatrix} 3/2 & 1/2 \\ 1/2 & 3/2 \end{pmatrix}^{-1} \begin{pmatrix} x_1 \\ x_2 \end{pmatrix}$$

$$= \begin{pmatrix} 1 & 0 \\ \sqrt{2}/2 & \sqrt{2}/2 \\ 0 & 1 \end{pmatrix} \begin{pmatrix} 3/4 & -1/4 \\ -1/4 & 3/4 \end{pmatrix} \begin{pmatrix} x_1 \\ x_2 \end{pmatrix}$$

$$= \begin{pmatrix} 3/4 & -1/4 \\ \sqrt{2}/4 & \sqrt{2}/4 \\ -1/4 & 3/4 \end{pmatrix} \begin{pmatrix} x_1 \\ x_2 \end{pmatrix} = \frac{1}{4} \begin{pmatrix} 3 & -1 \\ \sqrt{2} & \sqrt{2} \\ -1 & 3 \end{pmatrix} \begin{pmatrix} x_1 \\ x_2 \end{pmatrix}$$

Stratovan

(Example - cont'd.)

→ Let's determine "the optimal" coefficient vectors for $\vec{y} = \begin{pmatrix} 1 \\ 0 \end{pmatrix}$ and $\vec{y} = \begin{pmatrix} \sqrt{2}/2 \\ \sqrt{2}/2 \end{pmatrix}$:

(i) $\vec{y} = \begin{pmatrix} 1 \\ 0 \end{pmatrix} = \begin{pmatrix} x_1 \\ x_2 \end{pmatrix}$

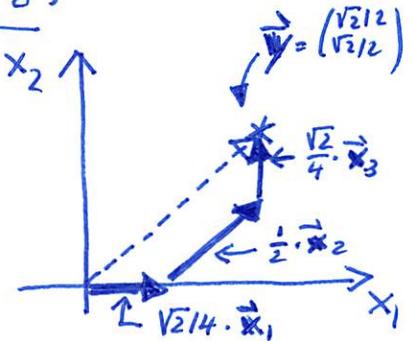
$\Rightarrow \vec{\alpha} = \frac{1}{4} \begin{pmatrix} 3 & -1 \\ \sqrt{2} & \sqrt{2} \\ -1 & 3 \end{pmatrix} \begin{pmatrix} 1 \\ 0 \end{pmatrix} = \frac{1}{4} \begin{pmatrix} 3 \\ \sqrt{2} \\ -1 \end{pmatrix}$

⇒ • IS "COORDINATE" VALUE $\ominus 1$ ALLOWABLE (for our application)?

• WHILE THIS $\vec{\alpha}$ COEFFICIENT VECTOR IS 'OPTIMAL' (in terms of its length), IS IT "BETTER" THAN $\vec{\alpha} = \begin{pmatrix} 1 \\ 0 \end{pmatrix}$???

(ii) $\vec{y} = \begin{pmatrix} \sqrt{2}/2 \\ \sqrt{2}/2 \end{pmatrix} = \begin{pmatrix} x_1 \\ x_2 \end{pmatrix}$

$\Rightarrow \vec{\alpha} = \frac{1}{4} \begin{pmatrix} 3 & -1 \\ \sqrt{2} & \sqrt{2} \\ -1 & 3 \end{pmatrix} \begin{pmatrix} \sqrt{2}/2 \\ \sqrt{2}/2 \end{pmatrix} = \frac{1}{4} \begin{pmatrix} \sqrt{2} \\ 2 \\ \sqrt{2} \end{pmatrix} \rightarrow$

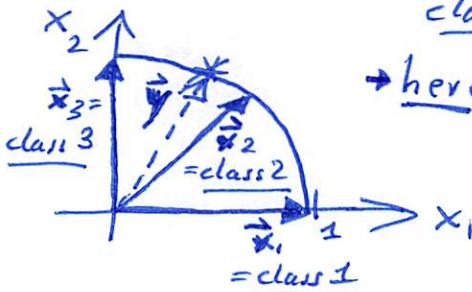


⇒ • IS THIS 'OPTIMAL' COEFFICIENT VECTOR 'BETTER' THAN $\vec{\alpha} = \begin{pmatrix} 1 \\ 0 \end{pmatrix}$???

④ Relationship to Classification - "Correct Classification"

"Our Context:

$\vec{x}_1, \vec{x}_2, \vec{x}_3$ are 3 "material classes"



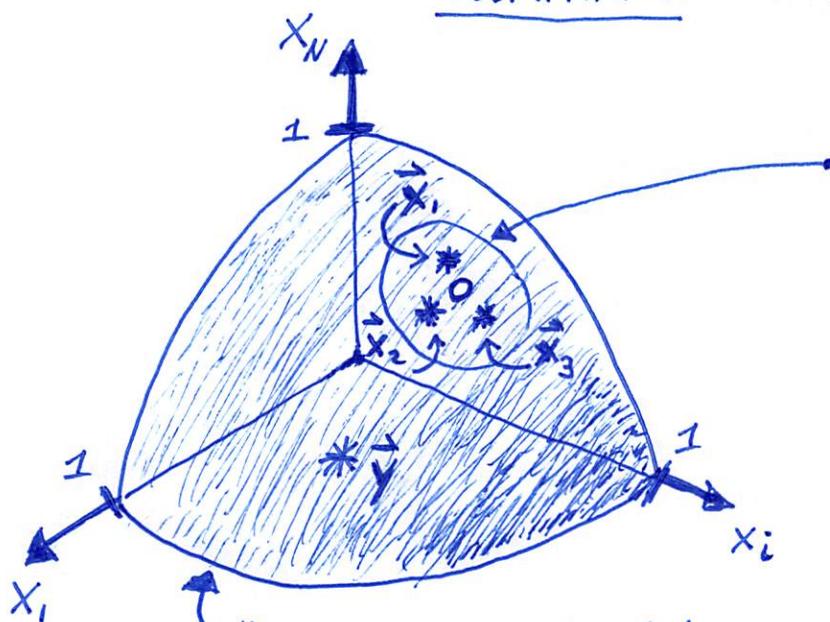
→ here: \vec{y} is close to \vec{x}_2

⇒ Is this the "proper question" to be answered?

• Given "material \vec{y} ", to which class does it belong, i.e., is \vec{y} close to an \vec{x}_i such that it can be viewed as belonging to \vec{x}_i ?

⑤ Possibility to 'avoid' using a linear systems approach -
Using a SCALAR PRODUCT / ANGLE criterion instead

→ Assumption: All histogram data consists of N-dim.
points/vectors with only non-negative
coordinates - and data is normalized!



Cluster $C_j, j=1 \dots L$,
consisting of K_j
(here: $K_j = 3$) associated
points $\vec{x}_1, \vec{x}_2, \vec{x}_3$ -
"defining" one CLASS/
MATERIAL TYPE □

"positive part" of hyper-
sphere $x_1^2 + \dots + x_N^2 = 1$,
 $0 \leq x_i \leq 1, i=1 \dots N$

• PROBLEM:

Given new point \vec{y} ,
does it belong to any
of the clusters/classes?

→ (i) One can associate a "center \vec{c}_j "
with each cluster C_j , by averaging its
associated points $\vec{x}_1, \vec{x}_2, \dots$ (for example):

$$\vec{c}_j = \frac{1}{K_j} \sum_{l=1}^{K_j} \vec{x}_l / \left\| \frac{1}{K_j} \sum_{l=1}^{K_j} \vec{x}_l \right\| \quad ('O' \text{ in figure})$$

$\vec{a} \cdot \vec{b}$
= scalar
product
of \vec{a} and \vec{b}

(ii) DOES \vec{y} REPRESENT A MATERIAL IN C_j ?

$\Leftrightarrow \vec{y} \cdot \vec{c}_j$ CLOSE TO 1.0? YES $\Rightarrow \vec{y}$ OF CLASS C_j

(iii) HOW "CLOSE" IS \vec{y} TO MATERIAL \vec{x}_l ? CONSIDER $\vec{y} \cdot \vec{x}_l$
BH ~