

Data Normalization

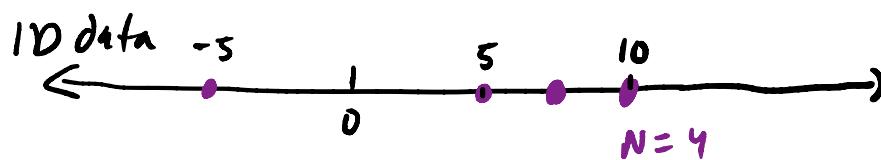
What if we want to apply 2+ transformations to data? e.g. a translation and a Scaling?

Normalization is one operation that requires both T and S .

Normalization for one variable:

Basic idea: Scale data so all values in range $[0, 1]$

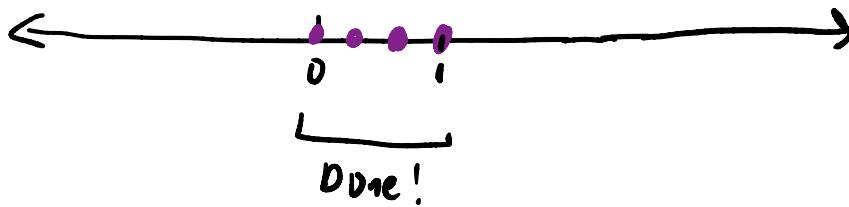
↑
inclusive



1) Subtract min from all samples:



2) Divide by range = max - min.



Formula

$$\frac{\vec{x} - \text{min}}{\text{max} - \text{min}} = \vec{x}_{\text{norm}}$$

Where is the translation?

$$\boxed{x - \min}$$

i.e. translate data by $-\min$.

Where is the Scaling?

$$\boxed{\frac{1}{\max - \min} = \frac{1}{\text{range}}}$$

i.e. divide data by range
[after translating]

Now thinking of >1 variables, how would we set up the matrix multiplication to apply both $T \circ S$ to D ?

$$A' = (T \circ S \circ A \cdot T) \cdot T \quad \text{or} \quad A' = (S \circ T \circ A \cdot T) \cdot T$$

translation gets applied to A first, then we do $S @ (\text{result})$

How do we fill in T & S to do normalization?

■ Option 1 : Global normalization (normalize together)

For T : translate by overall min over all data
[without homogenous coord]

For S : Scale by range over all data
[without homogenous coord]

$$\text{Example: Temperature data } A = \begin{bmatrix} \text{ME} & \text{FL} \\ 32 & 100 \\ 5 & 90 \\ 10 & 80 \end{bmatrix}$$

2D data $\Rightarrow T = \begin{bmatrix} 1 & 0 & t_x \\ 0 & 1 & t_y \\ 0 & 0 & 1 \end{bmatrix}$ need these!

Translate all variables by global min = 5

$$T = \begin{bmatrix} 1 & 0 & -5 \\ 0 & 1 & -5 \\ 0 & 0 & 1 \end{bmatrix} \quad \begin{array}{l} \text{ME + FL temps} \\ \text{both shifted by -5} \end{array}$$

$$T @ A @ T = \begin{bmatrix} 1 & 0 & -5 \\ 0 & 1 & -5 \\ 0 & 0 & 1 \end{bmatrix} \begin{bmatrix} 32 & 5 & 10 \\ 100 & 90 & 80 \\ 1 & 1 & 1 \end{bmatrix} = \begin{bmatrix} 27 & 0 & 5 \\ 95 & 85 & 75 \\ 1 & 1 & 1 \end{bmatrix}$$

Now for the Scaling:

$$S = \begin{bmatrix} S_x & 0 & 0 \\ 0 & S_y & 0 \\ 0 & 0 & 1 \end{bmatrix}$$

need these!

We scale by $\frac{1}{\text{global range}}$:

$\frac{\text{global Max} - \text{global Min}}{100 - 5} = \frac{1}{95}$

$$SeTeA.T = \begin{bmatrix} \frac{1}{95} & 0 & 0 \\ 0 & \frac{1}{95} & 0 \\ 0 & 0 & 1 \end{bmatrix} \begin{bmatrix} 27 & 0 & 5 \\ 95 & 85 & 75 \\ 1 & 1 & 1 \end{bmatrix}$$

$$= \begin{bmatrix} 27/95 & 0 & 5/95 \\ 1 & 85/95 & 75/95 \\ 1 & 1 & 1 \end{bmatrix}$$

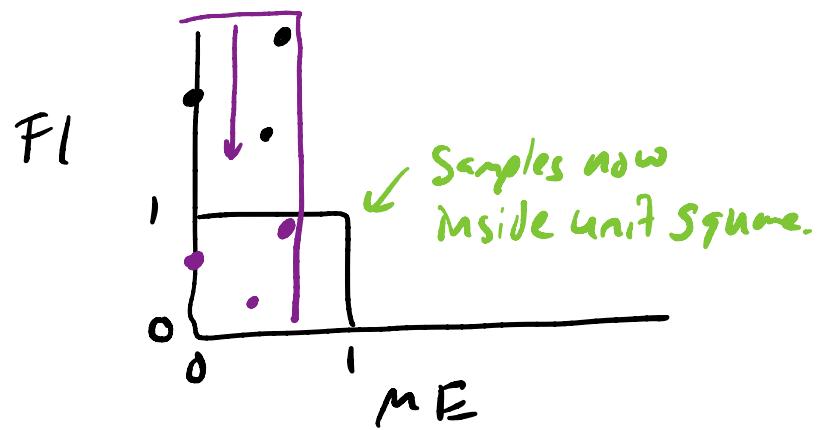
* all data now between 0 + 1.

$$(SeTeA.T).T = \begin{bmatrix} 27/95 & 1 & 1 \\ 0 & 85/95 & 1 \\ 5/95 & 75/95 & 1 \end{bmatrix}$$

Scalinity
Check

Across whole matrix, there should be 1 at overall max + 0 at the overall min — other values should be in between.

Geometry of global normalization:



■ Option 2 : Per-Variable normalization (normalize separately)

For T: translate by each var's min
[without homogenous coord]

For S: Scale by range each variable.
[without homogenous coord]

$$A = \begin{bmatrix} ME & FL & | \\ 32 & 100 & 1 \\ 5 & 90 & 1 \\ 10 & 80 & 1 \end{bmatrix}$$

$$T = \begin{bmatrix} 1 & 0 & t_x \\ 0 & 1 & t_y \\ 0 & 0 & 1 \end{bmatrix}$$

need these!

⇒ Translate by -min of ME and FL separately:

$$\underline{5} \quad \underline{80}$$

$$T @ A \cdot T = \begin{bmatrix} 1 & 0 & -5 \\ 0 & 1 & -80 \\ 0 & 0 & 1 \end{bmatrix} \begin{bmatrix} 32 & 5 & 10 \\ 100 & 90 & 80 \\ 1 & 1 & 1 \end{bmatrix} = \begin{bmatrix} 27 & 0 & 5 \\ 20 & 10 & 0 \\ 1 & 1 & 1 \end{bmatrix}$$

$$S = \begin{bmatrix} S_x & 0 & 0 \\ 0 & S_y & 0 \\ 0 & 0 & 1 \end{bmatrix}$$

need these

\Rightarrow Scale by $\frac{1}{\text{range}}$ for ME & FL computed separately

$$\Rightarrow ME \text{ max-min} = 32 - 5 = 27$$

$$FL \text{ max-min} = 100 - 80 = 20$$

$$S @ T @ A.T = \begin{bmatrix} 1/27 & 0 & 0 \\ 0 & 1/20 & 0 \\ 0 & 0 & 1 \end{bmatrix} \begin{bmatrix} 27 & 0 & 5 \\ 20 & 10 & 0 \\ 1 & 1 & 1 \end{bmatrix}$$

$$= \begin{bmatrix} 1 & 0 & 5/27 \\ 1 & 1/20 & 0 \\ 1 & 1 & 1 \end{bmatrix}$$

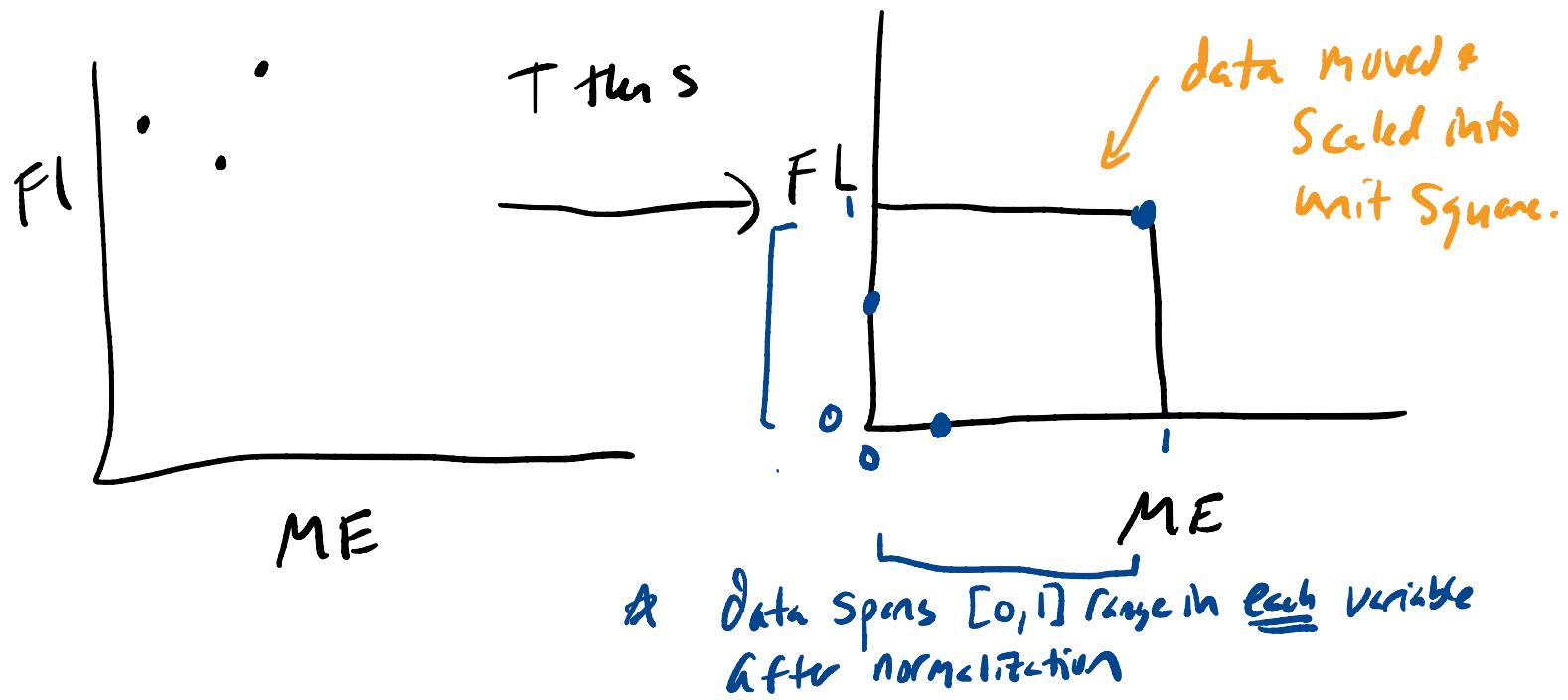
$$(S @ T @ A.T).T = \begin{bmatrix} 1 & 1 & 1 \\ 0 & 1/20 & 1 \\ 5/27 & 0 & 1 \end{bmatrix}$$

* all data now between 0 & 1.

Sanity check

A one and zeros in each column — one set per variable
 [unless there are ties for min/max]

Geometry of normalize separately:



performing 2+ transformations (e.g. T then S) called **Compound transformation**.

Because both T & S have the same shapes,

$(M+1, M+1)$ [e.g. $(3, 3)$ for 2D data]

we can multiply them in advance and get one

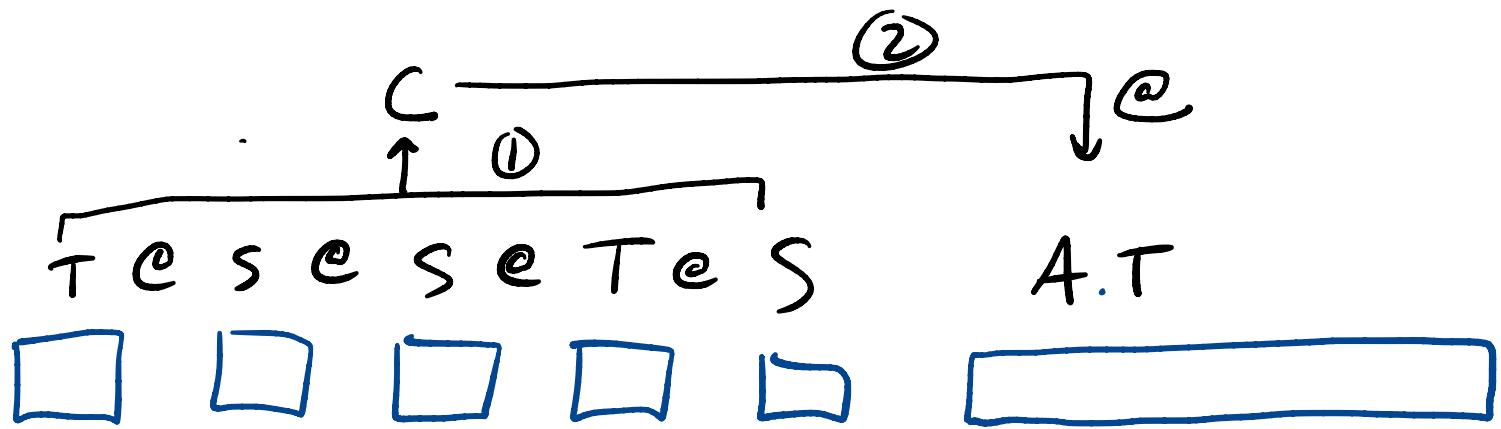
$(M+1, M+1)$ matrix **C**.

e.g. normalization: **$C = S @ T$**

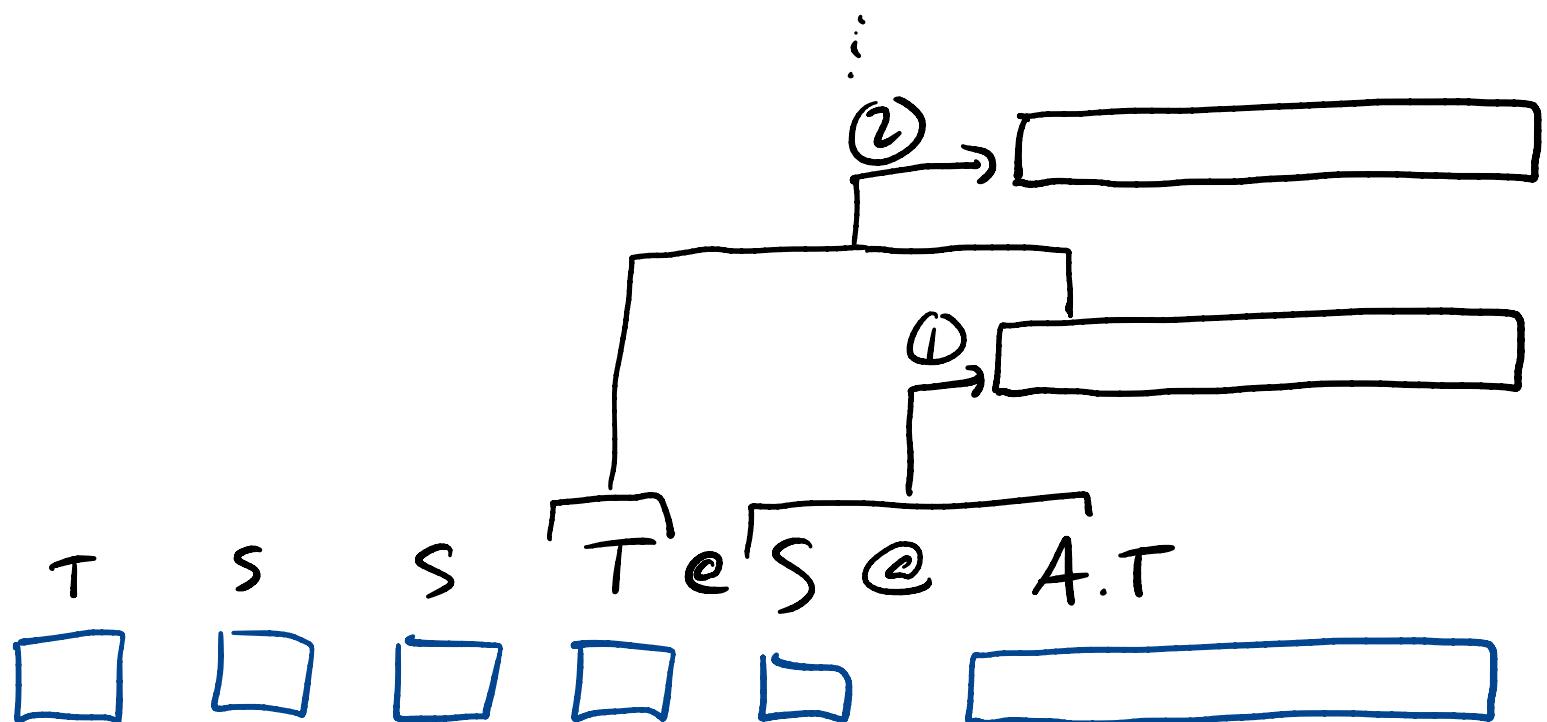
then we always do: **$A' = ((C @ A.T).T$**

Which would be more efficient:

- 1) Multiplying all the transformation matrices first
to get $C \Rightarrow$ then do one transformation on A ?



- 2) Multiply each transformation matrix with $A.T$
One-by-one?



#1 ! apply all transformations with one

matrix multiplication with data matrix

e.g. 1000×4
 $\nwarrow \quad \nearrow$

Why is computing C desirable? If $N > M$ (Many cases)

- Computing lots of 4×4 matrix multiplications to get C is fast. [#1]
- Doing lots of 1000×4 matrix multiplications slow. [#2]

* When Computing C, need to arrange transformation matrices in proper order:

$$C = S @ T \neq T @ S$$