

# Representing Data with Bases

# Objectives

1

- Introduce bases as building blocks for subspaces
- Introduce example uses of bases
- Define orthonormal bases

A subspace can be described as the span  
of a set of vectors. 2

$$\mathcal{S} = \left\{ \underline{x} : \underline{x} = \sum_{i=1}^n \underline{v}_i w_i, w_i \in \mathbb{R}, i=1,2,\dots,n \right\}$$

$\downarrow$   $\underline{x}$  being represented by weighted sum of  $\underline{v}_i$

$$= \text{Span} \{ \underline{v}_i \}$$

- 1)  $\underline{v}_i$  arbitrary - hard computing if linear dep.
  - 2)  $\underline{v}_i$  linearly independent
  - 3)  $\underline{v}_i$  orthonormal - easiest computing
- unique relationship between  $\underline{x}$  and  $w_i$  in 2), 3)

\* **Orthonormal:**  $\underline{v}_i$  are orthogonal to each other and unit length

$$\underline{v}_i^T \underline{v}_i = 1, \underline{v}_i^T \underline{v}_j = 0 \quad \forall i \neq j$$

① What is subspace: "smaller space" sits inside a bigger vector space →  $\star$  It is the space itself that includes infinitely many vectors  
satisfy the 3 requirements in 2.3

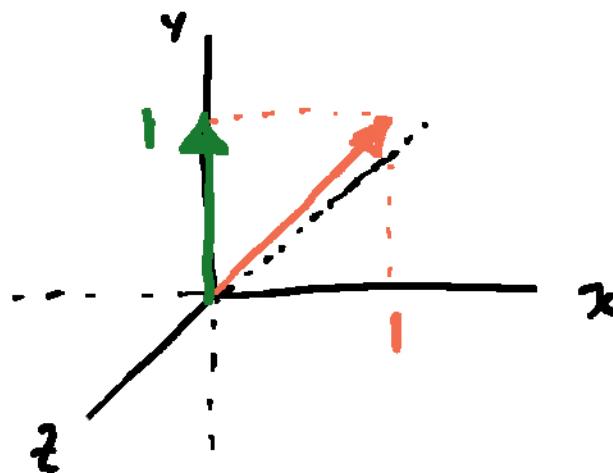
② What is span:  $v_1, \dots, v_m$ , the span is all possible linear combinations of those vectors → Finite set of vectors that give you the recipe for constructing the whole subspace

③ A basis is a minimal set of vectors that spans the subspace  
↓ must be linearly-independent

$\star$  A subspace and its basis always have the same dimension  
↳ The dimension of a subspace is defined as the number of vectors in its basis.

## Example 5

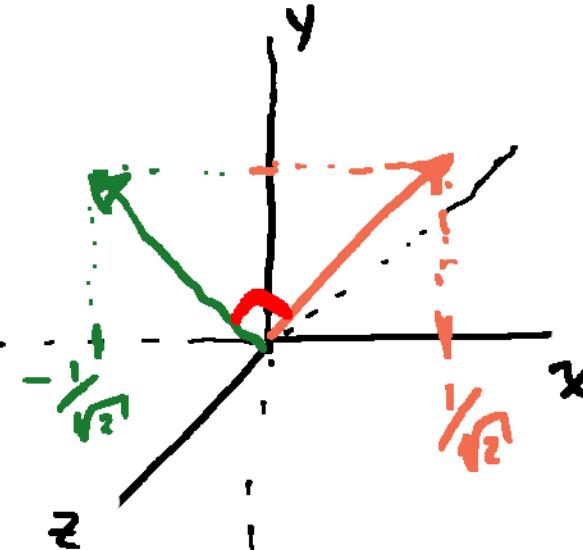
$$\underline{v}_1 = \begin{bmatrix} 1 \\ -1 \\ 0 \end{bmatrix}, \underline{v}_2 = \begin{bmatrix} 0 \\ 1 \\ 0 \end{bmatrix}$$



basis for  $x-y$  plane  
2-D subspace in  $\mathbb{R}^3$

$$f = \underline{v}_1 w_1 + \underline{v}_2 w_2$$

$$\underline{v}_1 = \frac{1}{\sqrt{2}} \begin{bmatrix} 1 \\ 1 \\ 0 \end{bmatrix}, \underline{v}_2 = \frac{1}{\sqrt{2}} \begin{bmatrix} -1 \\ 1 \\ 0 \end{bmatrix}, \underline{v}_3 = \frac{1}{2} \begin{bmatrix} 1 \\ 1 \\ 1 \end{bmatrix}, \underline{v}_4 = \frac{1}{2} \begin{bmatrix} 1 \\ -1 \\ -1 \end{bmatrix}$$



$$\underline{v}_1^\top \underline{v}_1 = \underline{v}_2^\top \underline{v}_2 = 1$$

$$\underline{v}_1^\top \underline{v}_2 = 0$$

orthonormal basis  
for  $x-y$  plane

$$\underline{v}_3 = \frac{1}{2} \begin{bmatrix} 1 \\ -1 \\ -1 \end{bmatrix}$$

(Any vector in the span can be expressed as a weighted sum of  $\underline{v}_1/\underline{v}_2/\underline{v}_3$ )

$$\underline{v}_1^\top \underline{v}_1 = 1$$

$$\underline{v}_1^\top \underline{v}_2 = 0$$

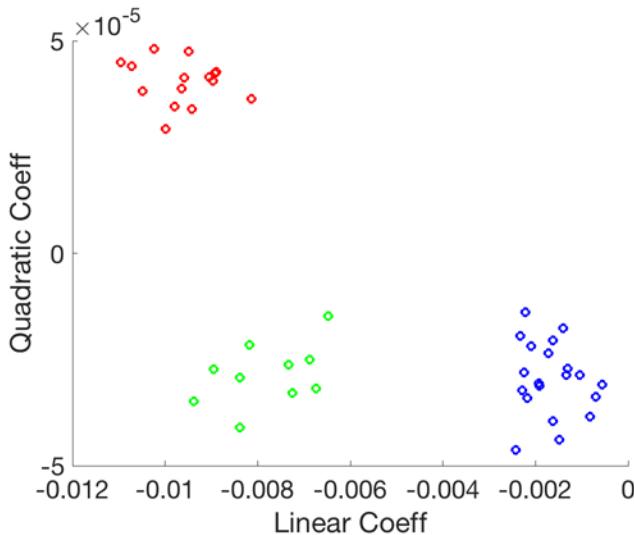
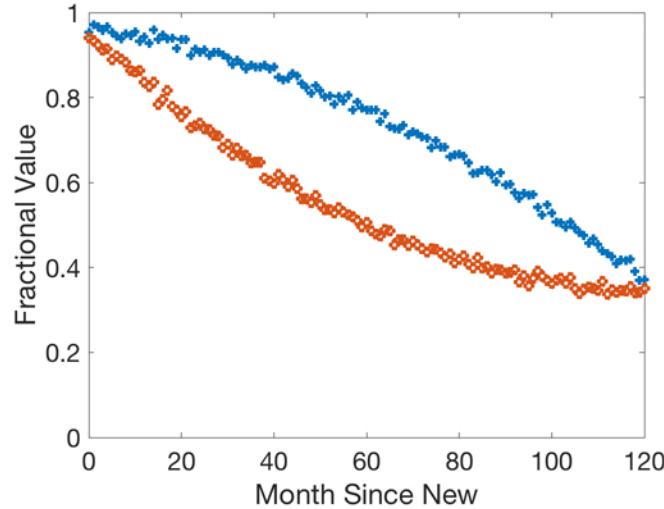
$$\underline{v}_1^\top \underline{v}_3 = 0$$

$$\underline{v}_2^\top \underline{v}_3 = 0$$

orthonormal basis

# Example : Modeling Depreciation

4



$$P_i = w_1 + t_i w_2 + t_i^2 w_3$$

$$\begin{bmatrix} P_0 \\ P_1 \\ \vdots \\ P_{120} \end{bmatrix} = \begin{bmatrix} 1 & t_0 & t_0^2 \\ 1 & t_1 & t_1^2 \\ \vdots & \vdots & \vdots \\ 1 & t_{120} & t_{120}^2 \end{bmatrix} \begin{bmatrix} w_1 \\ w_2 \\ w_3 \end{bmatrix}$$

$P$ : 121 dim  
 $w$ : 3 dim

basis coefficients

$v_1$     $v_2$     $v_3$  : bases

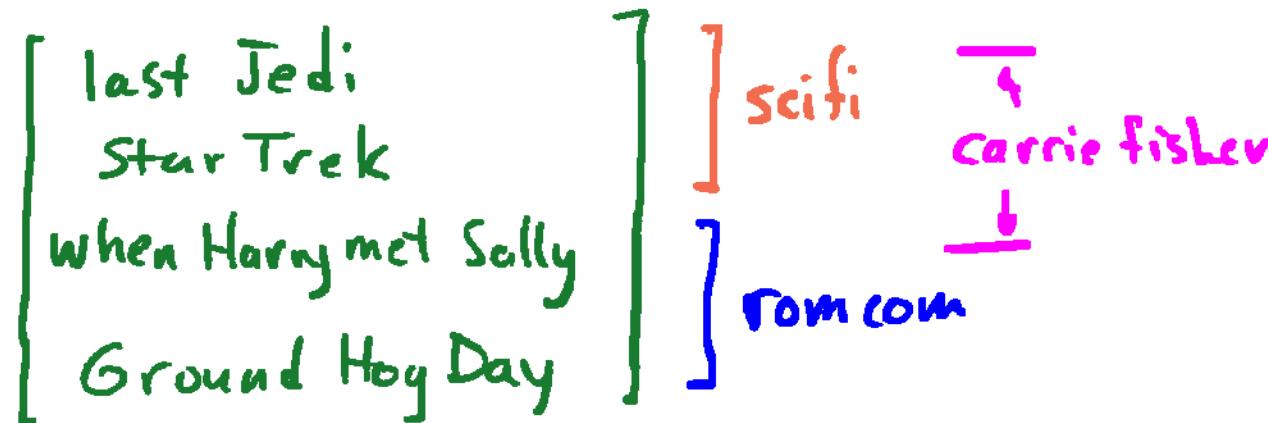
loses value initially

loses value

holds value initially

# Example: Movie Ratings

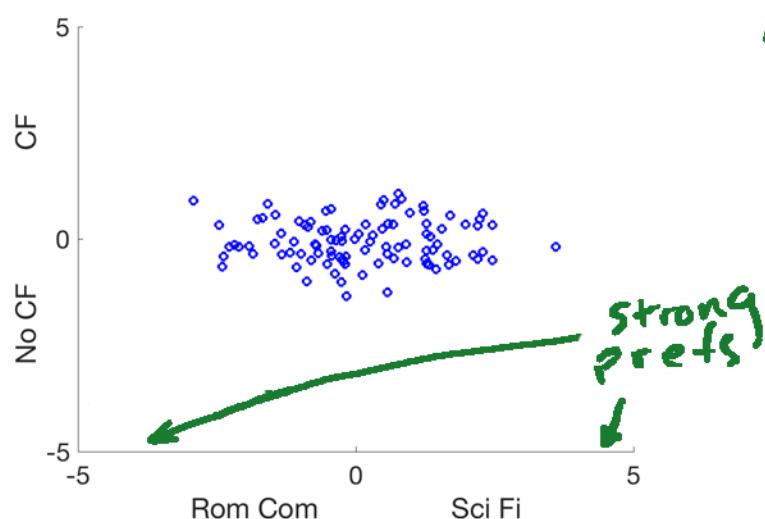
5



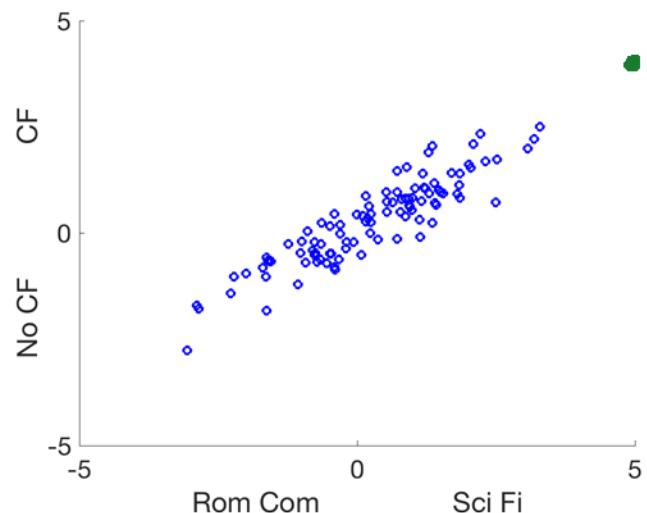
$$\underline{r}_i = \underline{t}_1 w_1 + \underline{t}_2 w_2$$

100 people  $\times$  4 dimensions

cf vs no cf



- sf vs rc stronger
- no link



- sf-rc linked with cf-ncf

## Taste profiles (bases)

$$\underline{t}_1 = \begin{bmatrix} 1 \\ 1 \\ -1 \\ -1 \end{bmatrix}$$

sf vs. rc

$$\underline{t}_2 = \begin{bmatrix} 1 \\ -1 \\ 1 \\ -1 \end{bmatrix}$$

# Open Issues

- Choosing a good/useful basis
- Choosing dimension
- Finding basis coefficients given bases
- Finding orthonormal bases

Copyright 2019  
Barry Van Veen