# Non-negative Matrix Factorization (NMF)

#### **Problem Motivation**

- Soft clustering
  - Clustering where each observation can have partial membership in each cluster
- Identify latent features
  - Identify topics in a corpus of text
  - Identify parts of faces in facial recognition

- Given a cake, we can directly measure carbohydrates, fat, and protein
- We cannot directly observe how much flour and sugar went into the cake
- NMF can learn ingredients from the final product
  - NMF learns the ingredients and the amount of each ingredient in each cake

- Recipe is a linear combination of ingredients
  - o 2 cups flour
  - 1 cup sugar

- Ingredients are linear combination of carbohydrates, fat, and protein
  - Flour:
    - 20g carbohydrates, 5g protein, 1g fat = **[20, 5, 1]**
  - Sugar:
    - 10g carbohydrates, 0g protein, 1g fat = [10, 0, 1]

Recipe and ingredients can be represented by vectors/matrices

Recipe = [flour, sugar] = [2, 1]
Ingredients = 
$$\begin{bmatrix} f_c & f_p & f_f \\ s_c & s_p & s_f \end{bmatrix}$$
= 
$$\begin{bmatrix} 20 & 5 & 1 \\ 10 & 0 & 1 \end{bmatrix}$$

Cake is product of recipe vector and ingredient matrix

Cake = 
$$(recipe)[Ingredients]$$
  
=  $(2 \ 1)\begin{bmatrix} 20 & 5 & 1 \\ 10 & 0 & 1 \end{bmatrix}$   
=  $(2 \ 20 \ 5 \ 1) + 1(10 \ 0 \ 1)$   
=  $(50 \ 10 \ 3)$ 

Many cakes can be represented by product of 2 matrices

$$\begin{bmatrix} c_1 & p_1 & f_1 \\ c_2 & p_2 & f_2 \\ c_3 & p_3 & f_3 \\ \vdots & \vdots & \vdots \end{bmatrix} = \begin{bmatrix} w_{1f} & w_{1s} \\ w_{2f} & w_{2s} \\ w_{3f} & w_{3s} \\ \vdots & \vdots & \vdots \end{bmatrix} \begin{bmatrix} f_c & f_p & f_f \\ s_c & s_p & s_f \end{bmatrix}$$

Finding the recipe

Given cakes and a set of ingredients, can we figure out the recipe?

$$\begin{bmatrix} 30 & 5 & 2 \\ 20 & 5 & 3 \\ 25 & 3 & 3 \\ \vdots & \vdots & \vdots \end{bmatrix} = \begin{bmatrix} w_{1f} & w_{1s} \\ w_{2f} & w_{2s} \\ w_{3f} & w_{3s} \\ \vdots & \vdots & \vdots \end{bmatrix} \begin{bmatrix} 20 & 5 & 1 \\ 10 & 0 & 1 \end{bmatrix}$$

Finding the recipe

Can perform OLS minimization to find the recipe

$$\begin{bmatrix} 25 & 3 & 3 \end{bmatrix} = \begin{pmatrix} w_f & w_s \end{pmatrix} \begin{bmatrix} 20 & 5 & 1 \\ 10 & 0 & 1 \end{bmatrix}$$

$$25 \approx \hat{c} = 20w_{3f} + 10w_{3s}$$

$$3 \approx \hat{p} = 5w_{3f} + 0w_{3s}$$

$$3 \approx \hat{f} = 1w_{3f} + 1w_{3s}$$

$$\underset{\vec{w}}{\operatorname{argmin}} \sum_{y \in (c, p, f)} (y - \hat{y})^2$$

Finding the ingredients

Given cakes and a set of recipes, can we figure out the ingredients?

$$\begin{bmatrix} 30 & 5 & 2 \\ 20 & 5 & 3 \\ 25 & 3 & 3 \\ \vdots & \vdots & \vdots \end{bmatrix} = \begin{bmatrix} 2 & 1 \\ 1 & 1 \\ 1 & 2 \\ \vdots & \vdots \end{bmatrix} \begin{bmatrix} f_c & f_p & f_f \\ s_c & s_p & s_f \end{bmatrix}$$

Finding the ingredients

Can perform OLS minimization to find the ingredients

$$[25 \ 3 \ 3] = (1 \ 2) \begin{bmatrix} f_c & f_p & f_f \\ s_c & s_p & s_f \end{bmatrix}$$

$$25 = \hat{c} \approx 1f_c + 2s_c = 1f_c + 2s_c + 0f_p + 0s_p + 0f_f + 0s_f$$

$$3 = \hat{p} \approx 1f_p + 2s_p = 0f_c + 0s_c + 1f_p + 2s_p + 0f_f + 0s_f$$

$$3 = \hat{f} \approx 1f_f + 2s_f = 0f_c + 0s_c + 0f_p + 0s_p + 1f_f + 2s_f$$

...continued for all observations

Finding the ingredients and the recipes

What if you don't know the recipe proportions *or* the ingredients?

$$\begin{bmatrix} 30 & 5 & 2 \\ 20 & 5 & 3 \\ 25 & 3 & 3 \\ \vdots & \vdots & \vdots \end{bmatrix} = \begin{bmatrix} w_{1f} & w_{1s} \\ w_{2f} & w_{2s} \\ w_{3f} & w_{3s} \\ \vdots & \vdots & \end{bmatrix} \begin{bmatrix} f_c & f_p & f_f \\ s_c & s_p & s_f \end{bmatrix}$$

## Alternating Least Squares

$$X = WH$$

- 1. Initialize\* W and H
- 2. Solve for H holding W constant (minimize OLS)
- 3. Solve W holding H constant (minimize OLS)
- 4. Repeat steps 2 and 3 until convergence

<sup>\*</sup>Initialization methods vary. Random initialization and k-means are common.

$$X = WH$$

X = (observations) x (features)

W = (observations) x (latent features)

H = (latent features) x (observed features)

**Topic Modeling** 

$$W = (documents) x (?)$$

$$H = (?) \times (word counts)$$

→ Latent features correspond to topics. This is explored in today's assignment.

**Face Identification** 

Suppose X is a set of images of faces (all the same size)

X = (images) x (pixel intensities)

W = (image) x (?)

H = (?) x (pixel intensities)

→ Latent features correspond to parts of faces

**Face Identification** 

Basis images from NMF on faces <a href="http://www.cs.ucsb.edu/~mturk/pubs/IJPRAI05.pdf">http://www.cs.ucsb.edu/~mturk/pubs/IJPRAI05.pdf</a>



#### Extensions

- Regularization
  - Can add lasso or ridge term to ALS
- Cross-validation
  - See "Rank Selection in Low-rank Matrix Approximations: A Study of Cross-Validation for NMFs" <a href="https://www.cs.umd.edu/~bhargav/nips2010.pdf">https://www.cs.umd.edu/~bhargav/nips2010.pdf</a>
- Other cost functions
- Other optimization methods
  - Multiplicative update rule