

# CS/ECE/ME 532 Activity 13

- Today - unit 4 continued

- Bias-variance tradeoff

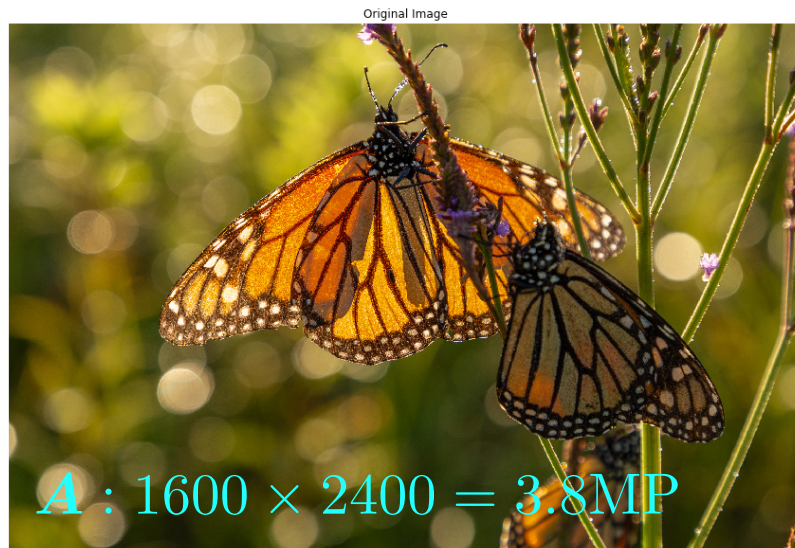
- Applications of SVD

- Image compression
- Face recognition
- Demo's of both ***bonus notebooks***

- Note about handing in code:

- limit what you print out
- indicate where a question is answered:

```
#####  
#   problem 2c  
#####  
    some_code = ...
```



## Eigenfaces for face recognition



### [PDF] Face recognition using eigenfaces

[M Turk, A Pentland - ... . 1991 IEEE computer society conference](#)

We present an approach to the detection and identification of human faces using a set of orthonormal basis functions called eigenfaces. We present a working, near-real-time face recognition system which tracks a subject's face and recognizes the person by comparing characteristics of the face to the eigenfaces.

☆ 99 Cited by 7662 Related articles All 65 versions »

# Today – More SVD – bias, variance trade-off

$$\mathbf{A} = \begin{bmatrix} \mathbf{A} \\ N \times M \end{bmatrix} = \begin{bmatrix} \mathbf{U} \\ N \times N \end{bmatrix} \begin{bmatrix} \mathbf{\Sigma} \\ N \times M \end{bmatrix} \begin{bmatrix} \mathbf{V}^T \\ M \times M \end{bmatrix}$$

## Bias

SVD defines the ‘best’ rank  $r$  approximation (EY, 1936):

$$\min \|\mathbf{A} - \hat{\mathbf{A}}\|_F^2$$

over all matrices  $\hat{\mathbf{A}}$  with rank  $\leq r$

$$\text{given by } \hat{\mathbf{A}} = \sum_{i=1}^r \sigma_i \mathbf{u}_i \mathbf{v}_i^T$$

*sum of outer products*

How far off is  $\hat{\mathbf{A}}$ ?  $\|\mathbf{A} - \hat{\mathbf{A}}\|_F^2 = \sum_{i=r+1}^n \sigma_i^2$

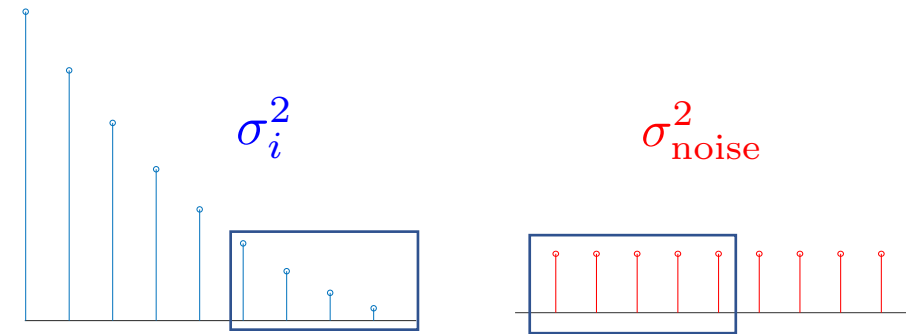
*bias!*

*Sum of the singular values not used in approximation*

## Variance

$$\mathbf{A} = \mathbf{S} + \mathbf{N}$$

*noise often isotropic  
(singular values same)*



*bias*

*variance*

$$\sum_{i=r+1}^n \sigma_i^2 \quad \text{vs.} \quad r \sigma_{\text{noise}}^2$$