

# *CS/ECE 281B – Advanced Computer Vision*

- ❖ Focus on the DL/AI development in computer vision over the past 5 years
- ❖ Though DL/AI activities are in all sub-areas of CV, our focus will be on 3 “D”
  - ❑ Detection
  - ❑ Description, and
  - ❑ Discrimination (Recognition)

# *Traditional vs. DL/AI frameworks*

- ❖ Low-level
- ❖ Flat
- ❖ Hand-crafted
- ❖ Separate processes
- ❖ High-level
- ❖ Hierarchical
- ❖ Automated
- ❖ Integrated processes

# *Traditional Detection Mechanism*

- ❖ Interesting patterns (the question of *where*)
  - ❑ Not flat
  - ❑ Not Simple
- ❖ Harris Corner Detector
- ❖ Fast Corner Detector

# *Description Mechanisms*

- ❖  $(x,y,z)$  is a descriptor of a 3D vector
- ❖  $(r, \theta)$  is a descriptor of a 2D vector
- ❖ Fourier transform coefficients form a descriptor, so are wavelet coefficients
- ❖ Multiple possibilities
- ❖ suitable for different tasks

# *High Dimensional Contents*

- ❖ Sparsity
- ❖ Correlation and redundancy
- ❖ The necessity of a paradigm shift
  - ❑ description fitting the data rather than
  - ❑ data fitting the description
  - ❑ Principal Components
  - ❑ SVD
  - ❑ Dimension Reduction
  - ❑ Etc.

# *Traditional Description Mechanism*

- ❖ Local patterns (the question of *what*)
  - ❑ Chain codes (e.g., Fast)
  - ❑ Localized and normalized gradient histograms HOG (e.g., Sift)
- ❖ Observations
  - ❑ *Where* is relatively simple and uniform
  - ❑ *What* is much harder
  - ❑ Answers to where and what can be separate processes



# *Traditional Discrimination Mechanisms*

- ❖ Not strictly a CV problem, with solutions shared with PR, ML, and AI (e.g., <http://www.cs.ucsb.edu/~cs281b/lectureMI.html> )
- ❖ A rich set of tools
  - ❑ Linear discrimination function, logistic regression, support vector machine, decision trees, simple and multi-layer perceptrons, etc.

# *What can traditional techs do?*

- ❖ Precise localization and low-level features
- ❖ Precision matching (pairing) of them over multiple frames, for accurate
  - ❑ Movement
  - ❑ Disparity
  - ❑ Homography
  - ❑ Etc.



# *Programming assignments*

- ❖ 2D lip tracking
- ❖ Picture stitching
- ❖ Stereo reconstruction
- ❖ 3D depth inference

# *What can't traditional techs do?*

- ❖ The detection/description mechanisms are separated
  - ❑ Heard a noise (easy), what kind of noise (party, fight, TV, music, etc.) is hard
  - ❑ Need better integration of the 1<sup>st</sup> 2D (face, pedestrian, vehicle)
- ❖ The description mechanism is deficient
  - ❑ Hand crafted – limited by human imagination
  - ❑ No hierarchy – describing simple patterns

# *What can't traditional techs do?*

- ❖ The discrimination mechanisms are separated
  - ❑ Integrating discrimination with the first 2D
  - ❑ Simpler network structures
  - ❑ Faster implementation

# *This Course*

## ❖ Convolution

- ❑ For discrete images
- ❑ Generalization of 2D filter (edge, corner, etc.)

## ❖ Recurrent

- ❑ For video and time dependent analysis
- ❑ A generalization of Kalman filter

## ❖ Advanced topics

- ❑ Discussions by my Ph.D. students
- ❑ Your research presentations and demos

# CNN

- ❖ Single layer
- ❖ Pre-trained weights
- ❖ Multi-layers
- ❖ Unknown weights (learned from data)
- ❖ Nonlinearity a must
- ❖ Batch normalization is useful
- ❖ Avoid saturation (Resnet and Densenet)

$x$		$y$
-1	0	1
-2	0	2
-1	0	1

-1	-2	-1
0	0	0
1	2	1

# RNN

- ❖ Known internal/external (observation) matrices
- ❖ Unknown internal state
- ❖ Unknown observation matrices
- ❖ Avoid saturation (peephole connection, projection, bi-directional, normalization, etc.)



# *Logistics*

- ❖ Many variances
  - ❑ Tensorflow (Google)
  - ❑ Pytorch (Facebook)
  - ❑ Mxnet (Apache, Amazon)
- ❖ Tensorflow is running in CSIL
  - ❑ CPU, not GPU
  - ❑ Cannot do much without GPU
- ❖ Python encoding only (Matlab? C++?)