Visual Speech Recognition

Tian Zhang
Tongshuang Wu
Yi-Sheng Hsieh

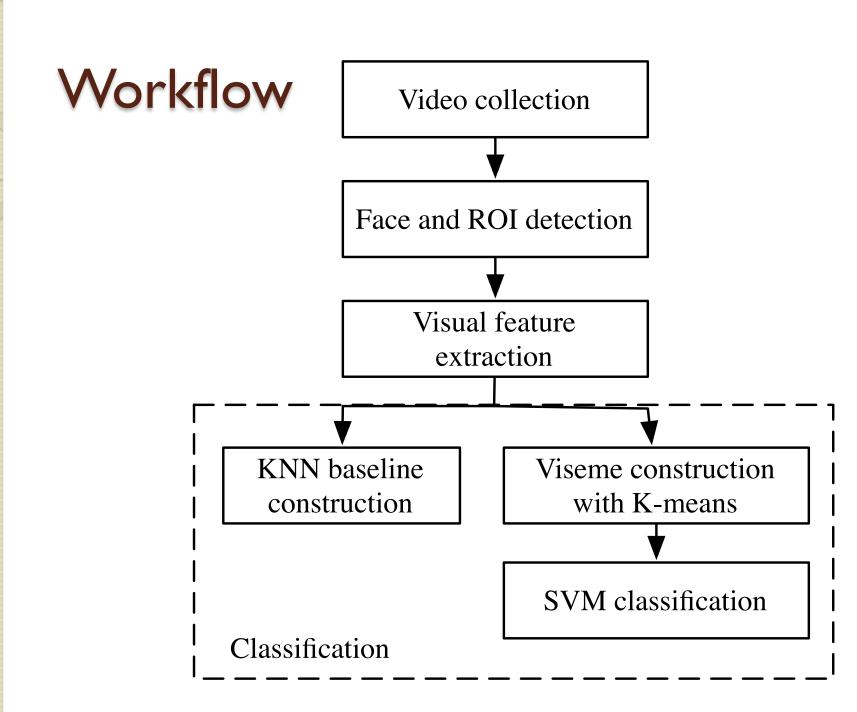


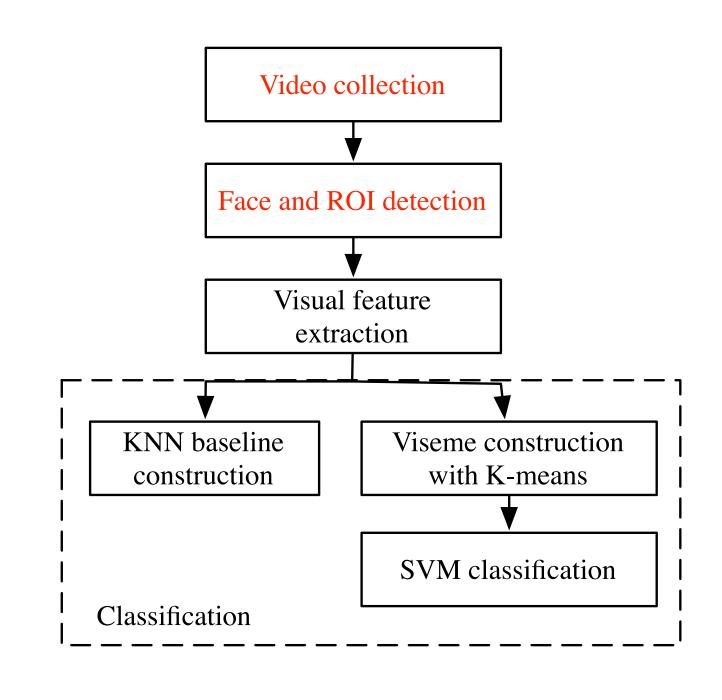
Visual Speech Recognition?

What? lip reading by computer!!
How? Lip motion recognition
Why? - Useful when audio information cannot be effectively conveyed
- Great social impact: for those with auditory dysfunction

What work have been done?

- Several specialized visual system for:
 - Digits
 - Chinese characters
 - Etc.
- Why it's not good enough:
 - Visual clues + automatic speech recognition
 - How about extend visual features to the full extend?





Dataset

- Video source : self-made
- 2 people
- 2 word : beach dark
- 10 times

Mouth detection

- MouthDetect = vision.CascadeObjectDetector ('Mouth','MergeThreshold',40);
 - Viola-jones algorithm
 - Classification model = mouth
 - MergeThreshold

Sample result

• Word : Beach

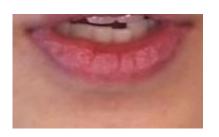


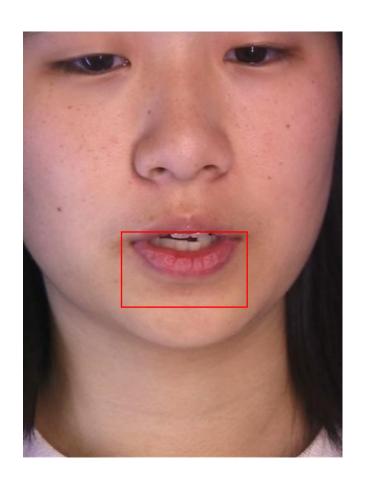
Dataset

- Video source : CMU-AMP database
- 3 people
- 150 word
- 10 times

Problem

Imperfect ROI





Observation

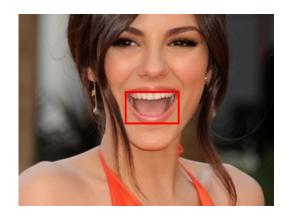


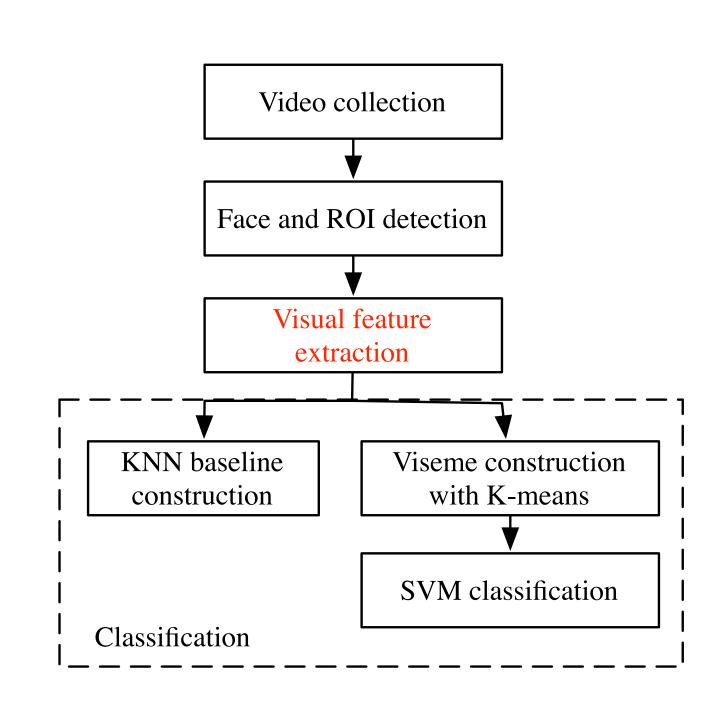










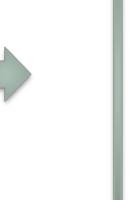


Matrix Representation



8 x I Vector Representation

Width
Length
Mutual Information
Quality Measure Feature
The ratio of vertical to horizontal features
The ratio of vertical to horizontal edges
The proportion of tongue
The proportion of teeth



Feature Description

- Width, Height
- Mutual Information: temporal correlation between two frames

$$M(X;Y) = \sum_{x} \sum_{y} p(x,y) \log \left(\frac{p(x,y)}{p(x)p(y)} \right)$$

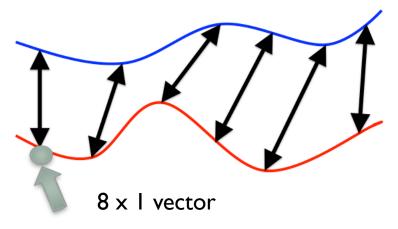
Quality Measure Feature: distortion between two frames

$$Q = \frac{4\sigma_{xy}\overline{x}\,\overline{y}}{(\sigma_{x}^{2} + \sigma_{y}^{2})[(\overline{x})^{2} + (\overline{y})^{2}]}$$

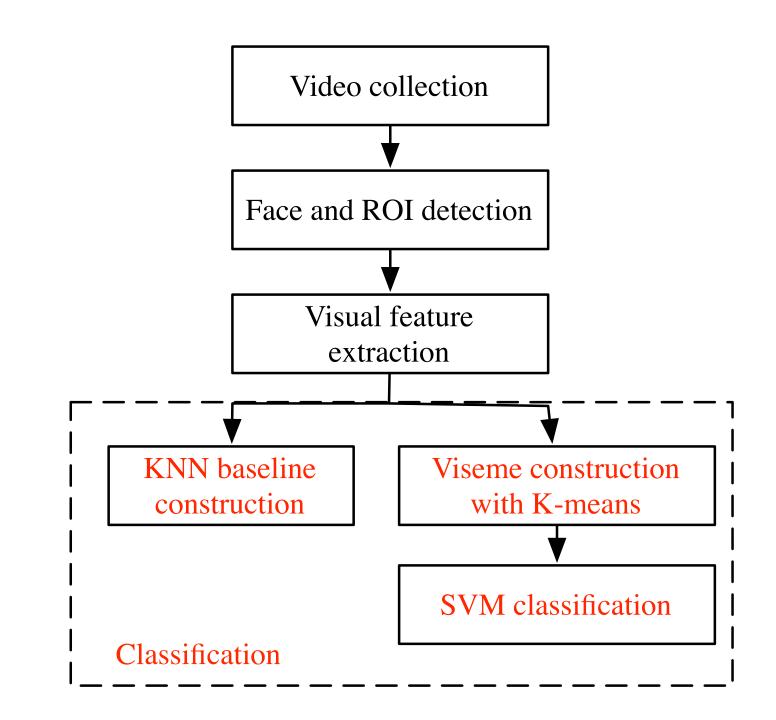
- The ratio of vertical to horizontal features: High deviation from the mean in the Discrete Wavelet Transform domain is usually associated with feature.
- The ratio of vertical to horizontal edges: Sobel Edge detector
- The proportion of tongue: proportion of dark red color
- The proportion of teeth: proportion of white color

Dynamic Time Warping

• DTW: Distance between two time series of different length.



Can be used in K Nearest Neighbor

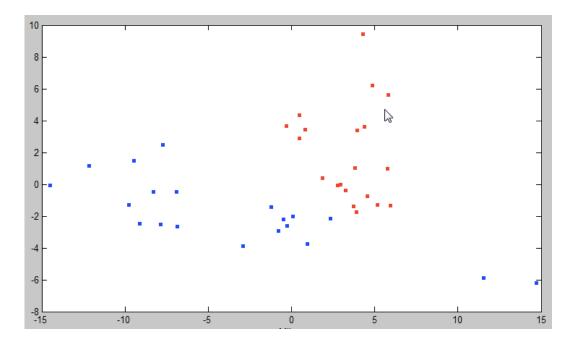


Classification

- Most popular techniques:
 - K Nearest Neighbors (used as baseline!)
 - Hidden Markov Model (borrowed its idea for state construction!)
- What we used:
 - HMM-like data processing, with kmeans and SVM technique

Baseline: KNN

- K-Nearest Neighbor algorithm
- Two words spoken by three people
- Accuracy = 75%



Hybrid K-means SVM

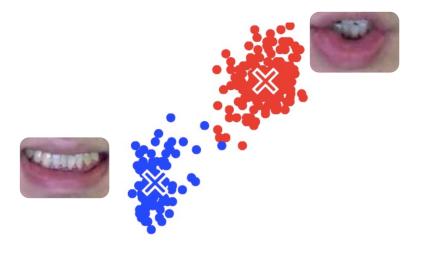
K-means on feature vectors to cluster viseme with cluster centroid, with cross validation for determining k

Construct k by k feature matrix for all the samples, each entry (i, j) means the probability of transforming from state i to state j

Use training data to train the SVM, with states being a cluster resulted in K-means

Process test data in the same way as did to training data,

Compare test and training model similarity with 1NN



12331132....

[0.1]	0.5	0.09
0.11	0.25	0.01
10.08	0.1	0.31

Hybrid K-means SVM

Results comparison using provided features, full size data				
	CMU features (only used height and width)	Our features		
KNN	N.A.	~75%		
Kmeans-SVM	~80%	~100%		

- Feature out-performs!
- Kmeans-SVM works!
- How about multiple words?

Guideline

- Problem description and literature review
- Technical and experimental description
 - Region of Interest detection
 - Feature extraction
 - Classification
- Future work

Future Work

- Use customized model for better ROI extraction
- Shoot some more standard videos on our own for classifying multiple word testing

Results comparison	using	provided	features, full	size		
data						

	CMU features	Our features
KNN	15%	?
Kmeans-SVM	?	?

Reference

- [1] G.Erten.Audiovisualspeechprocessing,2001.
- [2] C.G.Fishe, .Confusions among visually perceived consonants. *Journal of Speech, Language, and Hearing Research*, 11(4):796–804, 1968.
- [3] A. B.A. Hassanat. Visual speech recognition. CoRR, abs/1409.1411, 2014.
- [4] O. H. Jensen. *Implementing the Viola-Jones face detection algorithm*. PhD thesis, Technical University of Denmark, DTU, DK-2800 Kgs. Lyngby, Denmark, 2008.
- [5] K. Kumar, T. Chen, and R. M. Stern. Profile view lip reading. In Acoustics, Speech and Signal Processing, 2007. ICASSP 2007. IEEE International Conference on, volume 4, pages IV—429. IEEE, 2007.
- [6] K. Messer, J. Matas, J. Kittler, and K. Jonsson. Xm2vtsdb: The ex-tended m2vts database. In *In Second International Conference on Au- dio and Video-based Biometric Person Authentication*, pages 72–77, 1999.
- [7] A.V. Nefian, L. Liang, X. Pi, L. Xiaoxiang, C. Mao, and K. Mur-phy. A coupled hmm for audio-visual speech recognition. In *Acous-tics*, *Speech*, *and Signal Processing (ICASSP)*, 2002 IEEE International Conference on, volume 2, pages II–2013. IEEE, 2002.
- [8] T.-L.Pao, W.-Y.Liao, and Y.-T. Chen. Audio-visual speech recognition with weighted knn-based classification in mandarin database. In *In-telligent Information Hiding and Multimedia Signal Processing*, 2007. *IIHMSP* 2007. Third International Conference on, volume 1, pages 39–42. IEEE, 2007.
- [9] L. Rabiner. A tutorial on hidden markov models and selected applications in speech recognition. *Proceedings of the IEEE*, 77(2):257–286, 1989.

Thanks! Any questions?





