

# Clustering Applications in Biometrics

Biometric Works



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# Agenda

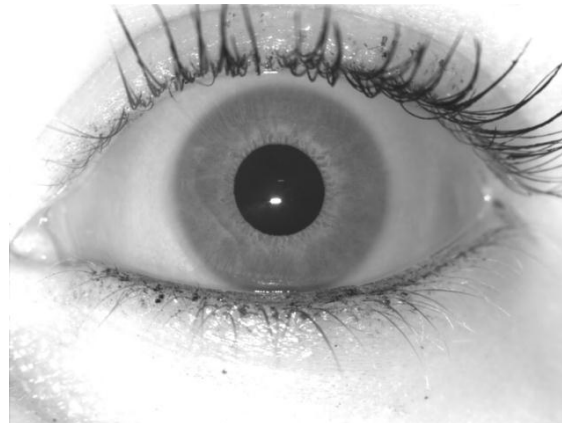


- Clustering in Iris Recognition
  - Binning
  - Hadoop and Biometric Databases
  - Pattern Recognition
- Clustering in Keystroke Dynamics
  - Three feature clustering

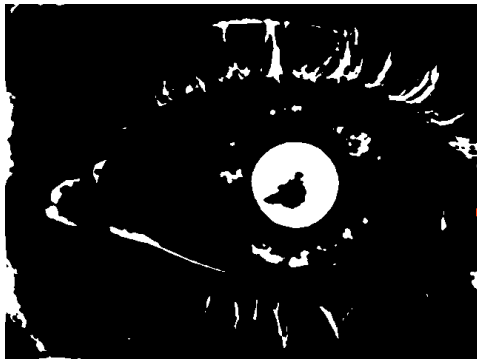
# Iris Recognition



- Steps:
  - Locate Pupil
  - Locate Limbic Boundary
  - Generate Template
  - Compare Templates



# Pupil Location



The original filtered image.



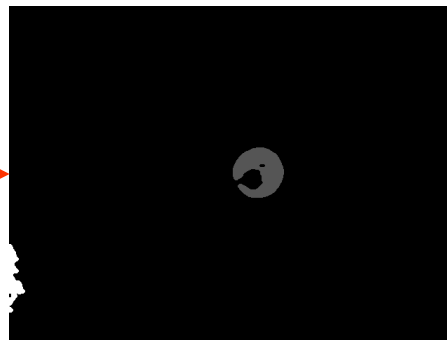
The eroded image.



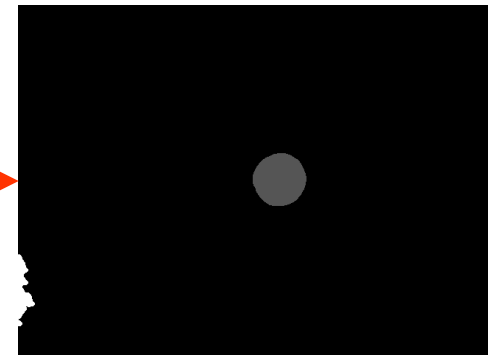
The image after erosion and dilation.



The image after erosion and dilation.



The image after labeling.  
Objects below a certain size are ignored.

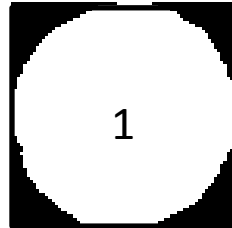


The labeled image after being filled.

# Pupil Location



- Pupil Selection:

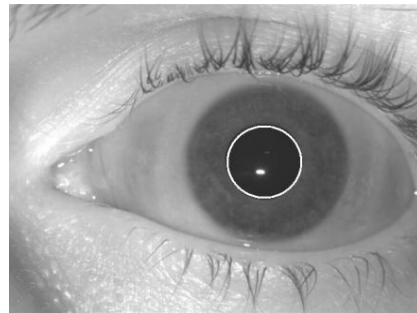


- Criteria:
  - Size
  - Eccentricity
  - Roundness

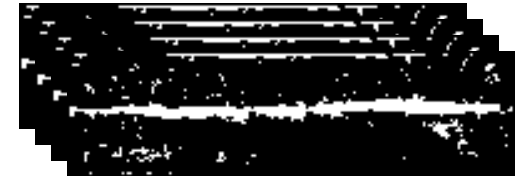
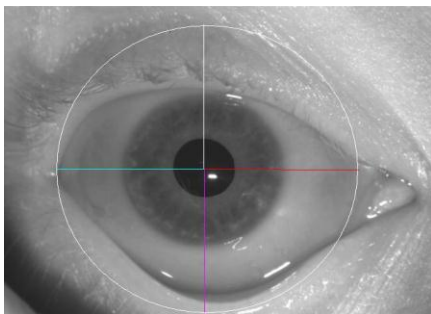
# Limbic Location



- Given an image and a pupil location



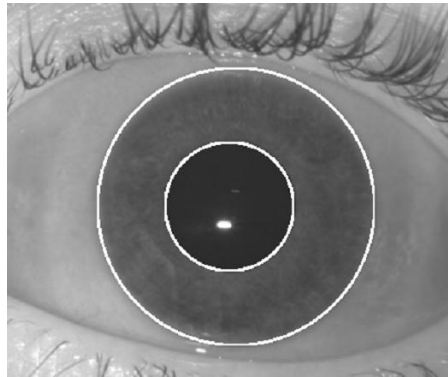
- “Unwrap”, filter, find straightest line



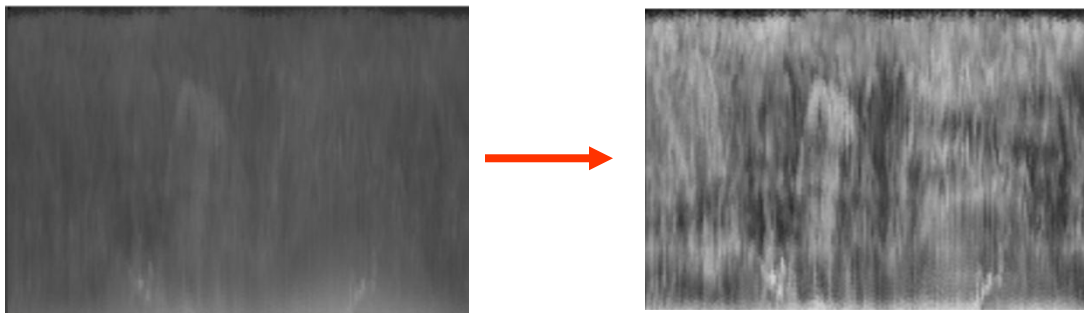
# Template Generation



- Given an image, a pupil location, and a limbic location



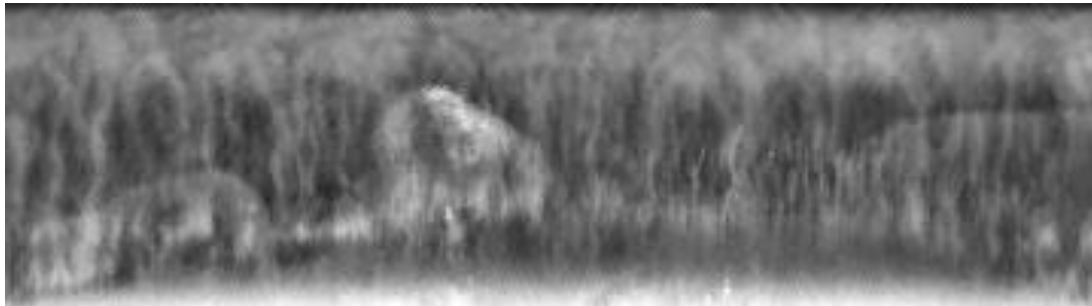
- Unwrap and contrast adjust



# Template Generation



- Feature Extraction



- What features are algorithm specific (and highly proprietary)
- Extracted using feature kernels and convolution ([wiki](#))



# Template Generation



- Get a matrix of convolution responses
- Threshold and convert to binary



# Template Comparison



- Every iris image becomes a template
  - Small
  - Easier to compare
  - Faster
- Comparing two templates = a distance metric
- Hamming Distance:

010111001
010101011

# Template Comparison



- Imposter Comparisons:



“Bill”



“Joe”

- Should have a fractional Hamming distance of  $\sim .5$

# Template Comparison



- Genuine Comparisons:



“Bill Yesterday”



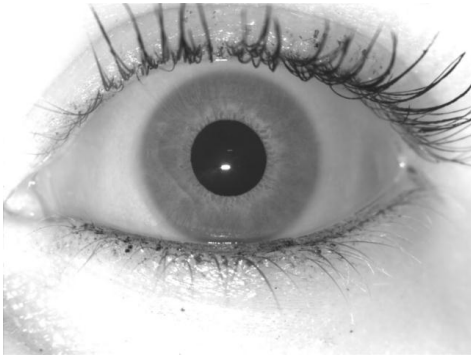
“Bill Today”

- Should have much lower fractional Hamming distance
- Can model both as a normal distribution

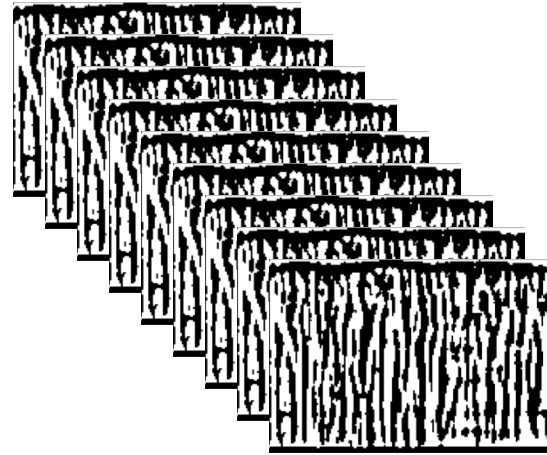
# Iris Recognition



- Binning



Who is this?



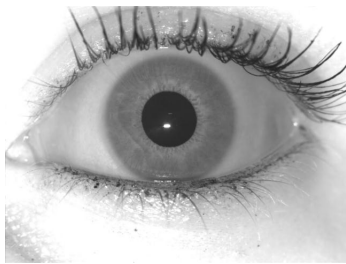
Search all these

- Binning reduces the amount of templates you have to search (prune the search space)

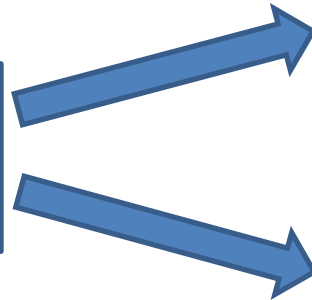
# Iris Recognition



- Naive Binning



Left or  
Right  
Eye?



All left  
templates



All right  
templates

- No “natural” order
- False no-match if wrong bin is searched

# Iris Recognition

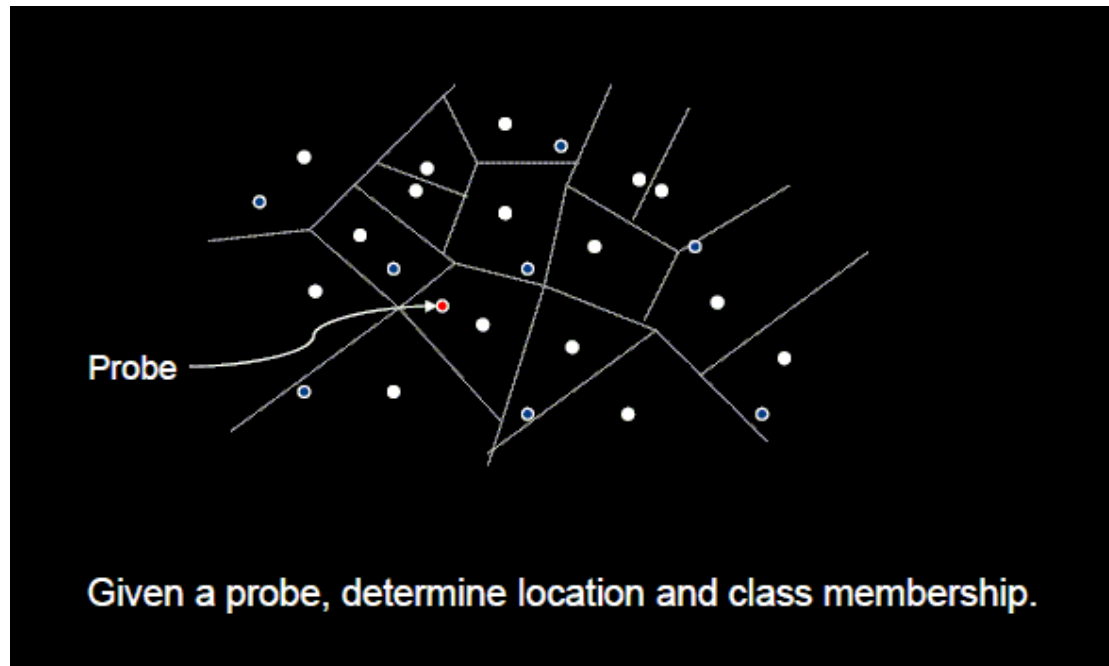


- Feature Based Binning (Mehrotra, 2010)
  - 27 features of each template identified
    - Center of gravity
    - Area of black pixels
    - Etc.
  - Cluster an existing database based on these features
  - When a new probe image is introduced it gets “fuzzily” assigned to a cluster
  - The template comparison routine only searches those clusters

# Iris Recognition



- Similarity Based Binning (Becker, Potts, 2007)
  - Don't bin on features
  - Bin on a similarity metric





# Hadoop and Iris Recognition

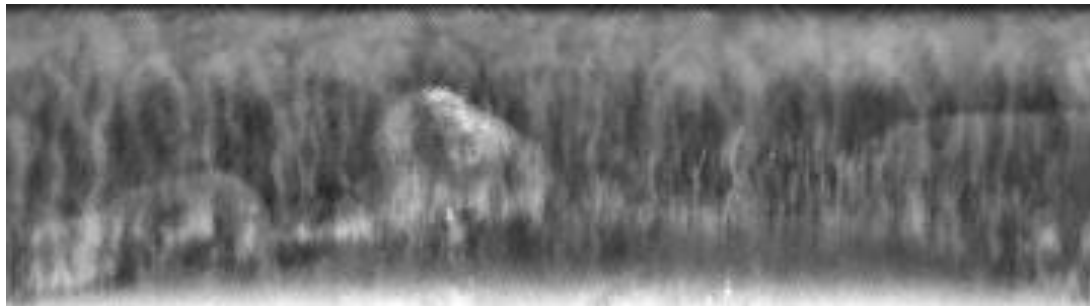


- Trost, 2010
- HDFS provides a framework for distribution, and redundancy of database
- Mahout/MapReduce to find clusters to reduce search space
- {Key, Value} stored in redundant SequenceFiles to allow for parallel search.

# Iris Recognition



- Clustering to find patterns (pattern recognition)



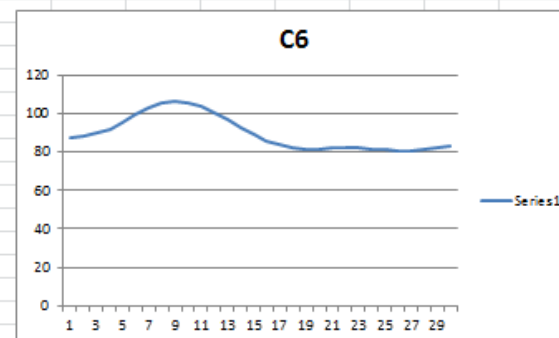
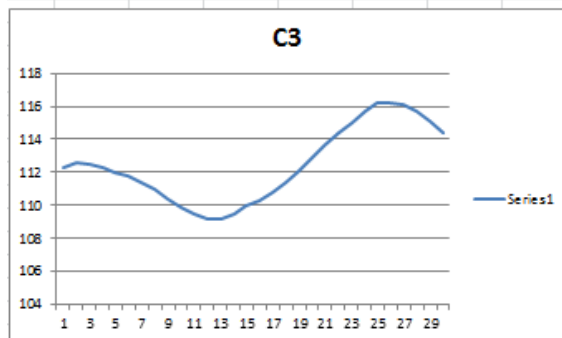
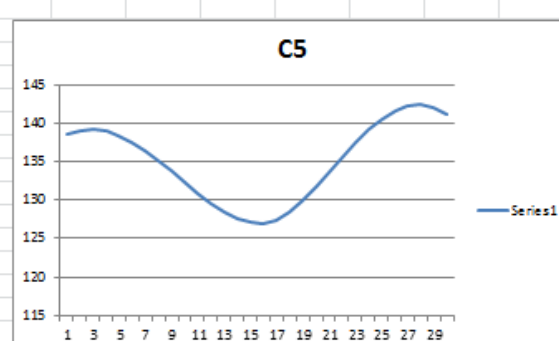
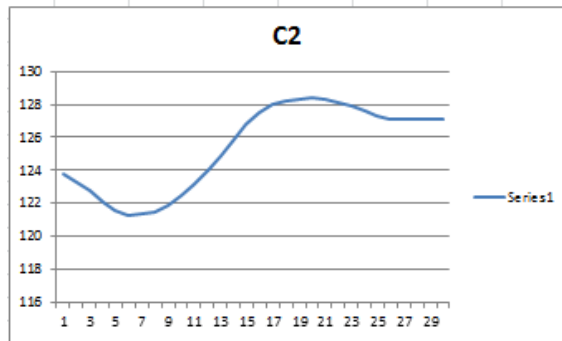
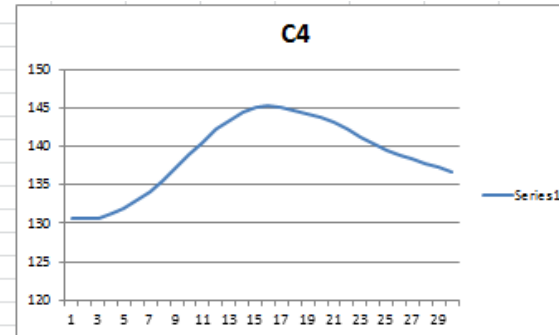
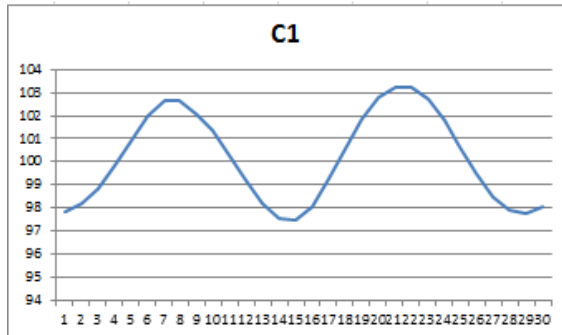
- Use sliding window of size  $k$
- Generate 36,000 row vectors
- Cluster using k-means or k-medoids using a high number of clusters

# Iris Recognition



- Resulting Clusters should be most common patterns in the image
- Can then use these to generate kernels that get extremely high responses

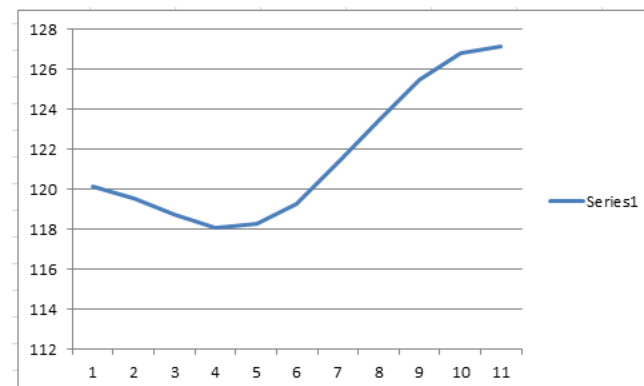
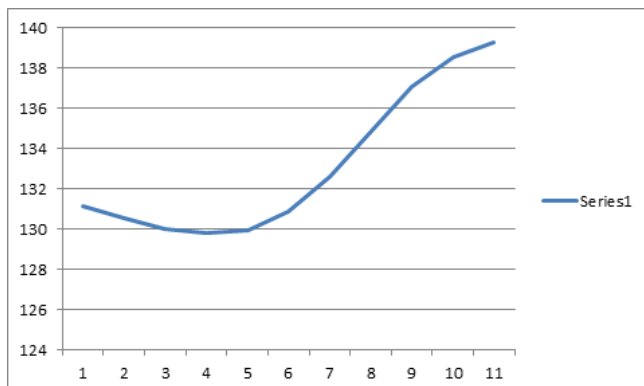
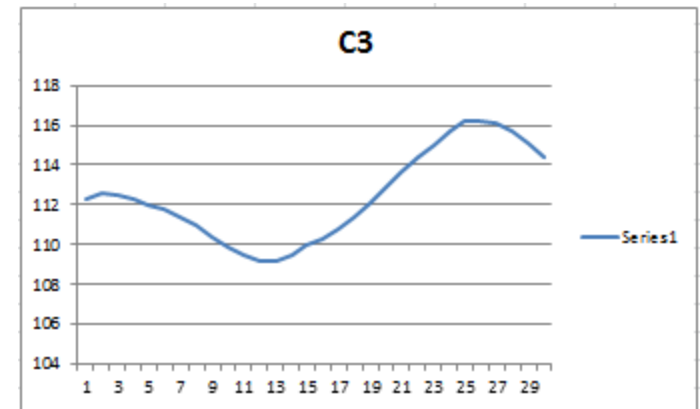
# Iris Recognition



# Iris Recognition



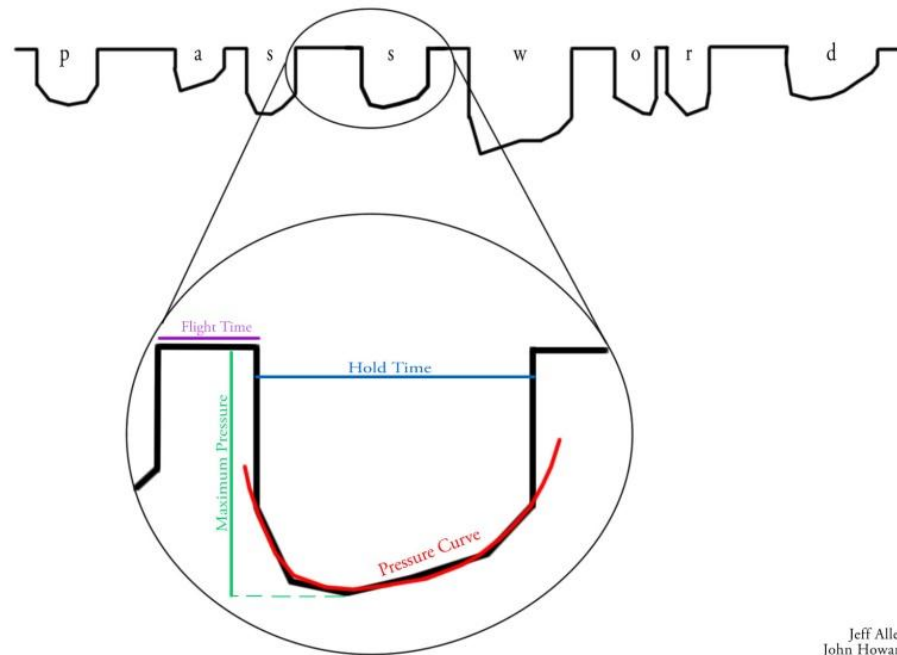
- Most common pattern not necessarily the most interesting
- Need to filter the vectors
  - Must have a certain % change
  - Must be normalized to the same starting value



# Keystroke Dynamics



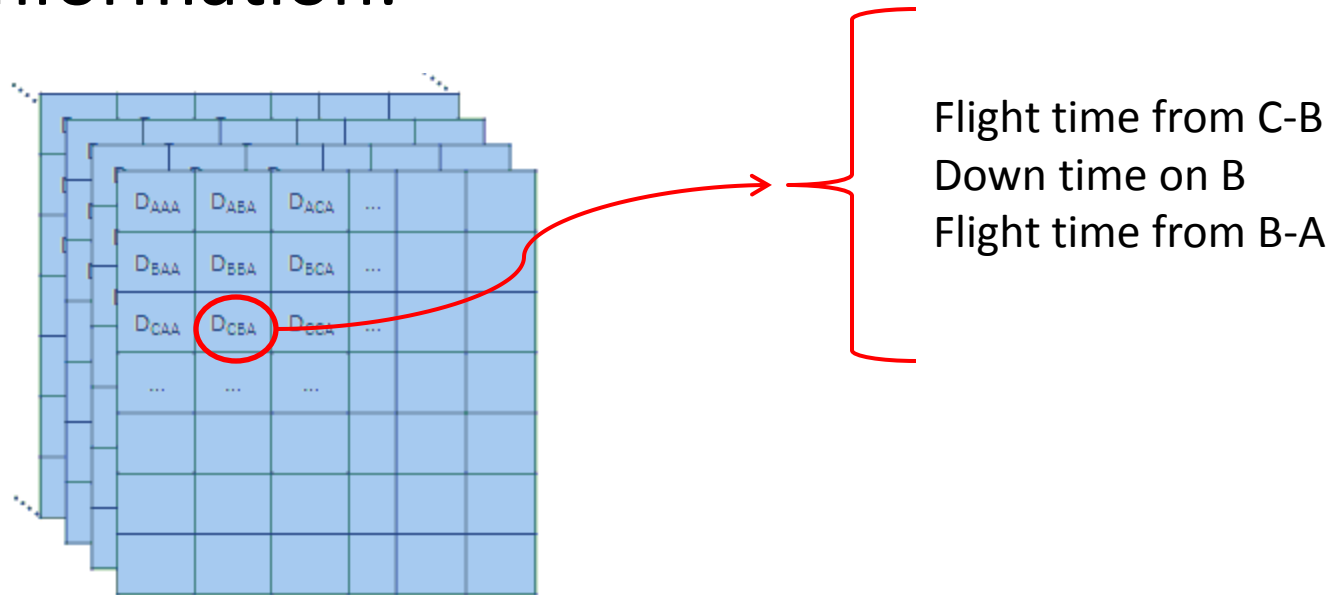
- Identify people based on their typing patterns



# Keystroke Dynamics



- Build up 3 dimensional matrices of timing information:



- Plot in 3D space
- Cluster (show example)

# Clustering Applications in Biometrics



- Biometrics and Clustering are closely related
- Many applications across the field
- More uses in the future
- Question?



# References



- Trost, J. Hadoop for Large Scale Biometric Databases. Hadoop Summit 2010.
- Becker, G. and Potts, M. Non-metric Biometric Clustering. Biometric Consortium Conference, 2007.
- Mehrotra, H., et. al. Feature Level Clustering of Large Biometric Databases.