## **Boosting the Efficiency of Target Face Recognition through Image Hybridization**

Michael NETT<sup>1,2</sup> Jichao SUN<sup>3</sup> Vincent ORIA<sup>3</sup> Shin'ichi SATOH<sup>1</sup> Michael E. HOULE<sup>1</sup> <sup>2</sup> University of Tokyo <sup>3</sup>New Jersey Institute of Technology <sup>1</sup> National Institute of Informatics

## Why

One of the great challenges in security is the ability to efficiently and accurately recognize individuals of interest from images or video streams. Face recognition of target individuals is problematic due to the high-dimensional representations of faces that are necessary for high accuracy, and the need to avoid the erroneous identification of targets.



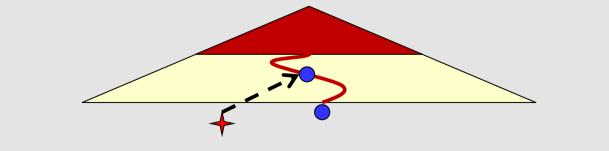
One popular way of reducing false positive errors is by searching for targets within a large database dominated by control images. We propose techniques for boosting the efficiency of handling identification queries by means of hybridizations between target faces and control faces. Hybridization improves indexability while still allowing for reasonable query times.

A Cautionary Tale			
TOSHIBA NEDO PROJECT	POSSIBLE SOLUTION?	PERFORMANCE	INDEXING EXPERIMENT
<ul> <li>scale Systems for Human Face Recognition" (2009 – 2011).</li> <li>Goal: fast recognition of a limited number of target faces from video streams (10<sup>2</sup>→10<sup>3</sup>).</li> <li>Example: terrorist suspects from airport surveillance video.</li> </ul>	Build large background set of <u>control</u> (non-target) images (~10 <sup>6</sup> ). Combine targets and controls into a single index. For each query, return <i>k</i> most similar faces ( $k = 10^2 \rightarrow 10^3$ ). Within each query result, check for instances of target faces, and verify.	<ul> <li>&gt; Toshiba face vector → 2560 features.</li> <li>&gt; 98%+ accuracy using sequential search.</li> <li>&gt; Time to execute a single query on 10<sup>6</sup> faces: 25s!</li> <li>&gt; More efficient indexing strategy is needed</li> </ul>	<ul> <li>Indices considered: LSH, Cover Tre SASH, RCT</li> <li>Data sets considered:         <ul> <li>Toshiba Faces (10<sup>6</sup> morphs generated from 9863 Toshiba employee ID photos).</li> <li>WikiFaces (2* 10<sup>5</sup> public domai face images crawled from the Wikimedia Commons archive)</li> <li>Both vectorized using Toshiba features (2560 dimensions).</li> </ul> </li> </ul>
<ul> <li>OUTCOMES</li> <li>&gt; Uniformly poor!</li> <li>&gt; Trade-off between accuracy and time essentially no better than sequential search.</li> <li>&gt; However, some sets with dimensionality on the order of 10<sup>5</sup>→10<sup>6</sup> are indexable.</li> </ul>	<ul> <li>Proposed measure of the intrinsic dimensionality of data, generalized the expansion dimension of Karge Ruhl.</li> <li>Dimension <i>m</i> can be computed expression volume and radii of spheres.</li> </ul>	er & log Vol $(a, r_a)$ - log Vol $(a, r_a)$	Intrinsic dimension of face data is much higher than that observed in indexable data. $\int_{\text{UDEXABILITY}}^{14\%} \int_{\text{UDEXABILITY}}^{12\%} \int_{\text{UDEX}}^{12\%} \int_{\text{UDEXABILITY}}^{12\%} \int_{\text{UDEX}}^{12\%} \int_{\text$
• Example: Reuters2 Corpus (subset): 554,651 documents, 320,647 keyword dimensions.	<ul> <li>In practice, volumes can be estimply numbers of points enclosed in balls.</li> <li>Distributional assumptions depensimilarity measure.</li> <li>Max expansion dimension: max v m' over range of choices of k<sub>1</sub> and m' = log k<sub>2</sub> - log k<sub>1</sub>/log r<sub>2</sub> - log r<sub>1</sub></li> </ul>	the d on alue of d $k_2$ :	$ \sum_{y=1}^{3} \sum_{$
<ul> <li>Question: why are some high-dimensional data sets indexable? Why are faces not?</li> </ul>	Average expansion dimension: us for characterizing difficulty of data and difficulty of queries.	setul sets	Conclusion: similarity values are meaningless across different individuals

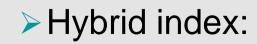
## **Hybridization Strategy**

## Promote transitivity by replacing control faces by <u>hybrids</u> between targets and controls.

- Each target face is more similar to its hybrids than to other faces; other face relationships have no particular similarity.
- At lower levels of the index, positive queries require that the target have at least one hybrid node that is similar to the query.



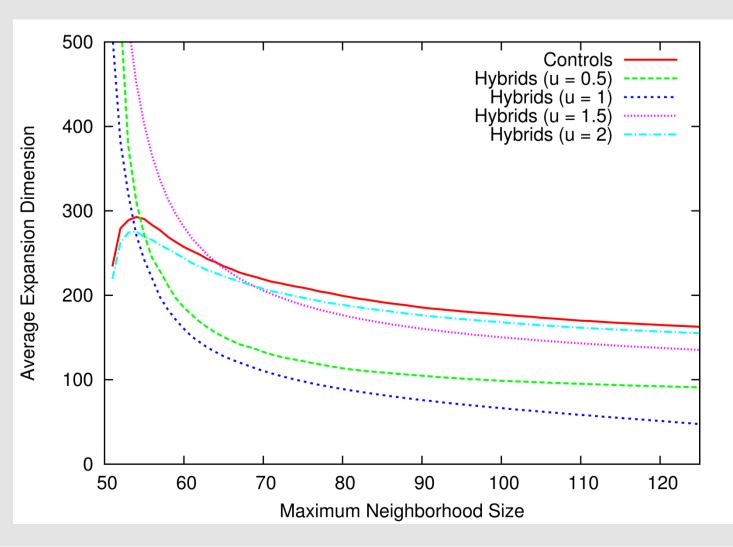
- > Search:
  - Visit all nodes at top levels of the structure.
  - At lower levels, visit a limited number of nodes similar to the query.
  - Classify by voting among query result hybrids as to whether one target dominates.



Target set *T* of size *m*.

 $\mathbf{O}\mathbf{V}$ 

- Number of hybrids per target: m<sup>u</sup>. 0
- Probability of level *i* containing no hybrids of a given target is at most  $\Pr[H(v) \cap L_i] \leq e^{-m^u/2^i}$
- >  $u=1 \Rightarrow O(m)$  query time in index of lower intrinsic dimensionality.
- Classification: not yet assessed.



連絡先: Michael E. HOULE (フール マイケル) / 国立情報学研究所 客員教授 **TEL: 03-4212-2538** Email : meh@nii.ac.jp