

## Face Retrieval for Security Applications

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Face Retrieval for Security Applications

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Detection of faces





Detection of faces



Extraction of face features

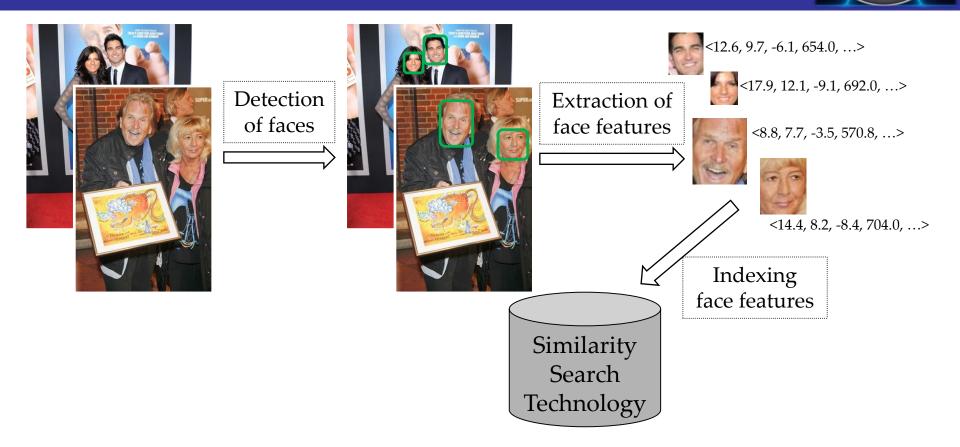
<12.6, 9.7, -6.1, 654.0, ...>

<8.8, 7.7, -3.5, 570.8, ...>

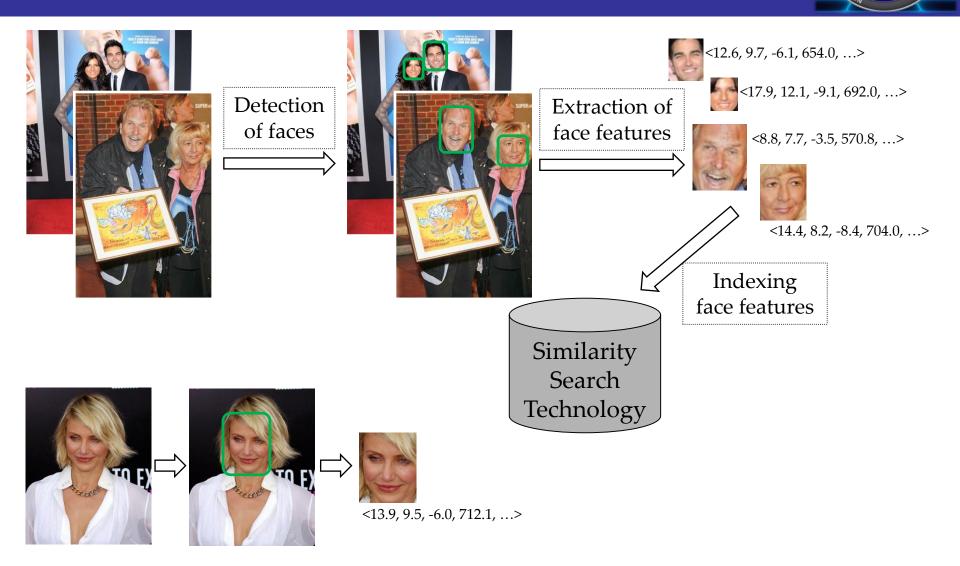


<14.4, 8.2, -8.4, 704.0, ...>

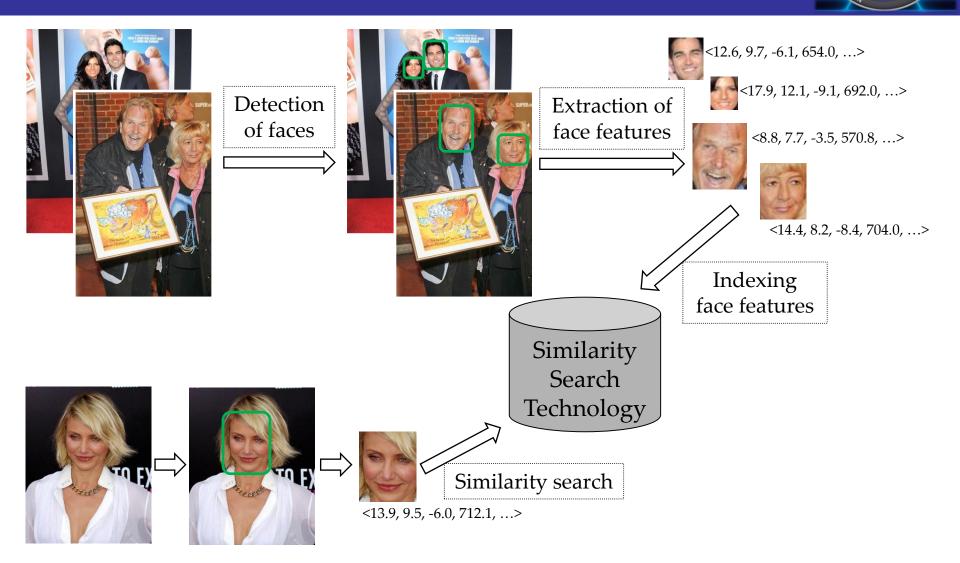
Workshop



Workshop



Workshop

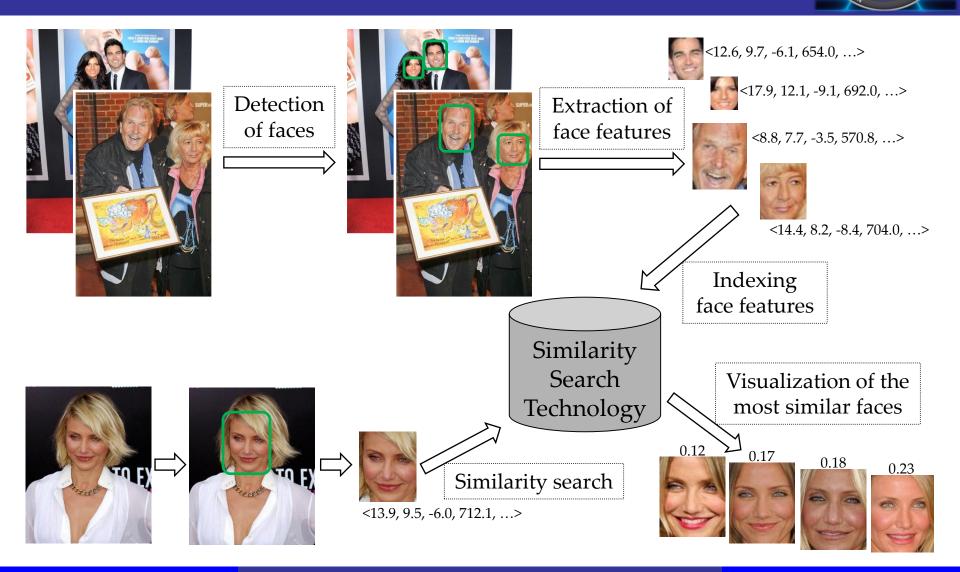


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Workshop

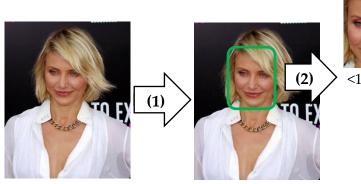


Workshop

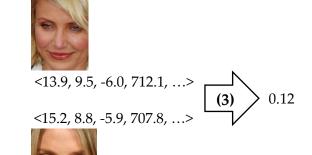
# Face Recognition Technology: Components

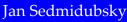


- Components of general face recognition technology
  - 1) Face detection function
    - Localizes bounding boxes of faces within a static image
  - 2) Feature extraction function
    - Extracts a characteristic feature of a face image
  - 3) Face recognition function
    - Computes similarity score between features of two faces



<13.9, 9.5, -6.0, 712.1, ...>





- Our objective to effectively and efficiently retrieve the most similar faces to a query face(s)
- Improvement of detection/recognition components:
  - 1) Face detection function combining multiple state-ofthe-art approaches for more effective detection
  - 2) Feature extraction func. aggregating multiple features
  - 3) Face recognition function
    - Combining multiple approaches for more effective recognition
    - Multi-face query evaluation for more effective retrieval
    - Pre-selection of candidate set and its re-ranking for more efficient retrieval

• Aggregation of existing detection approaches to increase recall/precision



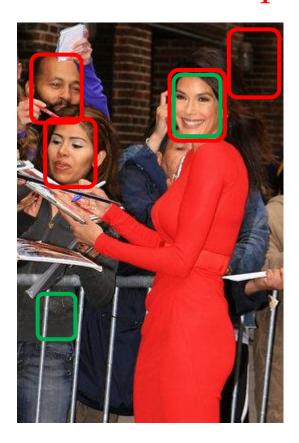
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• Aggregation of existing detection approaches to increase recall/precision



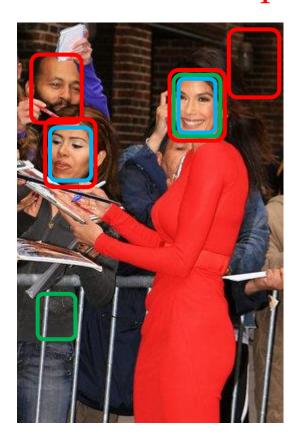
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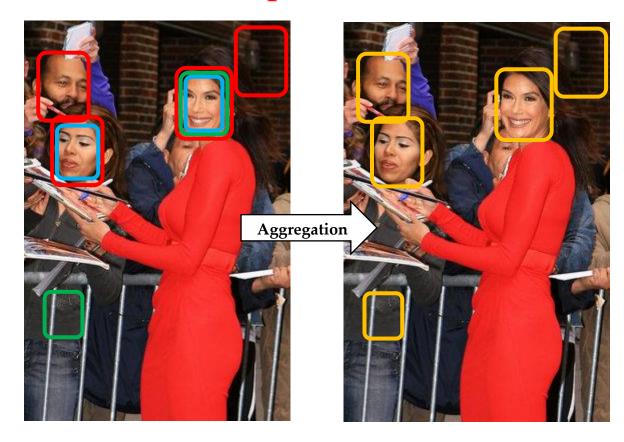
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• Aggregation of existing detection approaches to increase recall/precision



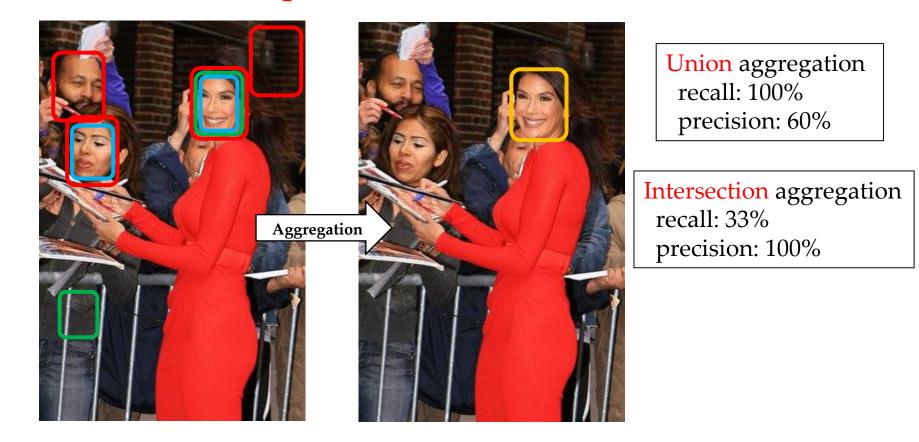
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• Aggregation of existing detection approaches to increase recall/precision

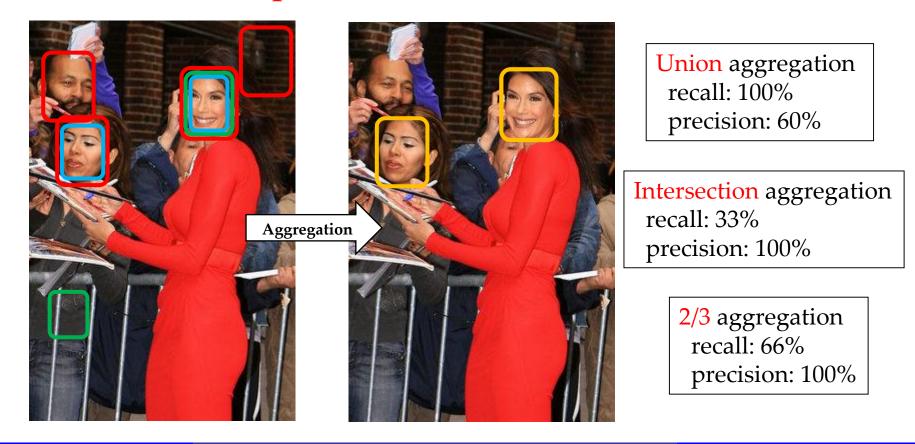


Union aggregation recall: 100% precision: 60%

• Aggregation of existing detection approaches to increase recall/precision



• Aggregation of existing detection approaches to increase recall/precision

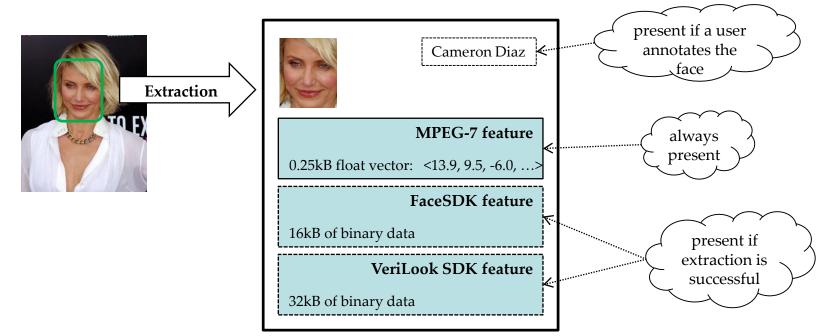


- Current implementation:
  - Aggregation of at least 2 approaches out of 3
    - OpenCV open source from Computer vision library
    - FaceSDK commercial software from Luxand
    - VeriLook SDK commercial software from Neurotechnology

	Low-quality face images # of faces: 1000		High-quality face images # of faces: 10,000	
	Recall	Precision	Recall	Precision
OpenCV	55	89	92	86
FaceSDK	63	83	95	94
VeriLook SDK	73	84	100	96
Our aggregation	62	98	97	100

#### Our Face Recognition Technology: Feature Extraction

- Utilization of existing extraction approaches to characterize a face by a set of different features
- Current version integrates three approaches of MPEG-7, FaceSDK and VeriLook SDK features

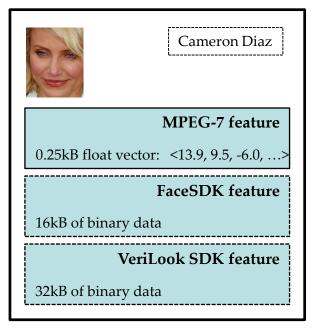


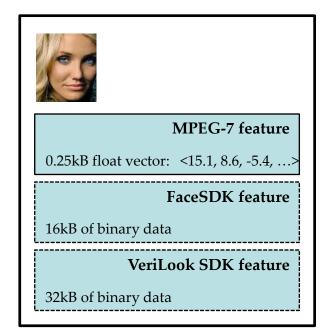
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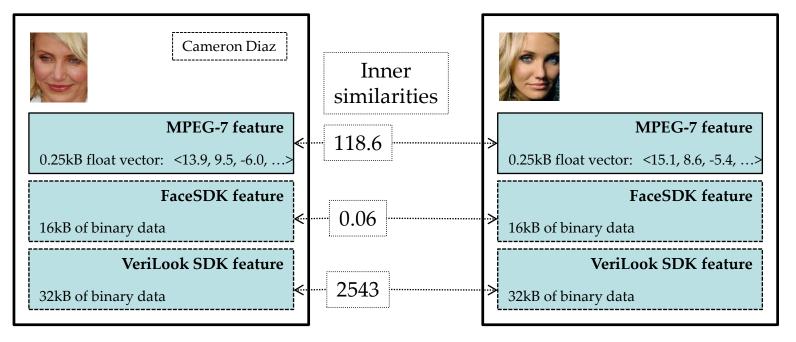


• Similarity of two faces is computed as aggregation of normalized inner similarities between features

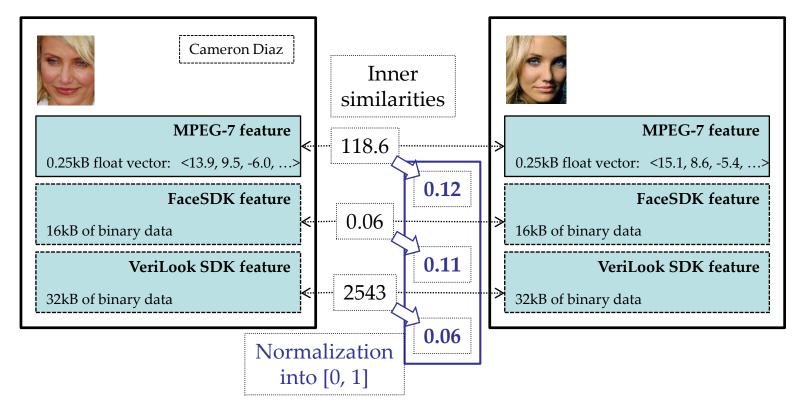




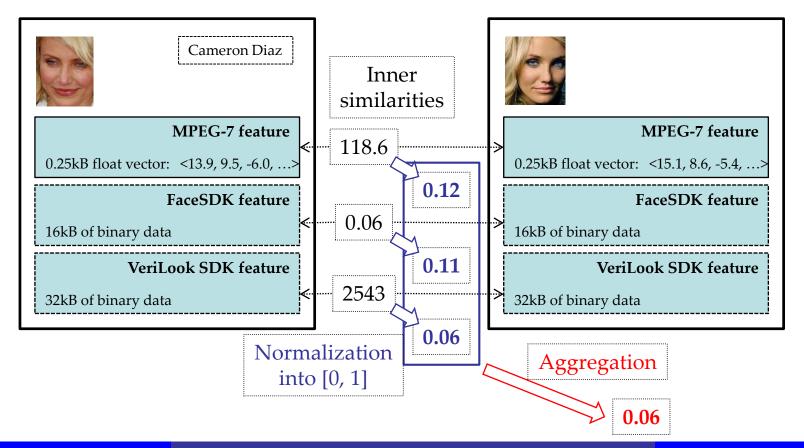
- Workshop 2014
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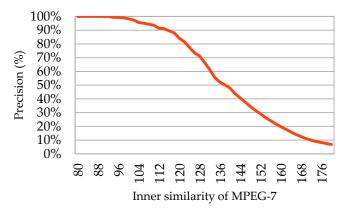


• Similarity of two faces is computed as aggregation of normalized inner similarities between features





- Inner similarities are computed by integrated recognition approaches themselves
- Each inner similarity is then normalized according to specifically learned normalization function
  - Normalization func. learned from training data (example)



Normalized value = 100% - <precision of inner similarity>

Example: inner similarity =  $118.6 \Rightarrow \text{precision} = 88\% \Rightarrow$ normalized value =  $100\% - 88\% = 12\% \Rightarrow 0.12$ 

• Aggregation is minimum of normalized similarities

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- Workshop 2014
- Aggregation approach increases the quality of face recognition in terms of recall and precision
- Face images of high/low quality have a distance between eyes >100/<30 pixels</li>

	Low-quality face images # of faces: 1000		High-quality face images # of faces: 10,000	
	<b>Recall</b> precision=85%	<b>Recall</b> precision=95%	<b>Recall</b> precision=85%	<b>Recall</b> precision=95%
MPEG-7	24	14	8	3
FaceSDK	23	16	14	0
VeriLook SDK	12	11	53	51
Our aggregation	31	24	54	51

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- Multi-face query:
  - Query composed of a number of reference faces



2nd iteration

1st iteration



- Relevance feedback on 1M dataset:
  - Manual selection of positive (correct) retrieval results
  - Iterative search where positive results represent query faces
  - 1st iteration: **precision=6%**, 5th iteration: **precision=30%** (k=100)



- Preprocessing (indexing) phase:
  - All faces are indexed according to the MPEG-7 feature
    - M-index (Novak, 2012) structure is utilized
    - Indexing of 1M faces takes ~10minutes
- Efficient (scalable) retrieval of 100 faces:
  - Candidate set of 10,000 the most similar faces is efficiently retrieved by M-index (only MPEG-7 feature is utilized)
    - Retrieval out of 1M faces takes ~0.15s (~10s without M-index)
  - Retrieved 10,000 faces are sorted (re-ranked) according to our aggregation approach
  - Top-ranked 100 faces are returned



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    Increases
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Increases effectiveness efficiency

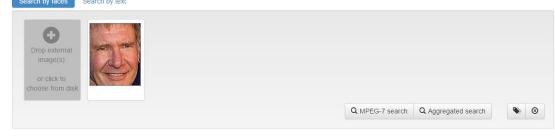
## Our Face Recognition Technology: Summary



- Advantages against other face recognition systems:
  - Support for multi-face query evaluation
  - Aggregation of multiple existing approaches for more accurate detection and recognition
    - Significant improve in the recognition accuracy (naive approach: 6% accuracy, our approach: 40% accuracy)
  - Tradeoff between the recognition accuracy and performance can be controlled by a user
    - Online search within millions of faces

## Our Face Recognition Technology: Demonstration Application

- Demonstrated by an online web application:
  - Implemented within Java + Apache Tomcat
  - Commercial licenses of FaceSDK and VeriLook SDK
  - 1 million database of extracted faces
  - <u>http://disa.fi.muni.cz/FaceMatch/</u>



Search results of double enhanced aggregated function (100)



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#### Thank you for your attention.

Try our online web application: <u>http://disa.fi.muni.cz/FaceMatch/</u>

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