

Using Video Compression to Exploit Similarities in Biometric Databases

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Objective

General incentive

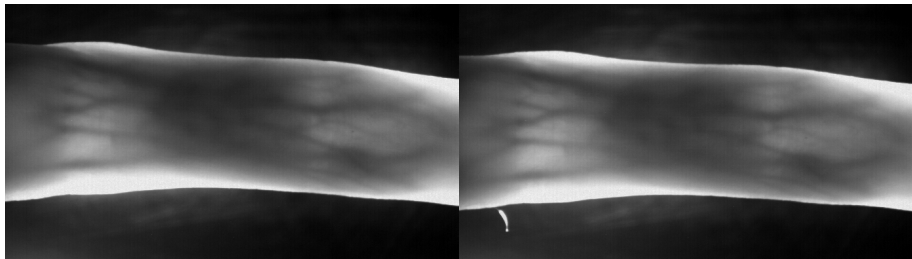
Is it possible to effectively apply **video compression** to very similar **pictures**?

- can we achieve *better* results with video compression than with image compression?
- which codec is best suited for our purposes?
- how does the change of video codec parameters affect the results?
- how to determine the quality of the results?

Dataset

The database consists of finger vein images of different fingers of different persons

- 6 fingers per person, with 4 pictures per finger \rightarrow 24 pictures per person
- 60 persons at all
- we worked with a subset of those



Video Compression

Why video compression?

Video Compression

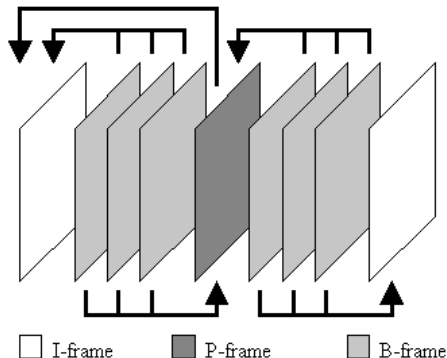
Why video compression?

- Very similar images
- Image compression only compresses individual images
- Video compression does 2 things:
 - 1 Compresses images
 - 2 Exploits similarities between images

I,P,B-Frames

3 different types of pictures

- I-Frame: Intra-coded picture
- P-Frame: Predictive-coded picture
- B-Frame: Bidirectional predictive-coded picture



Group Of Pictures

- usually defined with two numbers
 - ① defines distance of two I-Frames
 - ② defines distance of two anchor frames (I or P)
- we used GOP to adapt the encoding to the database
 - 24 pictures per person: use GOP 24 \rightarrow 1 I-Frame per person
 - 4 pictures per finger: use GOP 4 \rightarrow 1 I-Frame per finger
- P- and B-Frames allow higher compression \rightarrow GOP affects the compression rate

JPEG2000

- used as a baseline for comparison
- standard encoding settings, except number of layers
- ImageMagick (7.0.5-10) with integrated OpenJPEG library
 - 1 encode pictures with jp2, with different compression rates
 - 2 determine quality of the pictures
 - 3 compare with pictures compressed with video codecs

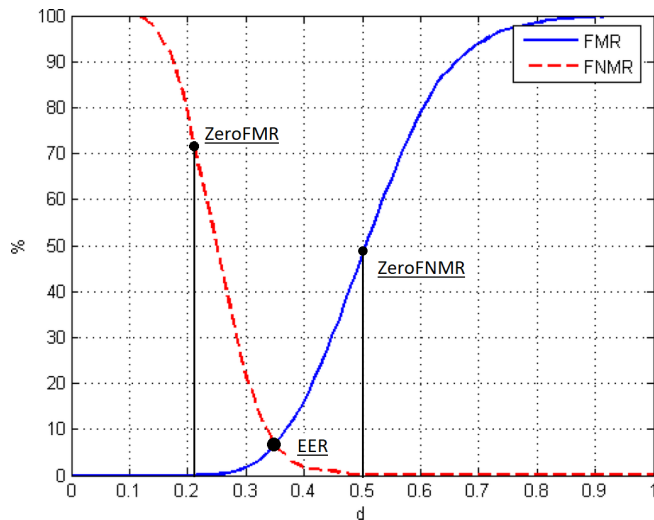
Matcher

- Used as a "black box"
- Compares original and compressed images
→ Checks if matches found
- calculates different error metrics
- Target:
Compare results of jpeg2000 and video compression

Error Metrics

- FMR: False Match Rate
- FNMR: False Non Match Rate
- EER: Equal Error Rate
- ZeroFNMR: FMR at $\text{FNMR} = 0$
- ZeroFMR: FNMR at $\text{FMR} = 0$
- Lower values are always better

Error Metrics



http://www.mdpi.com/sensors/sensors-11-09499/article_deploy/html/images/sensors-11-09499f22-1024.png

CRF

- CRF value (Constant Rate Factor)
 - 1 The range of the quantizer scale is 0-59
 - 2 A lower value means better quality (0 for best quality, lossless)
 - 3 Default value is 23
 - 4 A higher value means bad quality (59 for worst quality)

Presets

- presets (they provide a certain encoding speed)
 - 1 ultrafast , superfast , veryfast , faster , fast
 - 2 medium (default)
 - 3 slow, slower, veryslow, placebo
 - 4 we focused more to the slower presets (medium-veryslow)

Settings

- what are the settings behind them?

	medium	veryslow
-b-adapt	1	2
-bframes	3	8
-ref	3	16

Settings

Quick explanation of the settings :

- `-b-adapt "Mod"`:
 - algorithm for the adaptive distribution of B-frames
 - values : 0,1,2
- `-bframes "Max"`:
 - Defines how many B-frames can be positioned directly behind each other
 - values are between 0 and 16 (3 is default)
- `-ref "frames"`:
 - amount of valid reference frames

Settings (qscale mpeg4)

- -qscale:v n
- configure and select a video quality level
- value for n : 1-31
- 1: highest quality and largest filesize
- 31: lowest quality for smallest filesize

Video Compression

Setup:

- Used ffmpeg v.3.3.2 (latest version)
- Compressed 240 images
- Different crf values (0-50)
- Different qscale values for mpeg4 part 2
- Varying group of pictures (1, 4, 24)
- Two presets (medium, veryslow)

Video Compression

Repeat for each (crf, gop, preset) - combination

- 1 Compress images into single video
- 2 Get videosize (for compression rate)
- 3 Decompress video → get images
- 4 Put into folder named with settings

Additional steps

- Collect image names → parameters for matcher
- Rename decompressed videos

Matching

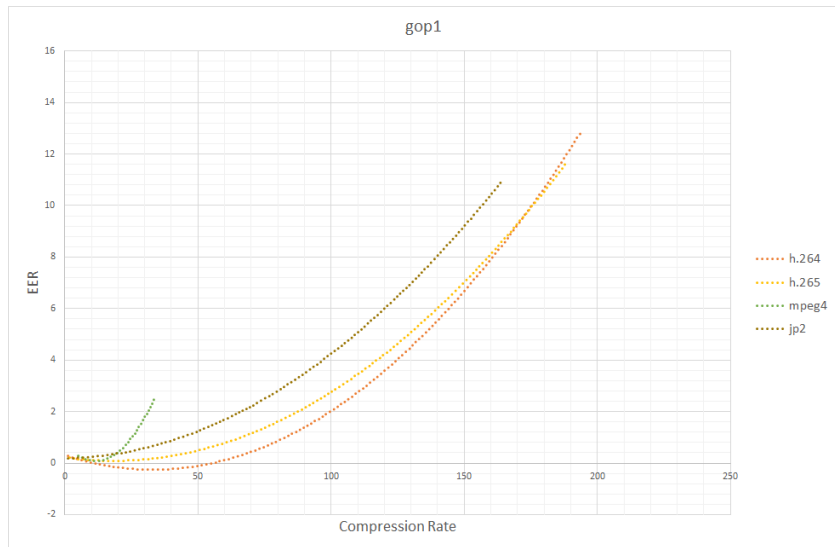
- Used matcher as "black box"
- Input:
 - ① original images
 - ② compression output folders (crf, gop, preset)
- Match each output folder
- Retrieve error metrics
- Evaluate Error rate dependent on compression rate

Evaluation

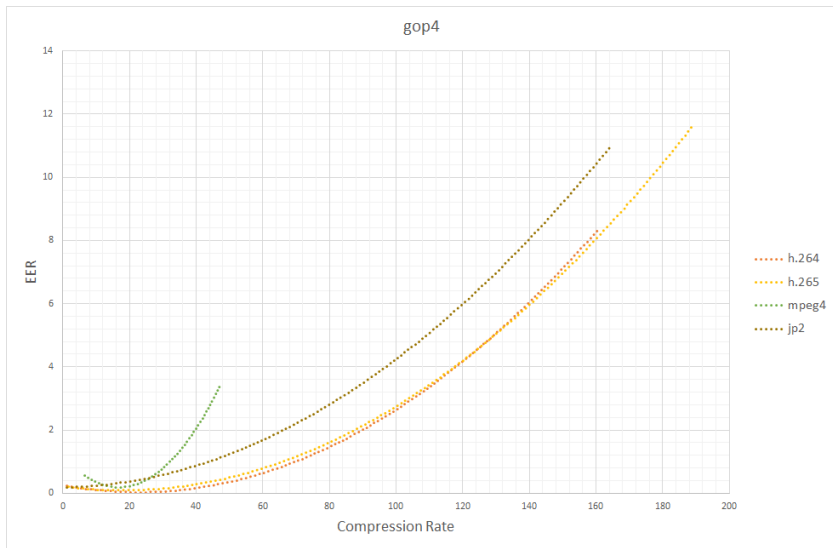
Compression rate vs. matching errors

- Group of pictures (1, 4, 24)
- Codecs (mpeg4 part 2, h.264, h.265)
- presets (medium, veryslow)
- Best result of jpeg2000 compression as baseline

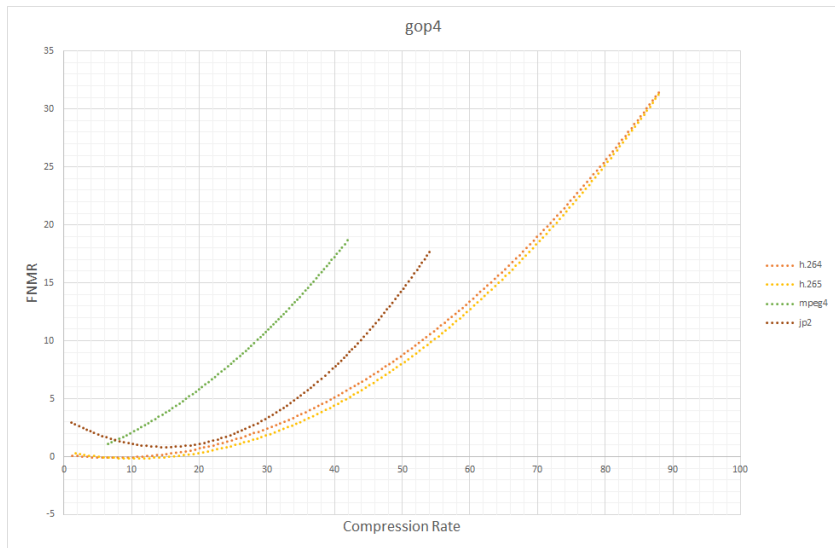
Comparing codecs



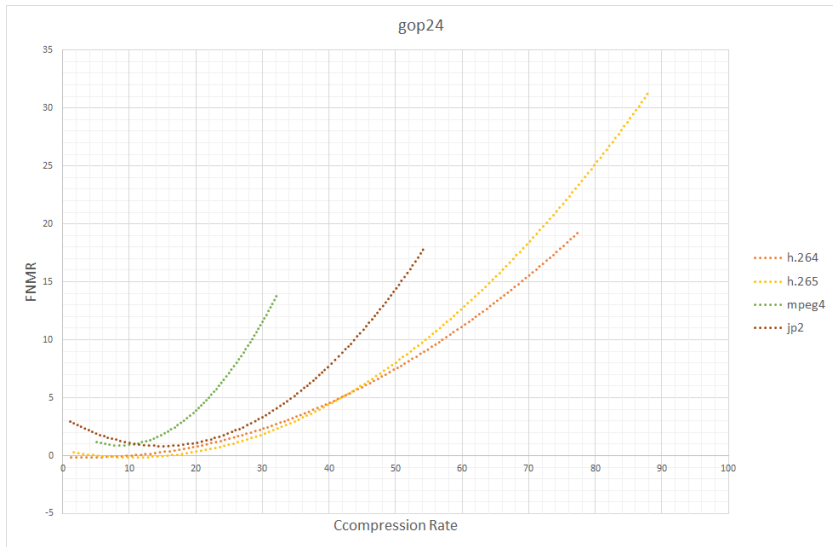
Comparing codecs



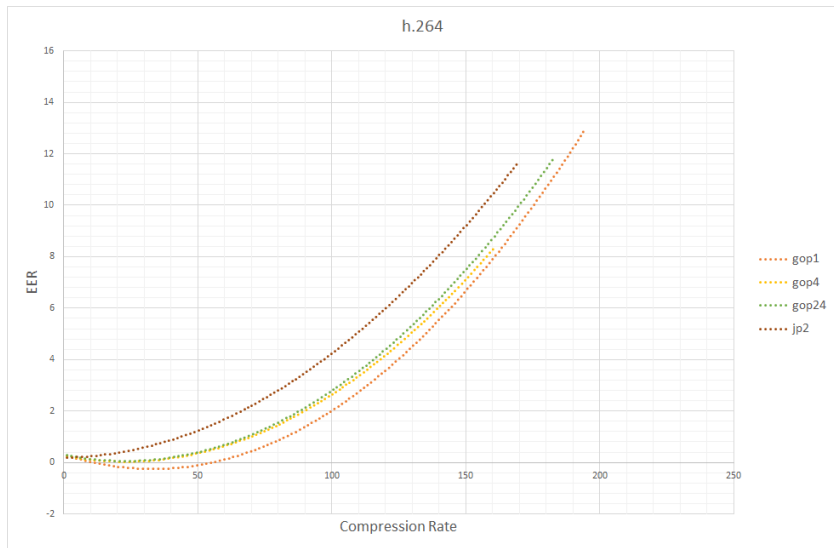
Comparing codecs



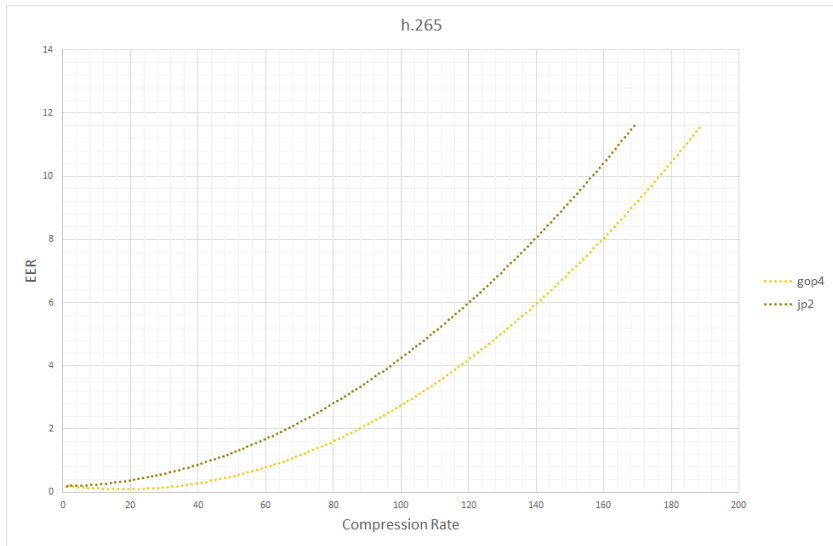
Comparing codecs



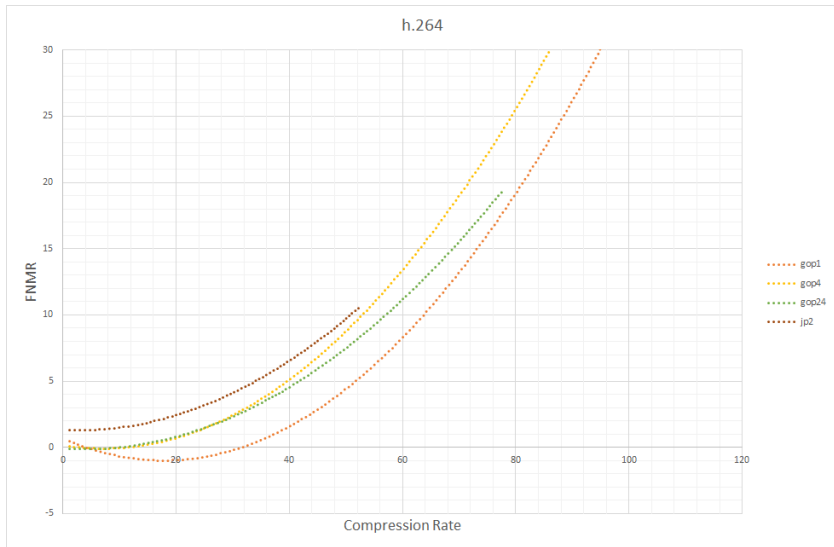
Comparing group of pictures



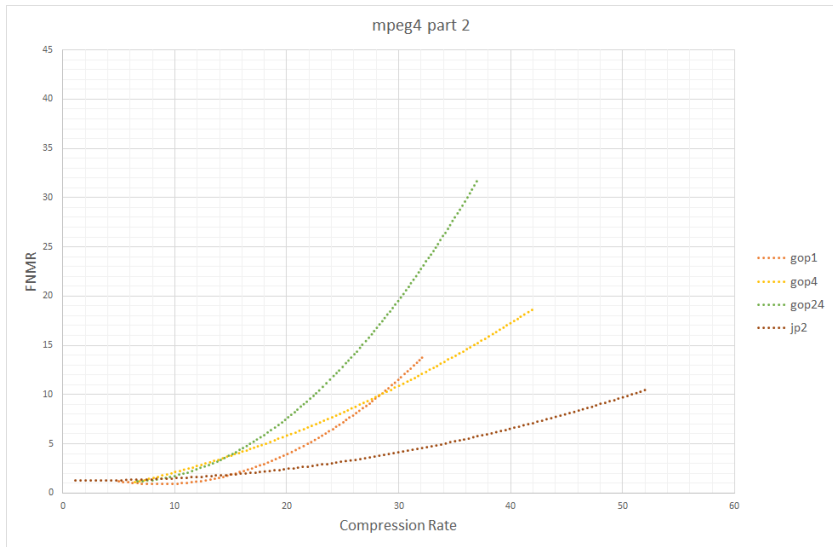
Comparing group of pictures



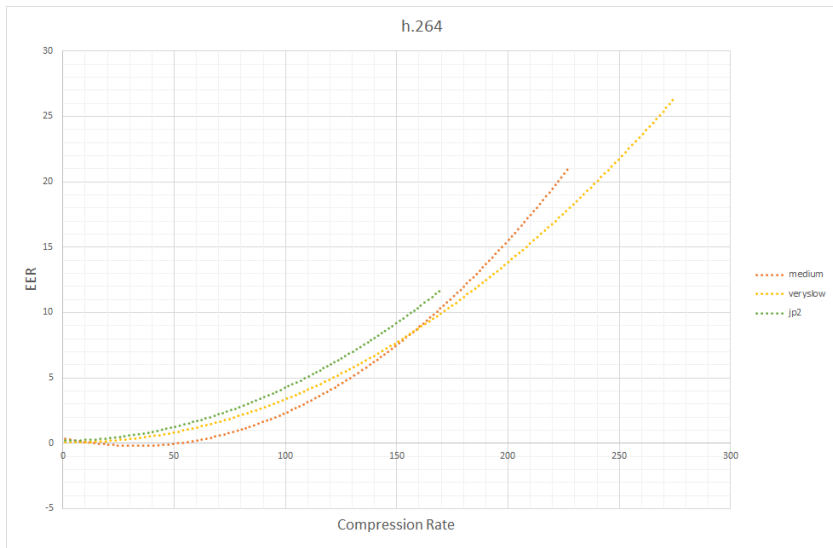
Comparing group of pictures



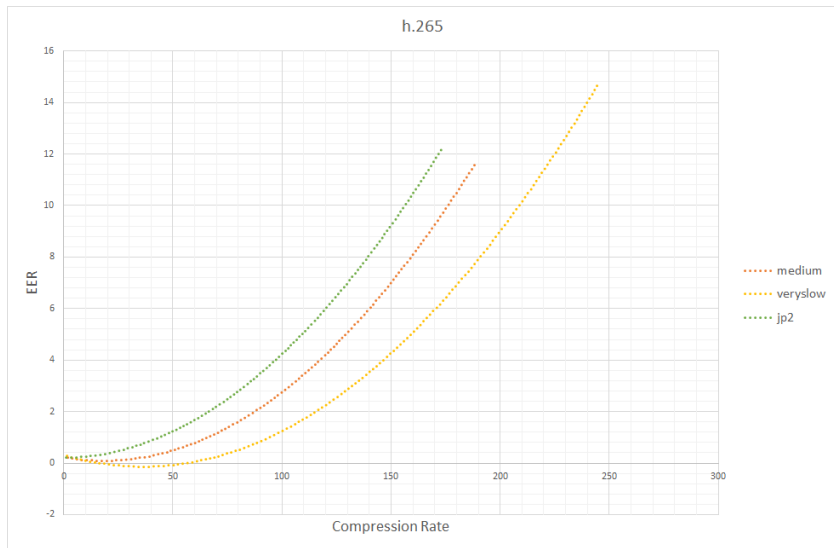
Comparing group of pictures



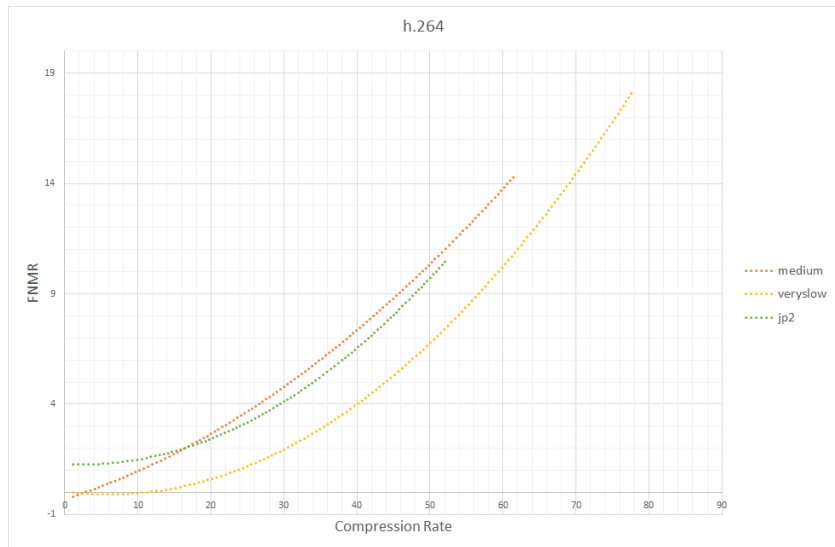
Comparing presets



Comparing presets



Comparing presets



Conclusion

Best Results:

- GOP 1 or GOP 4
- h.264 or h.265
- veryslow

40% higher compression at same EER (h.265 veryslow)

30% higher compression at same FNMR (h.264 veryslow)

Thank you for your attention!