Optimal Hyper-parametrization for efficient Video Compression



Video Compression Hackathon - SBIPowered by Microsoft Corporation Pvt. Ltd.

Problem

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- Storage and retrieval of these video files is a challenge especially given the expected tsunami of video files that are expected to be generated on extending more use-cases to our client base of over 45 crore customers.

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- I made use of the same to learn optimal parameters for a vast array of these.

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- Then, these were made use of for encoding using x265.

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- We also used zlib for further compression.

Video used

• phase3.mp4

• Profile: H.264

• Dimensions : 1920×1080

• FPS: 30.0

• Bit-Rate: 17036 kbps

• Size: 127.8 MB

• The file may be found here.

Decompressed Video

- phase31.mp4
- Profile (Main): H.265
- ullet Dimensions : 1920×1080
- FPS: 30.0 (as expected)
- Bit-Rate: 1381 kbps
- Size (compressed): 9.5 MB

Hardware Details

CPU : Intel(R) Core(TM) i5-1035G1 CPU 1.00GHz

Memory : 8 GB

Memory Clock: 3200 MHz

• L1 Cache: 128 kB

• L2 Cache: 2 MiB

• L3 Cache: 6 MiB

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- Different Types of applications provide scope for different types of compression shorthands.
- Our approach enables us to learn these shorthands

Functional Requirements

- x265
- opencv4
- Python's cryptography
- Python's zlib
- ffmpeg

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All these tools are easily available in Azure Cloud Services and thus, can be used to leverage the application into production.

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- Infrastructure: Azure Cloud Deployment
- Time Required to setup: 2-3 Weeks (Debugging included!)

Non-functional Requirements

- The Video isn't recorded in camcorders which make use of Compressed Sensing.
- The Noise is minimal and patterns in the images remain consistent.
- The Machine on which the Application is used has \geq 8GB RAM.
- This would enable greater compression due to easier Identification of Generic features.

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- Use-case specific optimization in HEVC is generically obtained by making use of hyperparameter optimization.
- Feedback functions could also be tinkered to better reflect our Requirements (should they ever change!)

Github Repository

My Actual Github ID is thevaliantthird

I've kept the Video Hackathon submission Repository via another profile, SBI-Video-Hackathon

Video Demonstration

I have demonstrated usage of my application here.

Thank You!

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Github ID (original) : thevaliantthird

Website: thevaliantthird