

ENCODING COMPLEXITY ANALYSIS THROUGH THE MEASUREMENT OF THE SEARCH SPACE

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Introduction

- Based on the paper:
 - **VVC Search Space Analysis including an Open, Optimized Implementation**
 - In IEEE Transactions on Consumer Electronics, vol. 68, no. 2, pp. 127-138, May 2022.

- Comparisons based on 3 encoders
 - HM 16.22
 - VTM 11.0
 - VVenC 1.0.0

Agenda

- What is complexity in video encoding?
 - Encoding vs decoding
- Empirical search space quantification
 - Measurement
 - Application to partitioning
 - *Application to mode search*
- Conclusion

Video coding complexity

- Video coding is complex
 - In literature: VVC encoding 10x more complex than HEVC, decoding 2x
- Coding complexity is measurable
 - Runtime
 - Energy consumption (electric bill, device heat)
- What causes the complexity and how can it be controlled?
- Does the complexity depend on the standard used?

Encoding vs decoding complexity

- What is **video decoding**?
 - Fixed sequence of steps described in the standard specification
 - The **worst case** can be quantified, using e.g. number of Multiply-Add-Operations

- What is **video encoding**?
 - Video encoders incorporate the decoder
 - Beyond that its basically a **search problem**
 - Decoder is part of the **cost function**

Video encoding as a search problem

- Usual formulation: find a sequence of **N bits**, such that (2^N possibilities)
 - The sequence is **compliant** with a given standard
 - **Minimizing** the distortion between the **decoded video** and the original
- Components influencing enc. complexity: **bitrate**, **complexity of decode**, **accuracy**
- Contradiction
 - Assumption: VVC provides **50%** bitrate savings vs HEVC
 - N half as large for VVC as for HEVC, at double decode complexity
 - **VVC encoding** should actually be **less complex**?
- Smart search algorithms only evaluate fraction of overall search space
 - How big is the **actual visited search space**?
 - Why is this search space larger for each next generation codec?

Empirical search space quantification

- Recap
 - Decoder is part of the encoder cost function
- Idea
 - Quantify how many times a sample is decoded during encoding
- Problem
 - When is a sample decoded during encoding?
- Solution
 - During video decoding each sample is
 - Only contained in a single block
 - Dequantized maximally once
 - Measure the plurality of partitioning and quantization test per sample

Empirical search space quantification

Partitioning search

- Encoder search visits a set of $i_p \in 1 \dots N_p$ blocks during encoding of a frame
 - A frame has N_s samples
 - Each block has a height of $W_p(i)$ and $H_p(i)$
- Partitioning overhead is thus:

$$S_p = \sum_{i=1 \dots N_p} W_p(i) \cdot H_p(i) / N_s$$

- Cumulative area of blocks visited during encoding normalized to frame size

Empirical search space quantification

Coding mode search

- In each visited block, the encoder tests a plurality of coding modes
 - Often most complex step is the rate-distortion-optimized quantization (RDOQ)
 - The RDOQ is applied to $i_Q = 1 \dots N_Q$ blocks of size $W_Q(i) \cdot H_Q(i)$
- Quantization overhead is thus:

$$S_Q = \sum_{i=1 \dots N_Q} W_Q(i) \cdot H_Q(i) / (N_S \cdot S_P)$$

- Cumulative quantized area, normalized to the partitioning area
- Overall encoding overhead (i.e. **empirical search space**):

$$S = S_Q \cdot S_P$$

Partitioning search space

Exemplary for HEVC

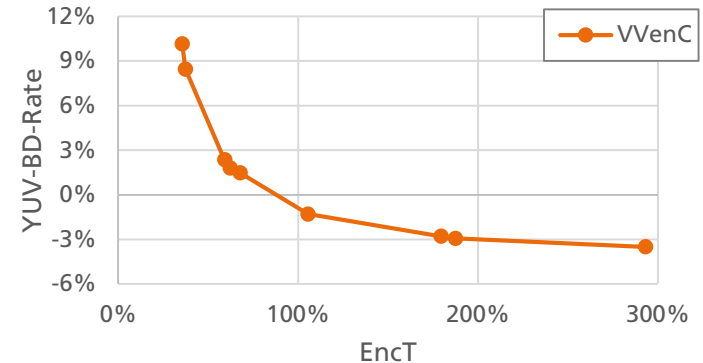
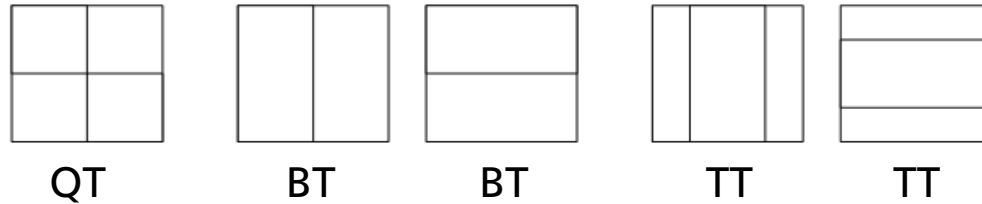
- Assuming the G-BFOS search algorithm
 - *Forward-only* search
 - Each sample once in each config.
- For HEVC assuming CTU of 64x64
 - 4 different CU sizes
 - PU splits also count as partitioning
 - Easy to enumerate
- Simple formulation of upper bound

| Depth / CU size | | Allow PU splits |
|--------------------------|---|--|
| 0 / 64x64 | | NxN, N/2xN/2, NxN/2, N/2xN, 4x asymmetric modes |
| 1 / 32x32 | | |
| 2 / 16x16 | | |
| 3 / 8x 8 | | NxN, N/2xN/2, NxN/2, N/2xN |
| Partitionings per sample | | |
| Intra | 4 | 5 |
| Inter | 4 | 28 |

Partitioning search space

Recap partitioning in VVC

- Five recursive splits
 - QT only with a QT
- Exponential growth
- Empirical upper bound
 - Assuming full traversal with G-BFOS, including
 - Chroma separate tree
 - Local dual tree, mode restrictions
 - Depends on high-level partitioning parameters
- Figure
 - CTU128, QT: 128x128 to 8x8
 - BT and TT splits: 0 to 4 recursion levels



Partitioning search space

Upper bounds for VVC search space

- QT depth is simply the difference between min and max block size
 - BTT depth has to be kept much lower to limit complexity
- VVC partitioning space can be much larger than HEVC
 - For practical encoders it can be very well kept lower

| | Intra Frames | | | | | | P/B Frames | | | | | |
|-----------------|--------------|----------|-----|-----------|-----|--------------|------------|----------|-----|-----------|---|---------------|
| | CTU | Max size | | Max depth | | bound | CTU | Max size | | Max depth | | bound |
| | | QT | BTT | QT | BTT | | | QT | BTT | | | |
| faster | 64 | 64 | N/A | 4 | 0 | 4.67 | 64 | 64 | N/A | 4 | 0 | 5.00 |
| fast | 64 | 64 | 32 | 4 | 1 | 14.33 | 64 | 64 | N/A | 4 | 0 | 5.00 |
| medium | 128 | 128 | 32 | 4 | 2 | 37.50 | 128 | 128 | 128 | 4 | 1 | 24.00 |
| slow | 128 | 128 | 32 | 4 | 3 | 85.42 | 128 | 128 | 128 | 4 | 2 | 75.00 |
| slower | 128 | 128 | 32 | 4 | 3 | 85.42 | 128 | 128 | 128 | 4 | 3 | 220.75 |
| VTM-11.0 | 128 | 128 | 32 | 4 | 3 | 85.42 | 128 | 128 | 128 | 4 | 3 | 220.75 |

Partitioning search space

Empirical results for Inter frames

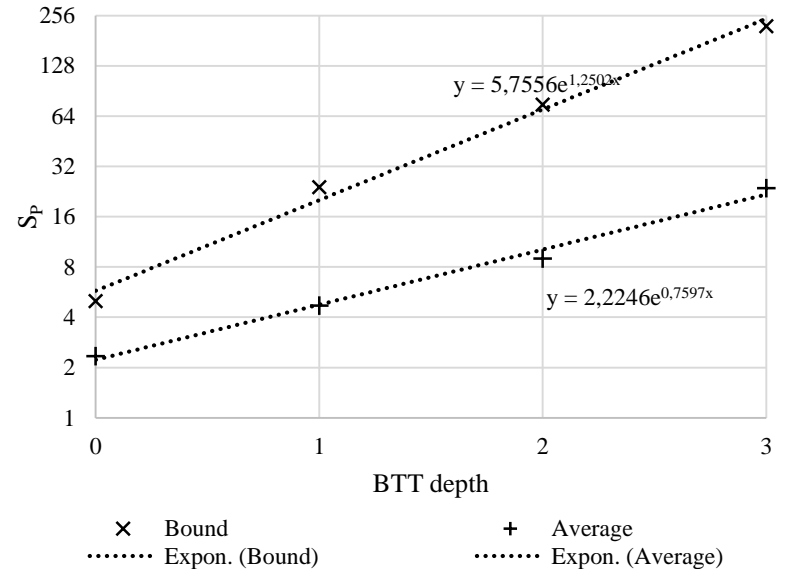
- Measured on JVET CTC Classes A1, A2, and B, in CTC encoding conditions
- Observations for HM
 - HM has no CU depth speedup
 - ... but has PU split speedups
- VVenC slower visits less partitions than VTM, bcs of more aggressive optimization
- The larger the search space, the easier it is to limit
- Search space varies, e.g. with target quality (more so for VVencC)

| | Partitioning S_p | | | |
|--------------|--------------------|-------|-------|--------|
| | avg | min | max | bound |
| VVenC faster | 2.39 | 1.71 | 3.31 | 5.00 |
| VVenC fast | 2.34 | 1.68 | 3.23 | 5.00 |
| VVenC medium | 4.69 | 3.32 | 6.75 | 24.00 |
| VVenC slow | 8.98 | 6.01 | 14.12 | 75.00 |
| VVenC slower | 23.71 | 13.49 | 42.04 | 220.75 |
| VTM | 29.18 | 17.54 | 49.60 | 220.75 |
| HM CU | 3.95 | 3.95 | 3.95 | 4.00 |
| HM CU+PU | 15.51 | 13.97 | 17.77 | 28.00 |

Partitioning search space

Exponential growth, theoretical and measured for P/B frames

- Fast algorithm can limit the growth
 - Both base and exponent reduced
 - Exponent bound: $\sim 1.25 \times \text{Depth}$
 - Exponent average: $\sim 0.76 \times \text{Depth}$
- The curves diverge
 - The larger the search space, the more opportunities for limitation



Overall search space

First observations

- Why is partitioning such a popular topic for optimization in literature?
 - $S_p \gg S_Q$ for both VTM and HM
 - In VVenC, which was Pareto-Optimized, $S_p \sim S_Q$ (except for *slower*)
- It is hard to define an upper bound for S_Q

| | Partitioning S_p | | | | Quantization S_Q | | |
|----------------|--------------------|-------|-------|--------|--------------------|-------|-------|
| | avg | min | max | bound | avg | min | max |
| VVenC faster | 2.39 | 1.71 | 3.31 | 5.00 | 3.15 | 2.92 | 3.55 |
| VVenC fast | 2.34 | 1.68 | 3.23 | 5.00 | 4.46 | 4.13 | 5.05 |
| VVenC medium | 4.69 | 3.32 | 6.75 | 24.00 | 5.81 | 5.29 | 6.59 |
| VVenC slow | 8.98 | 6.01 | 14.12 | 75.00 | 10.08 | 9.37 | 10.99 |
| VVenC slower | 23.71 | 13.49 | 42.04 | 220.75 | 13.08 | 11.85 | 14.84 |
| VTM-11.0 | 29.18 | 17.54 | 49.60 | 220.75 | 14.28 | 12.35 | 16.91 |
| HM-16.22 CU | 3.95 | 3.95 | 3.95 | 4.00 | 20.93 | 17.18 | 27.27 |
| HM-16.22 CU+PU | 15.51 | 13.97 | 17.77 | 28.00 | 5.32 | 4.84 | 6.05 |

Overall search space

Further observations

- VVenC *medium* has a search space similar to HM, but provides ~30% BD-rate gain
- Average S_Q ~3x larger in VTM than HM, S_P ~2x larger in VTM than HM
 - Partitioning adds complexity to VVC, but new encodings modes add even more
 - E.g. a lot of new merge modes

| | Partitioning S_P | | | | Quantization S_Q | | |
|----------------|--------------------|-------|-------|--------|--------------------|-------|-------|
| | avg | min | max | bound | avg | min | max |
| VVenC faster | 2.39 | 1.71 | 3.31 | 5.00 | 3.15 | 2.92 | 3.55 |
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Conclusions

Empirical search space measurement

- Recap of the presentation
 - Empirical measurement of search space was presented
 - Data was presented and discussed, with focus on partitioning
- Conclusions
 - Partitioning is complex, but is already very effectively reduced
 - Is partitioning in VTM over-dimensioned?
- Shortcomings
 - Only measures the CU-loop search space
- Outlook
 - Measure distortion calculations per sample (i.e. prediction overhead)
 - Evaluate the ratio of prediction overhead to quantization overhead

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Questions?

