

PPoPP 2020

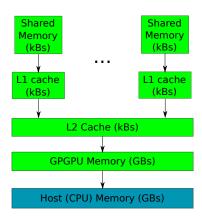


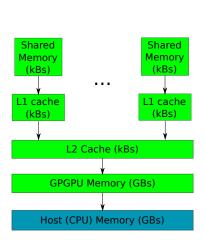
POSTER: Optimizing GPU Programs By Partial Evaluation

Aleksey Tyurin, Daniil Berezun, Semyon Grigorev

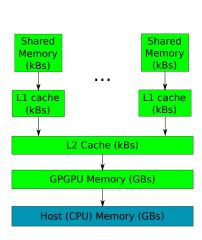
JetBrains Research, Programming Languages and Tools Lab Saint Petersburg University

February 24, 2020

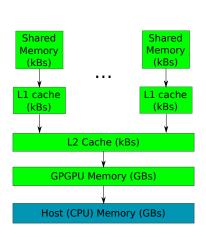




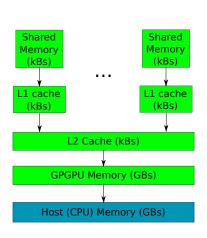
- Global memory
 - © Big
 - Slow



- Global memory
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- Shared memory
 - © Fast
 - Relatively small
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- Memory traffic is a bottleneck

- Substring matching
- 2D convolution
- Filtering by using Hidden Markov Models (HMM)

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- Substring matching ⇒ Data curving (cyber forensics)
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filterParams is static during one data processing session.

How can we use this fact to optimize our procedure?

```
handleData (filterParams, data)
{
  res = new List()
  for d in data
     for e in filterParams
        if d % e == 0
        then res.Add(d)
  return res
}
```

```
partial evaluator
\llbracket handleData 
rbracket [filterParams, data] = \llbracket \llbracket mix 
rbracket [filterParams] 
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                                                     if d % 2 == 0 ||
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Evaluation Setup

- We use AnyDSL framework for specialization
 - Special DSL which can be specialized and compiled
 - Ahead-of-time specialization

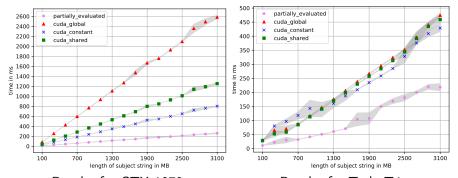
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- Algorithms
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Evaluation Setup

- We use AnyDSL framework for specialization
 - ▶ Special DSL which can be specialized and compiled
 - ► Ahead-of-time specialization
- Algorithms
 - Naïve multiple substring matching
 - 2D convolution
- Environment
 - ▶ GTX-1070: Pascal architecture, 8GB GDDR5, 1920 CUDA cores
 - ▶ Tesla T4: Turing architecture, 16 GB GDDR6, 2560 CUDA cores

- Application: data curving
- Subject string: byte sequence from real hard drive
- Patterns: 16 file signatures from GCK's file signatures table¹

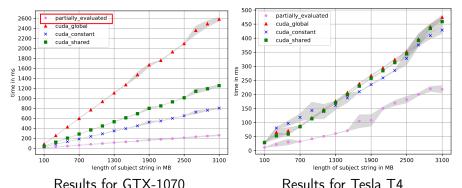


Results for GTX-1070

Results for Tesla T4

¹https://www.garykessler.net/library/file_sigs.html

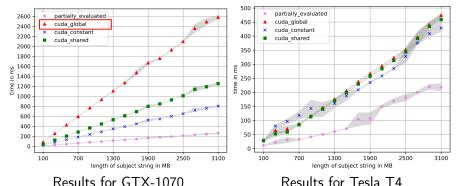
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Results for Tesla T4

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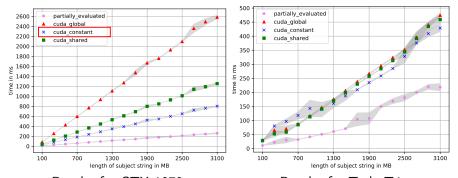
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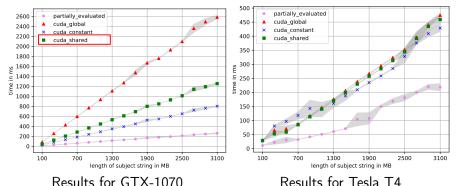


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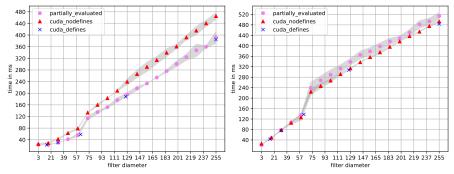
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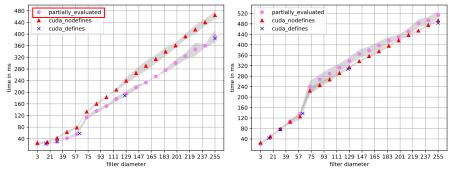
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- Subject image: random image of size 1Gb
- Filters: random square filters with diameter 3 to 255



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Results for Tesla T4

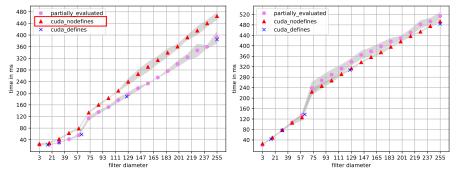
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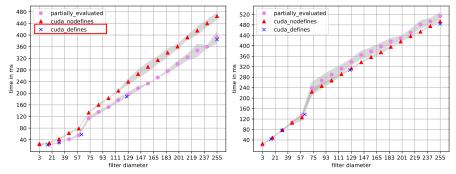
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Conclusion

- Partial evaluation improves performance of GPGPU procedures
 - Effect depends on GPU architecture
 - ★ Dependencies should be carefully analyzed
 - Effect depends on the initial memory access pattern
 - ★ Irregular pattern better performance improvement

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- Evaluation on real-world examples
 - Homology search in bioinformatics
 - Regular expression matching for traffic analysis, log processing
 - Graph database querying
 - Ray tracing, path tracing

Contact Information

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Thanks!