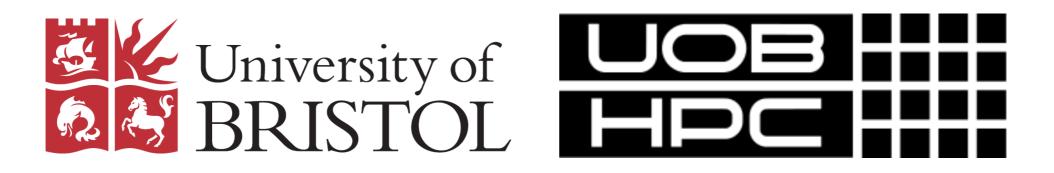
#### Analyzing and improving performance portability of OpenCL applications via auto-tuning

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### Overview

- Performance portability is one of the key concerns for developers targeting many different architectures
- Current work in this area has provided mixed results
- Here we propose a method of rigorously analyzing performance portability by exploiting the black-box nature of auto-tuning
- We also present a simple technique that can improve performance portability when using auto-tuning

# Analysis approach

- Expose set of possible implementation decisions as tuning options
- Dynamically generate kernels to provide greater flexibility
- Run auto-tuner to optimize the kernel for each device individually
- This produces a set of architecture-specific kernels
- We then run each architecture-specific kernel on every other device and measure efficiency

#### Benchmark #1: Jacobi method

- Work-group size / parallel decomposition
- Memory layout / memory access pattern
- Loop unrolling
- \*+ vs mad vs fma
- Branching vs masks
- Division by diagonal (inline or precompute)
- Address spaces of input vectors
- Embedding values into kernel as constants

Solve  $A\mathbf{x} = \mathbf{b}$ Split matrix  $\mathbf{A}$  into diagonal  $\mathbf{D}$ and remainder  $\mathbf{R}$  $\mathbf{x}_{i+1} = \mathbf{D}^{-1}(\mathbf{b} - \mathbf{R}\mathbf{x}_i)$ 

#### Benchmark #2: Bilateral filter

- Work-group size
- Tile size / tile layout
- Prefetch pixels into private/local memory
- Buffers vs images
- \*+ vs mad vs fma
- Loop interchange
- Native math functions

$$W(x, y, i, j) = e^{\left(-\frac{\sqrt{i^2 + j^2}}{2\sigma_d} - \frac{||(I(x, y) - I(x+i, y+j))||}{2\sigma_r}\right)}$$

 $O(x,y) = \frac{i = -r \, j = -r}{r \, r}$ 

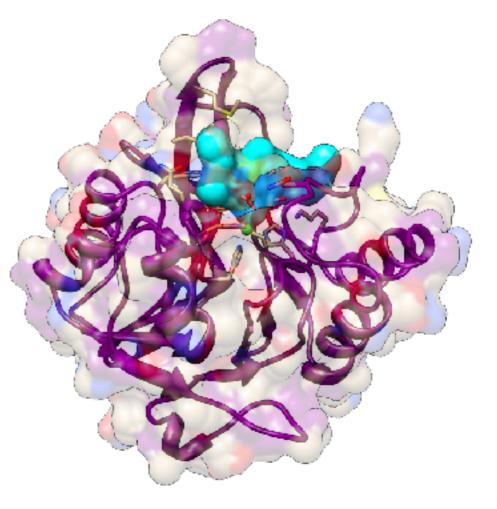
 $\sum \sum I(i,j) \cdot W(x,y,i,j)$ 

 $\sum \sum W(x,y,i,j)$ 

Embedding values into kernel as constants

#### **Benchmark #3: BUDE**

- Parallel decomposition / work-group size
- Data layout / memory access patterns
- Loop interchange
- Address space of molecules / forcefield
- Precomputing forcefield coefficients
- Native math functions
- Embedding values into kernel as constants



### Devices

Vendor	Product	Architecture	Compute units
	GeForce GTX 580	Fermi (GF110)	16
	GeForce GTX 680	Kepler (GK104)	8
NVIDIA	GeForce GTX 780 Ti	Kepler (GK110)	15
	GeForce GTX 980 Ti	Maxwell (GM200)	22
	GeForce GTX 1080 Ti	Pascal (GP102)	28
	Radeon HD 7970	Tahiti	32
AMD	Radeon R9 290X	Hawaii	44
	Radeon R9 Furyx	Fiji	64
	Radeon RX 480	Ellesmere	36
	Core i5-3550	Ivy Bridge	4
Intel	Core i5-4590	Haswell	4
	Core i5-6600	Skylake	4

#### Jacobi efficiencies

	Skylake CPU -	90%	69%	93%	90%	90%	45%	28%	14%	21%	2%	99%	100%
	Haswell CPU -	12%	16%	16%	39%	43%	37%	20%	11%	17%	2%	100%	97%
	Ivy Bridge CPU -	21%	14%	18%	41%	40%	19%	14%	6%	13%	100%	53%	44%
	RX 480 -	87%	74%	78%	95%	93%	99%	99%	97%	100%	2%	93%	81%
T	R9 Fury X -	24%	20%	20%	23%	37%	100%	99%	100%	99%	62%	80%	65%
	R9 290X -	91%	79%	84%	97%	97%	99%	100%	93%	100%	2%	70%	57%
	HD 7970 -	23%	19%	20%	21%	33%	100%	100%	99%	100%	73%	85%	71%
5	GTX 1080 Ti -	91%	97%	58%	98%	100%	68%	38%	28%	62%	1%	5%	4%
	GTX 980 Ti -	71%	69%	51%	100%	92%	53%	33%	49%	36%	5%	5%	5%
	GTX 780 Ti -	93%	66%	100%	95%	90%	68%	33%	32%	62%	5%	5%	4%
	GTX 680 -	92%	100%	71%	98%	98%	55%	38%	26%	57%	1%	5%	4%
	GTX 580 -	100%	87%	74%	53%	56%	Х	X	X	X	1%	5%	4%
		GTX 580 -	GTX 680 -	GTX 780 Ti -	GTY 980 Ti -	GTX 1080 Ti -	- 0262 CH	R9 2903 -	R9 Fury -	RY 480 -	IVY Bridge CPU -	Haswell CPU -	Skylake CpU -

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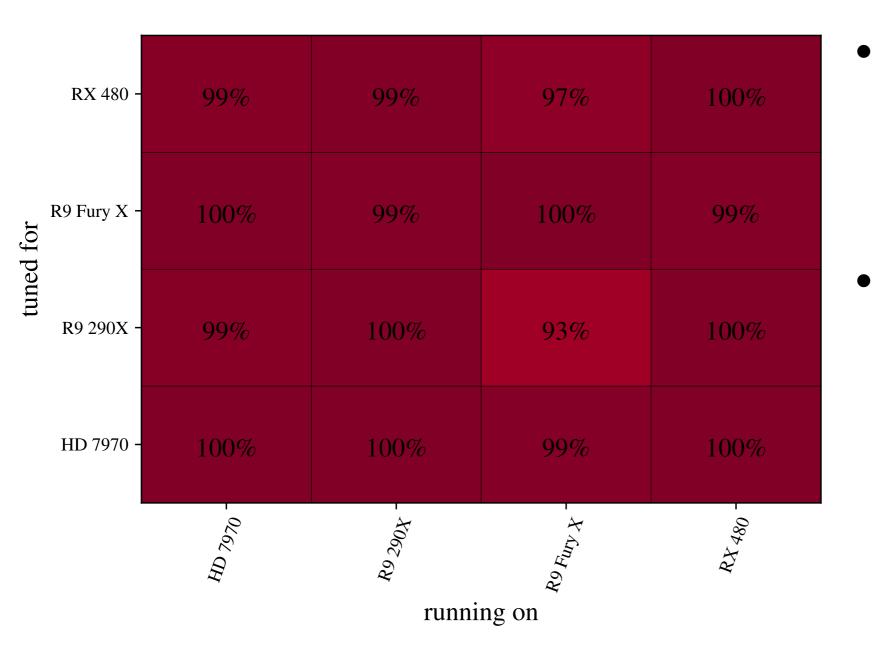
### Jacobi - NVIDIA

GTX 1080 Ti -	91%	97%	58%	98%	100%
GTX 980 Ti -	71%	69%	51%	100%	92%
GTX 780 Ti -	93%	66%	100%	95%	90%
GTX 680 -	92%	100%	71%	98%	98%
GTX 580 -	100%	87%	74%	53%	56%
-	GTX 580 -	GTX 680 -	GTX 780 Ti	67X 980 Ti	GTX 1080 Ti
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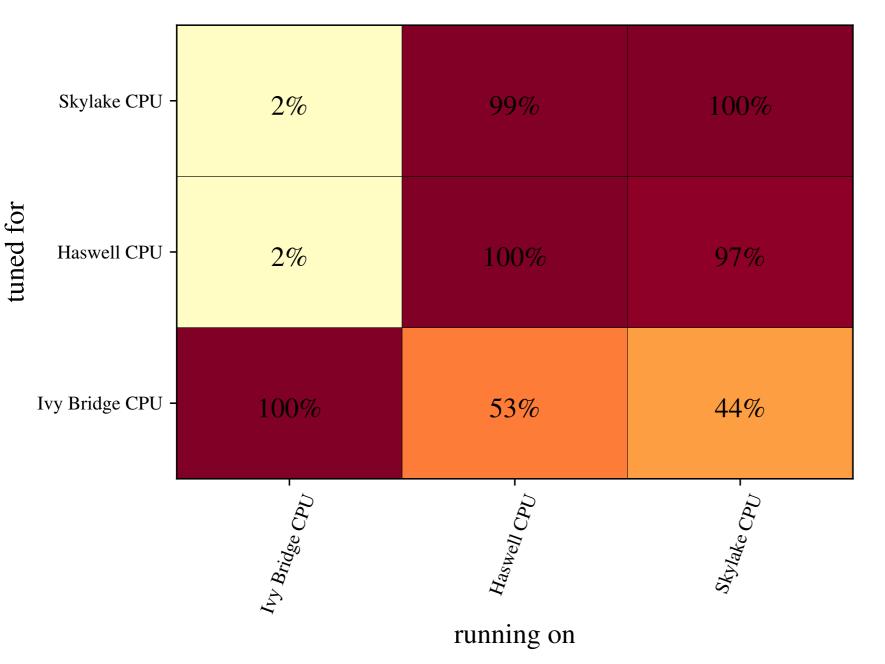
- Work-group size differs between devices
- Other drops due to address spaces and memory access pattern

### Jacobi - AMD



- Most parameters uniform between all AMD devices
- FuryX suffers a little with expensive division

### Jacobi - Intel



- Ivy Bridge performs poorly with **fma** builtin
- Vectorisation width differs

### Jacobi efficiencies

													<u>WCE</u>
Skylake CPU -	90%	69%	93%	90%	90%	45%	28%	14%	21%	2%	99%	100%	2%
Haswell CPU -	12%	16%	16%	39%	43%	37%	20%	11%	17%	2%	100%	97%	2%
Ivy Bridge CPU -	21%	14%	18%	41%	40%	19%	14%	6%	13%	100%	53%	44%	6%
RX 480 -	87%	74%	78%	95%	93%	99%	99%	97%	100%	2%	93%	81%	2%
R9 Fury X -	24%	20%	20%	23%	37%	100%	99%	100%	99%	62%	80%	65%	20%
R9 290X -	91%	79%	84%	97%	97%	99%	100%	93%	100%	2%	70%	57%	2%
HD 7970 -	23%	19%	20%	21%	33%	100%	100%	99%	100%	73%	85%	71%	19%
GTX 1080 Ti -	91%	97%	58%	98%	100%	68%	38%	28%	62%	1%	5%	4%	1%
GTX 980 Ti -	71%	69%	51%	100%	92%	53%	33%	49%	36%	5%	5%	5%	5%
GTX 780 Ti -	93%	66%	100%	95%	90%	68%	33%	32%	62%	5%	5%	4%	4%
GTX 680 -	92%	100%	71%	98%	98%	55%	38%	26%	57%	1%	5%	4%	1%
GTX 580 -	100%	87%	74%	53%	56%	X	X	X	X	1%	5%	4%	-
	580 -	680 -		$0 I_i$		- 026	- 406	A A	RY 480 -			Skylake CPU	
	GTX 580	GTX 680	GTX 780 Ti	GTX 980 Ti	GTX 1080 Tr	0262 (JH	R9 2908	R9 <sub>FuryX</sub>	RY	dge (	) llə4	<sup>lake</sup> (	
	-	-	61	61	ELS B			X		Ivy Bridge CpU	Haswell CpU	Skyn	
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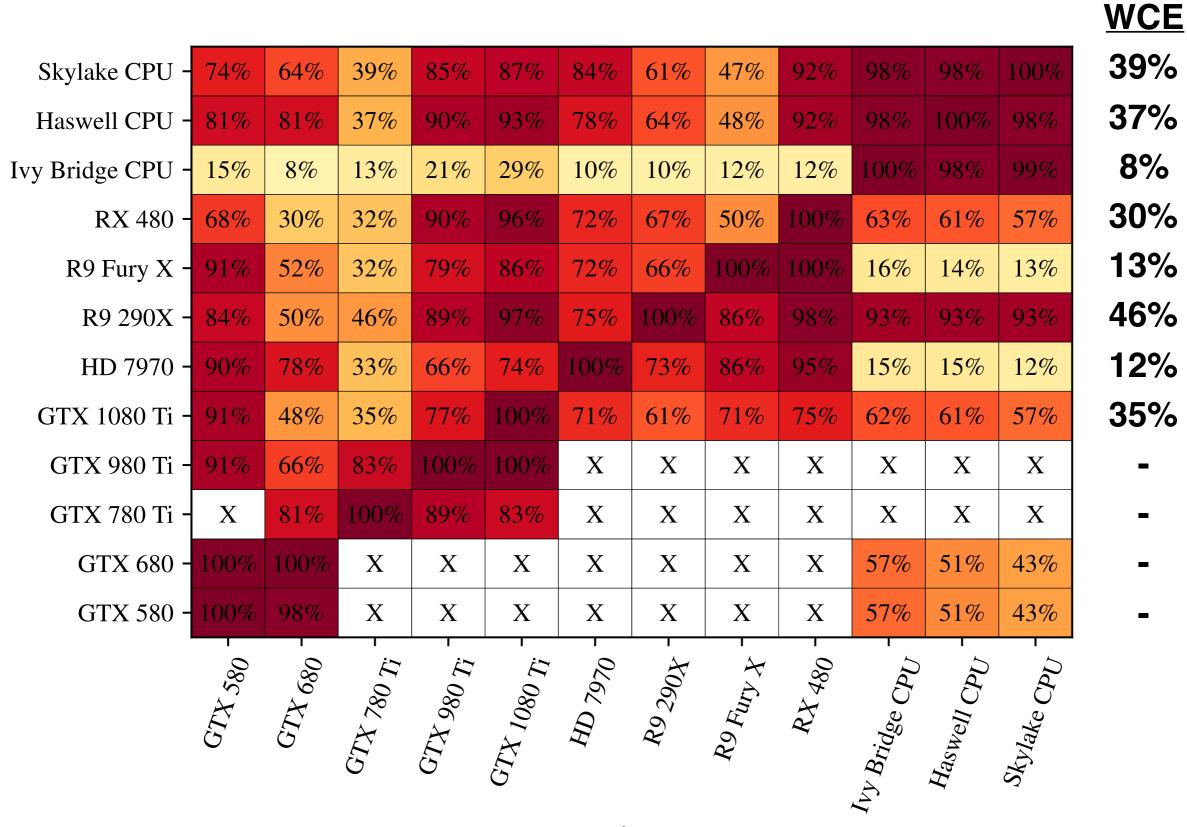
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### **Bilateral efficiencies**

19%	13%	18%	17%	12%	11%	8%	9%	98%	100%	100%
19%	13%	18%	17%	12%	11%	8%	9%	98%	100%	100%
8%	6%	7%	7%	7%	6%	5%	5%	100%	99%	99%
70%	70%	69%	84%	87%	100%	100%	100%	29%	27%	26%
70%	71%	76%	85%	87%	100%	100%	100%	28%	27%	26%
70%	68%	73%	85%	86%	100%	100%	100%	28%	27%	25%
93%	90%	97%	97%	100%	Х	X	Х	14%	72%	73%
100%	97%	99%	100%	98%	35%	31%	37%	14%	72%	71%
95%	96%	100%	98%	98%	37%	37%	45%	74%	71%	73%
96%	100%	99%	98%	97%	37%	38%	45%	74%	72%	74%
100%	98%	99%	100%	99%	35%	31%	29%	14%	73%	74%
GTX 580 -	GTY 680 -	GTX 780 Ti -	¥980 Ti -	GTX 1080 Ti -	R9 2907 -	$R_{9}F_{ury}_{X}$	RF 480 -	$d_{\mathcal{S}}e Cp_U$	Haswell CPU -	Skylake CPU -
	<ul> <li>19%</li> <li>8%</li> <li>70%</li> <li>70%</li> <li>70%</li> <li>93%</li> <li>100%</li> <li>96%</li> <li>100%</li> </ul>	119%       13%         8%       6%         70%       70%         70%       71%         70%       68%         93%       90%         100%       97%         95%       96%         100%       98%	19%       13%       18%         8%       6%       7%         70%       70%       69%         70%       71%       76%         70%       68%       73%         93%       90%       97%         93%       90%       99%         100%       96%       100%         96%       100%       99%         100%       98%       99%	19%13%18%17%8%6%7%7%70%70%69%84%70%71%76%85%70%68%73%85%93%90%97%97%100%97%99%100%96%100%99%98%100%98%99%100%	19%13%18%17%12%8%6%7%7%7%70%70%69%84%87%70%71%76%85%87%70%68%73%85%86%93%90%97%97%100%95%96%100%98%98%100%98%99%100%99%	19%13%18%17%12%11%8%6%7%7%6%70%70%69%84%87%100%70%71%76%85%87%100%70%68%73%85%86%100%93%90%97%97%100%X100%97%99%100%98%35%96%100%98%98%37%100%98%99%100%99%35%	19%13%18%17%12%11%8%8%6%7%7%7%6%5%70%70%69%84%87%100%100%70%71%76%85%87%100%100%70%68%73%85%86%100%100%93%90%97%97%100%XX100%97%99%100%98%35%31%96%100%98%97%37%38%100%98%99%100%99%35%31%	19%13%18%17%12%11%8%9%8%6%7%7%7%6%5%5%70%70%69%84%87%100%100%100%70%71%76%85%87%100%100%100%70%68%73%85%86%100%100%100%93%90%97%97%100%XXX100%97%98%35%31%37%95%96%100%98%97%37%38%45%100%98%99%100%99%31%29%	19%13%18%17%12%11%8%9%98%8%6%7%7%7%6%5%5%100%70%70%69%84%87%100%100%100%29%70%71%76%85%87%100%100%100%28%70%68%73%85%86%100%100%100%28%93%90%97%97%100%XXX14%100%97%99%100%98%35%31%37%14%96%100%98%97%37%38%45%74%100%98%99%100%99%35%31%29%14%	19%13%18%17%12%11%8%9%98%100%8%6%7%7%6%5%5%100%99%70%70%69%84%87%100%100%100%29%27%70%71%76%85%87%100%100%100%28%27%70%68%73%85%86%100%100%100%28%27%93%90%97%97%100%XX14%72%93%90%97%98%35%31%37%14%72%95%96%100%98%98%37%38%45%74%72%96%98%99%35%31%29%14%72%96%98%99%35%31%29%14%73%

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### **BUDE efficiencies**

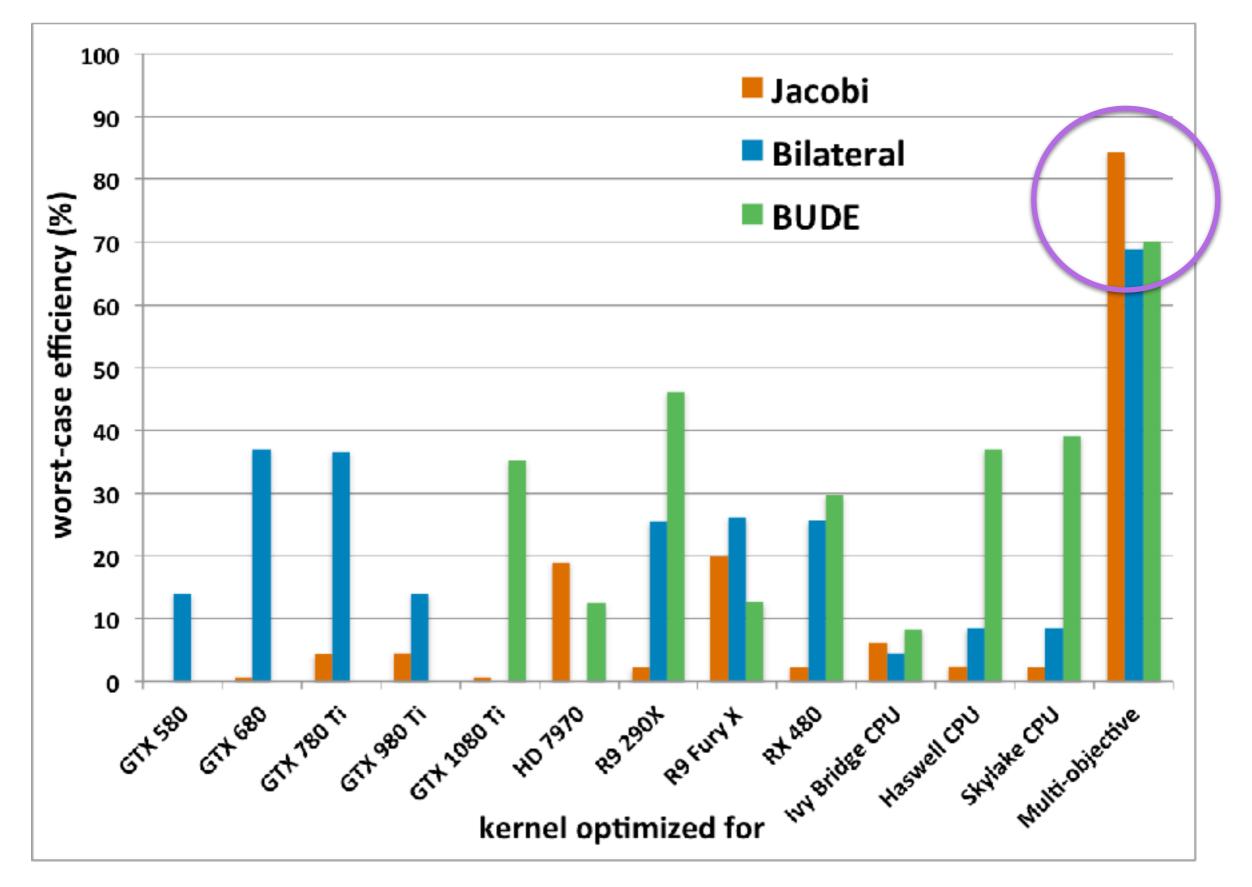


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### Multi-objective auto-tuning

- Extend tuning process to consider multiple devices at once
- Each time a kernel is generated, the auto-tuner evaluates it on every target device
- The performance values are then reduced into a single number representing the overall 'fitness'
- We use worst-case efficiency for this fitness function

#### **Multi-objective tuning results**



## Summary

- Over-optimisation hurts performance portability
- Auto-tuning can be a great way to expose these issues
- It can also help generate performance portable kernels
- Future work looking at tuning across different input/ problem configurations