### **How to Write Fast Numerical Code**

Fall 2018

Lecture: Balance Principles, Part II

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## **References**

- These slides and the work is from Kenneth Czechowksi, Rich Vuduc et al., Georgia Tech
- Kenneth Czechowski, Casey Battaglino, Chris McClanahan, Aparna Chandramowlishwaran, and Richard Vuduc. Balance principles for algorithmarchitecture co-design. In Proc. USENIX Wkshp. Hot Topics in Parallelism (HotPar), May 2011.
- Kenneth Czechowski, Chris McClanahan, Casey Battaglino, Kartik Iyer, P.-K. Yeung, Richard Vuduc. On the communication complexity of 3D FFTs and its implications for exascale. In Proceedings of the ACM International Conference on Supercomputing (ICS), 2012.

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## **Balance Principles II**

Czechowksi et al. 2011

$$T_{\text{mem}} \leq T_{\text{comp}}$$

$$\frac{p\pi}{\beta}\left(1+\frac{\alpha\beta/\lambda}{Q/D}\right) \leq \frac{W}{Q\lambda}\left(1+\frac{p}{W/D}\right)$$

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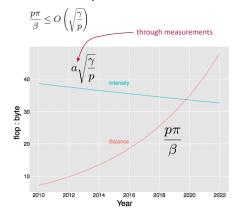
# **Application: Analyze Effect of HW Trends**

Czechowksi et al. 2012

#### 10 year extrapolation (2010 - 2020)

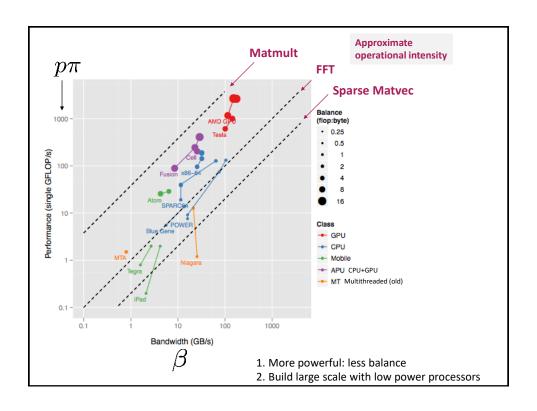
Parameter		2010 values	Doubling time (in years)	10-year increase factor	value
Peak:	$C_{\text{CPU}}$ $C_{\text{GPU}}$	50.4 GF/s 515 GF/s	1.7	59.0×	3.0 TF/s 30 TF/s
Cores:a	$\rho_{\mathrm{CPU}}$ $\rho_{\mathrm{GPU}}$	6 448	1.87	40.7×	134 18k
Memory bandwidth:	$\beta_{\text{CPU}}$ $\beta_{\text{GPU}}$	21.3 GB/s 144 GB/s	3.0	9.7×	206 GB/ 1.4 TB/s
Fast memory	$Z_{\text{CPU}}$ $Z_{\text{GPU}}$	$^{6}$ MB $^{2.7}$ MB $^{b}$	2.0	32.0×	192 MB 86.4 MB
Line size:	$L_{\text{CPU}}$ $L_{\text{GPU}}$	64 B 128 B	10.2	$2.0 \times$	128 B 256 B
Link bandwidth:	$\beta_{\mathrm{link}}$	$10~\mathrm{GB/s}$	2.25	21.8×	218 GB/
Machine peak:	$R_{\text{peak}}$	$4~\mathrm{PF/s}$	1.0	1000×	$4 \; \mathrm{EF/s}$
System memory:	E	635 TB	1.3	208×	132 PB
Nodes $(\frac{R_{\text{peak}}}{C})$ :	$P_{\text{CPU}}$ $P_{\text{GPU}}$	79,400 7,770	2.4	17.4×	1.3M 135,000

#### Matrix-multiplication on GPU



Even Matmult on GPU could become memory bound!

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# **Application: Analyze Effect of HW Trends**

Czechowksi et al. 2012

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