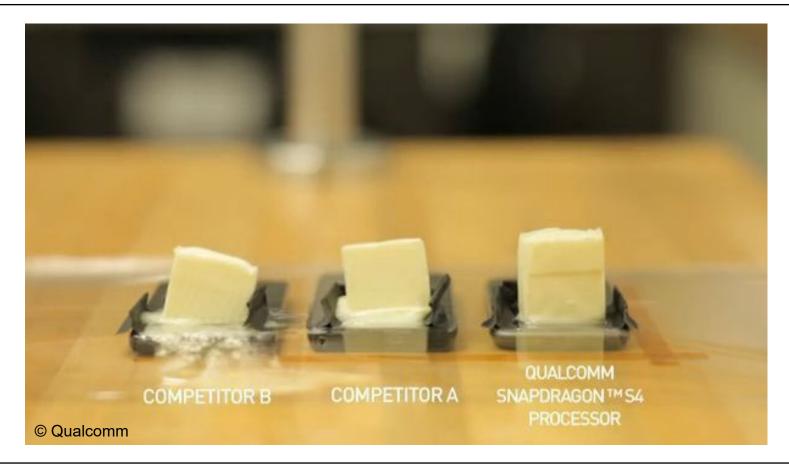


# Power and Energy Consumption of HPC Systems

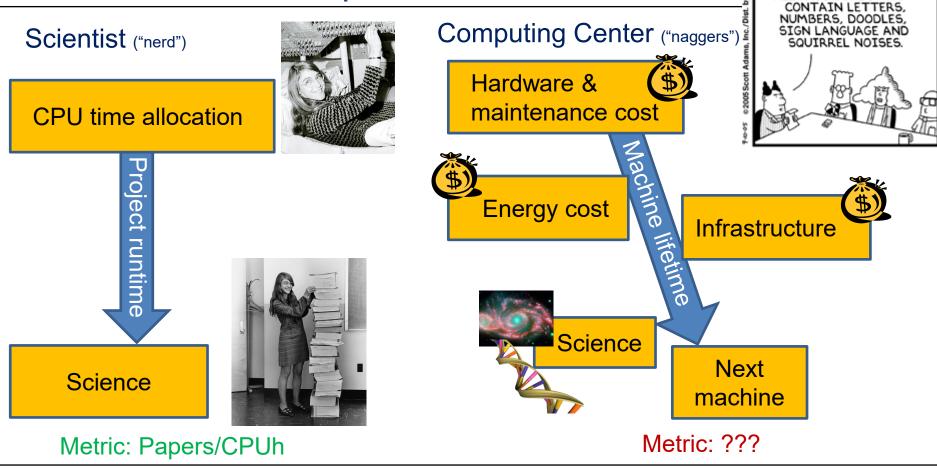
Georg Hager

Erlangen National High Performance Computing Center (NHR@FAU)

#### Points of view: mobile devices



#### Points of view: computational science



STARTING TODAY, ALL PASSWORDS MUST

#### HPC is "hot"

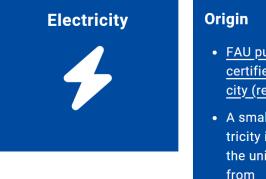
		Rmax	Rpeak	Power
System	Cores	(PFlop/s)	(PFlop/s)	(kW)
Frontier - HPE Cray EX235a, AMD Optimized 3rd Generation EPYC 64C 2GHz, AMD Instinct MI250X, Slingshot-11, HPE D0E/SC/Oak Ridge National Laboratory United States	8,699,904	1,206.00	1,714.81	22,786
<b>Aurora</b> - HPE Cray EX - Intel Exascale Compute Blade, Xeon CPU Max 9470 52C 2.4GHz, Intel Data Center GPU Max, Slingshot-11, <b>Intel</b> D0E/SC/Argonne National Laboratory <b>United States</b>	9,264,128	1,012.00	1,980.01	38,698
<b>Eagle</b> - Microsoft NDv5, Xeon Platinum 8480C 48C 2GHz, NVIDIA H100, NVIDIA Infiniband NDR, <b>Microsoft Azure</b> Microsoft Azure <b>United States</b>	2,073,600	561.20	846.84	
<b>Supercomputer Fugaku</b> - Supercomputer Fugaku, A64FX 48C 2.2GHz, Tofu interconnect D, <b>Fujitsu</b> RIKEN Center for Computational Science Japan	7,630,848	442.01	537.21	29,899
LUMI - HPE Cray EX235a, AMD Optimized 3rd Generation EPYC 64C 2GHz, AMD Instinct MI250X, Slingshot-11, HPE EuroHPC/CSC Finland	2,752,704	379.70	531.51	7,107
	Frontier - HPE Cray EX235a, AMD Optimized 3rd Generation EPYC 64C 2GHz, AMD Instinct MI250X, Slingshot-11, HPE DOE/SC/Oak Ridge National Laboratory United StatesAurora - HPE Cray EX - Intel Exascale Compute Blade, Xeon CPU Max 9470 52C 2.4GHz, Intel Data Center GPU Max, Slingshot-11, Intel DOE/SC/Argonne National Laboratory United StatesEagle - Microsoft NDv5, Xeon Platinum 8480C 48C 2GHz, NVIDIA H100, NVIDIA Infiniband NDR, Microsoft Azure Microsoft Azure United StatesSupercomputer Fugaku - Supercomputer Fugaku, A64FX 48C 2.2GHz, Tofu interconnect D, Fujitsu RIKEN Center for Computational Science JapanLUMI - HPE Cray EX235a, AMD Optimized 3rd Generation EPYC 64C 2GHz, AMD Instinct MI250X, Slingshot-11, HPE	Frontier - HPE Cray EX235a, AMD Optimized 3rd Generation EPYC 64C 2GHz, AMD Instinct MI250X, Slingshot-11, HPE DOE/SC/Oak Ridge National Laboratory United States8,699,904Aurora - HPE Cray EX - Intel Exascale Compute Blade, Xeon CPU Max 9470 52C 2.4GHz, Intel Data Center GPU Max, Slingshot-11, Intel 	SystemCores(PFLop/s)Frontier - HPE Cray EX235a, AMD Optimized 3rd Generation EPYC 64C 2GHz, AMD Instinct MI250X, Slingshot-11, HPE DDE/SC/Oak Ridge National Laboratory United States8,699,9041,206.00Aurora - HPE Cray EX - Intel Exascale Compute Blade, Xeon CPU Max 9470 52C 2.4GHz, Intel Data Center GPU Max, Slingshot-11, Intel DDE/SC/Argonne National Laboratory United States9,264,1281,012.00Bagle - Microsoft NDV5, Xeon Platinum 8480C 48C 2GHz, NVIDIA H100, NVIDIA Infiniband NDR, Microsoft Azure United States2,073,600561.20Supercomputer Fugaku - Supercomputer Fugaku, A64FX 48C 2.2GHz, Tofu interconnect D, Fujitsu RIKEN Center for Computational Science Japan7,630,848442.01LUMI - HPE Cray EX235a, AMD Optimized 3rd Generation EYPC 64C 2GHz, AMD Instinct MI250X, Slingshot-11, HPE2,752,704379.70	SystemCores(PFLop/s)(PFLop/s)Frontier - HPE Cray EX235a, AMD Optimized 3rd Generation EPYC 64C 2GHz, AMD Instinct MI250X, Slingshot-11, HPE DDE/SC/Oak Ridge National Laboratory United States8,699,9041,206.001,714.81Aurora - HPE Cray EX - Intel Exascale Compute Blade, Xeon CPU Max 9470 52C 2.4GHz, Intel Data Center GPU Max, Slingshot-11, Intel DDE/SC/Argonne National Laboratory United States9,264,1281,012.001,980.01Keon CPU Max 9470 52C 2.4GHz, Intel Data Center GPU Max, Slingshot-11, Intel DDE/SC/Argonne National Laboratory9,264,1281,012.001,980.01Keon CPU Max 9470 52C 2.4GHz, Intel Data Center GPU Max, Slingshot-11, Intel DDE/SC/Argonne National Laboratory9,264,1281,012.001,980.01Keon CPU Max 9470 52C 2.4GHz, Intel Data Center GPU Max, Slingshot-11, Intel DDE/SC/Argonne National Laboratory9,264,1281,012.001,980.01Keon CPU Max 9470 52C 2.4GHz, Intel Data Center GPU Max, Slingshot-11, Intel DDE/SC/Argonne National Laboratory9,264,1281,012.001,980.01Keon CPU Max 9470 52C 2.4GHz, MID NDK, Microsoft Azure NIDIA H100, NVIDIA H100, NVIDIA Infiniband NDR, Microsoft Azure NIDIA States2,073,600561.20846.84LUMI - HPE Cray EX235a, AMD Optimized 3rd Generation EPYC 64C 2GHz, AMD Instinct MI250X, Slingshot-11, HPE EuroHPC/CSC2,752,704379.70531.51

Source: Top500.org, June 2024

#### Electrical energy at FAU

#### Origin and consumption of energy at FAU

Approximately 38,000 students and 6,500 members of staff study, research and work in FAU buildings. They require energy, i.e. electricity and heating, in order to carry out their work.



- <u>FAU purchases 100%</u> <u>certified green electri-</u> <u>city (renewable)</u>
- A small share of electricity is produced by the university itself from
  - A cogeneration unit
  - Photovoltaic systems

#### **Necessary for**

- Lighting
- Ventilation and air conditioning systems
- All devices and appliances (from the office to the laboratory)
- Cooling systems etc.

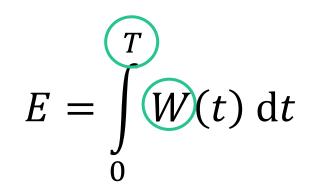
#### Consumption

• Approx. 68 GWh per year

https://www.fau.eu/fausavesenergy/infoszum-energieverbrauch/

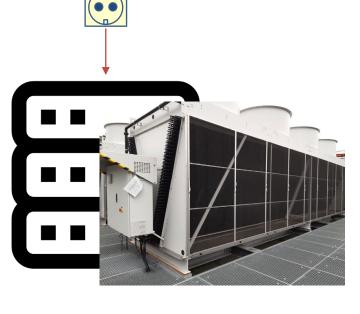
# Power and energy of a computer system

Energy "consumption" (goes into heat):





- T: Runtime
- E: Energy [Wh]
- Example: Fritz cluster at full load 650 kW x 8700 h = 5.7 x 10<sup>6</sup> kWh p.a.





# The price of energy (i.e., why should you care?)

- Energy price φ [€/kWh]
- Assuming 0.30 €/kWh, 1 year of Fritz is

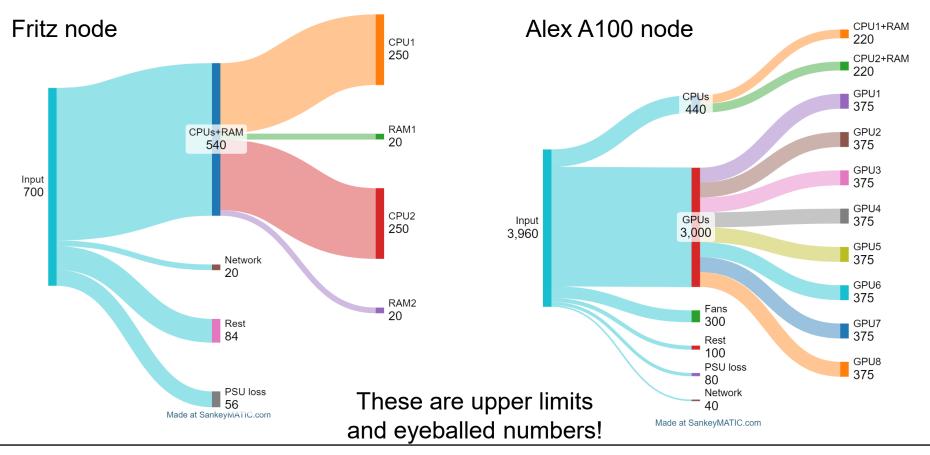
 $W \times T \times \phi = 650 \text{ kW} \times 8700 \text{ h} \times 0.30 \text{ €}/$ = 1.7 × 10<sup>6</sup> € on the energy bill

- How much does my job cost per day?
  - Fritz: 1 node (650 W) for 24 hours ≈ 4.70 €
  - Alex: 1 GPU (300 W) + host share (20 W) for 24 hours ≈ 2.30 €
  - This does not include the cooling infrastructure
    - + 5-10% for Fritz
    - + 20-30% for Alex



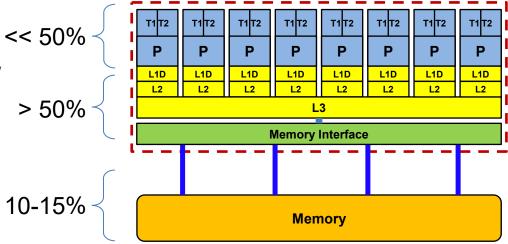


#### Where does the power go in a system?



# Power dissipation on the chip

- Data transfer is the most energy-intensive thing in a computer system today!
- All else being equal, reducing data transfers reduces the power dissipation
- The further away the data, the more energy it costs
- Good news: Long-distance <data lines are also slow and few
- Significant power even for "idle" chip!



#### How do I know how much power/energy my job uses?

- Tools (CPU)
  - LIKWID tools (based on RAPL)
    - Available as module on NHR@FAU clusters
    - https://github.com/RRZE-HPC/likwid
    - likwid-perfctr -g ENERGY [-m] -c N:0-71 ./a.out
    - likwid-mpirun -g ENERGY [-m] -np 360 ./a.out
  - ClusterCockpit (based on LIKWID)
    - <u>https://github.com/ClusterCockpit</u>
- Tools (GPU)
  - Nvidia + AMD tools nvidia-smi / rocm-smi
  - ClusterCockpit (based on nvidia-smi)

#### ClusterCockpit

#### Job-specific monitoring accessible from HPC portal

#### Job energy consumption will be available soon™

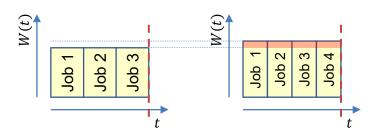


Power and Energy of HPC Systems | HPC Café

# Reducing the energy per job

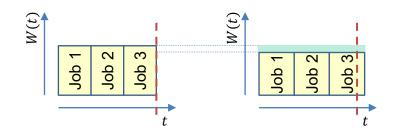
Reduce the job's runtime T

- Better overall use of resources
- Better use of your resource allotment
- More "science per CPUh"
- Probably higher or lower power dissipation of the system



Reduce the job's power W(t)

- Lower power dissipation of the system
- Probably some performance loss
  → probably less "science per CPUh"



# Tuning knobs for job energy reduction (user view)

- Clock speed reduction ("underclocking")
  - Power Energy T CPU/GPU-h T ⇒
- Concurrency throttling ("use less hardware")
  Power Energy CPU/GPU-h

#### Summary: It's complicated.

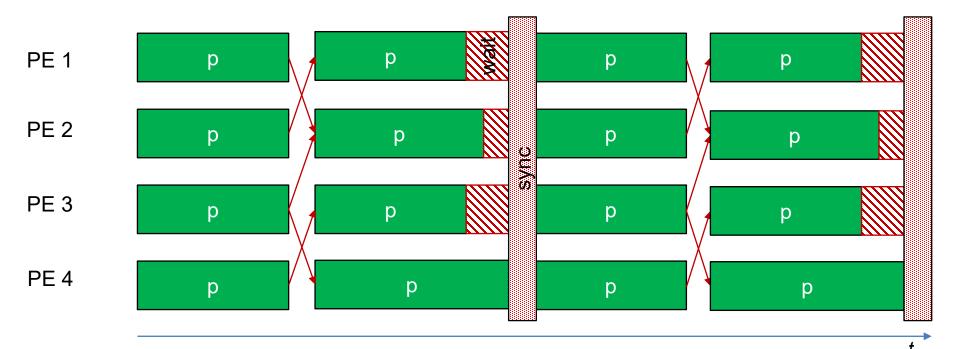
# Code optimization for runtime and energy

- Use the best algorithm
  - E.g.,  $N^2 \rightarrow N \log N$
- Use optimized libraries
  - E.g., OpenBLAS → MKL
- Use aggressive compiler optimization
  - E.g., -O1  $\rightarrow$  -Ofast -xHost
- Do less work
  - E.g., use sparse matrices instead of dense

- Balance the workload
  - All "devices" finish at the same time
- Transfer less data from far away
  - Cache blocking / register reuse
  - Avoid network communication
- Hide communication overhead
  - Asynchronous/bidirectional network communication
  - Asynchronous GPU data transfers

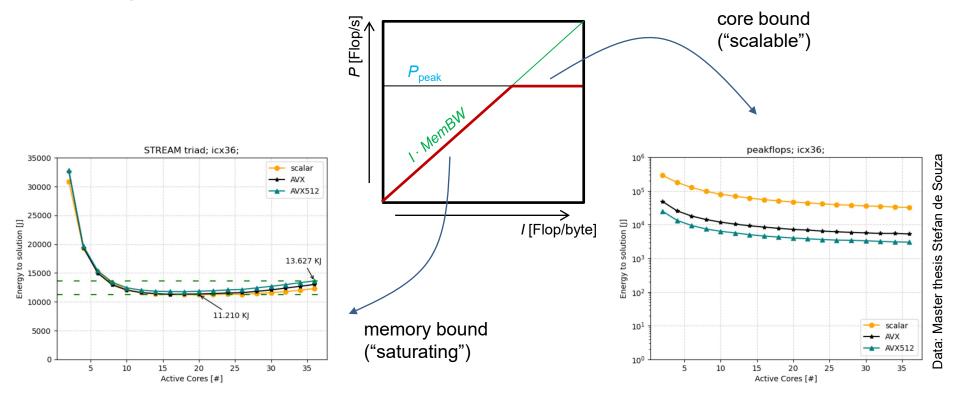
#### Concurrency and resource efficiency

- In general, less parallelism is better for resource efficiency
- Parallelism incurs overhead, which is waste



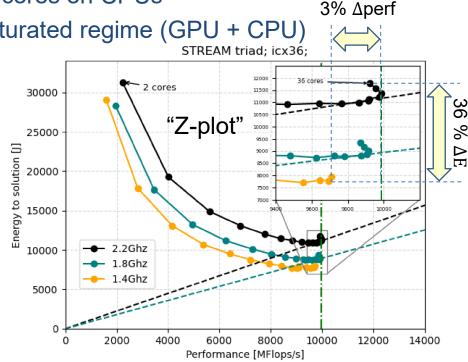
#### ... unless we have relevant shared resources

Roofline point of view on a CPU socket level



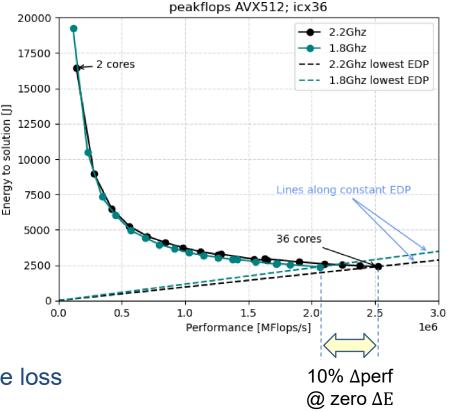
# Memory-bound code

- Shows "performance saturation" vs. # of cores on CPUs
- Weak sensitivity to clock frequency in saturated regime (GPU + CPU)
  - Significant energy saving potential
- Optimal "operating point" is crucial
  - # of cores at saturation point  $\rightarrow$  min E
  - Clock speed
- Setting the concurrency
  - OMP\_NUM\_THREADS
  - -nperdomain
  - ... whatever, but use pinning!
- Setting the clock speed (CPUs)
  - likwid-setFrequencies -f <GHz>
  - srun --cpu-freq=<kHzmin>-<kHzmax>:performance ./a.out



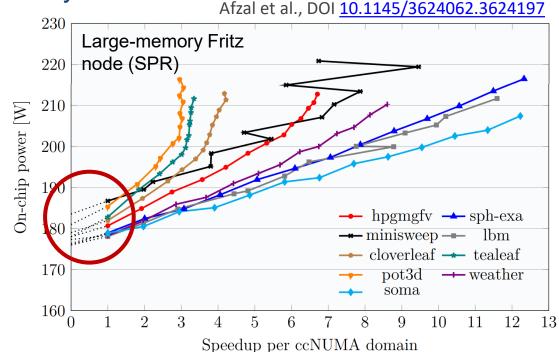
### Core-bound ("scalable") code

- No on-chip scaling bottleneck
- Performance scales with # cores
- The more cores, the lower the energy to solution
- Performance sensitive to clock speed
  Ideally, proportional
- Power is sensitive to clock speed
  - $W \propto f^{\alpha}, \, \alpha \in [1, \dots, 3]$
  - There is an optimal frequency for minimal energy to solution
  - ... at the price of significant performance loss



#### Chip baseline power today

- Much of the CPU/GPU chip power today is "baseline power"
- Extrapolation to "zero concurrency"
- Fritz:
  - ICL socket W<sub>0</sub> ≈ 100 W (TDP = 250W)
  - SPR socket W<sub>0</sub> ≈ 180 W (TDP = 350W)
- Sandy Bridge (2012): 20% baseline power



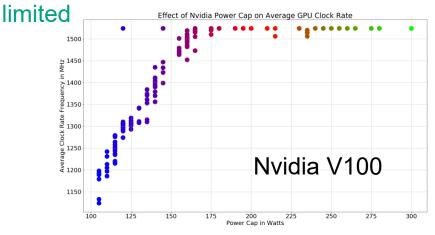
### What can the computing center do?

# Disable Turbo Mode, fixed frequency

- Reliable performance behavior
- Damped but still significant variations in power
- Pessimistic power power limit
- Infrastructure must provide power headroom

#### Device power capping

- Reliable upper power limit
- Power variations turn into performance variations
- Infrastructure can be tightly power



Patki et al., DOI: 10.1109/WORKS49585.2019.00009

# So... what should you do?

0<sup>th</sup> order advice:

- Make your code run faster
  - Latest code versions
  - Best/latest compilers
  - Best libraries
  - Code optimization
- Avoid waste
  - Reasonable concurrency
  - Controlled overhead
  - "If it finishes in tolerable time, don't scale further"

#### For memory-bound code:

- Significant energy savings via clock speed reduction
  - Run benchmarks
  - Talk to your friendly computing center
- Consider applying for a KONWIHR project
  - www.konwihr.de







Friedrich-Alexander-Universität Erlangen-Nürnberg

# Thank you.