



Sustainable Computing

Ivona Brandić

TU Wien

ivona.brandic@tuwien.ac.at



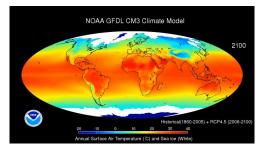


Computational Power

Simulation

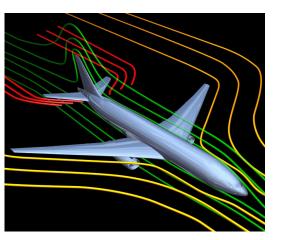


Mechanical Structure Simulation

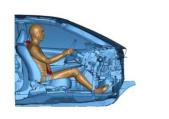


Climate Prediction

Optimization



Airflow Optimization

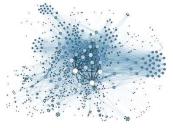


Finite Element Simulation Hyper Parameter Optimization

Today: Analytics, AI

A5ASC3.1	14 STREUPPSOT	DI LENZ DRAMEERY	OTCTOR O	CELEVERADEURPO	TED OCCTORO S	NEKEPOGOGGSAVOLY	NECOS TEL A	1.04
B4F917.1						FREEPDGDGSSAVOL		
9951V2.1						ADSGGDKTCSAVV/WW		
89GSN7.1						MEKOPOGOGSSAVOIN		
088056.1	30 SESTUPPTORT	HUTLYE PRINTER IL.	Dribkk.	GLESKEEHEEDHKK	ELVHP HHHHHU	RATASVEEGIKALOLY	ALS SKLALEVEN	100
	30 SESTUPPIOR	RUHVVRREVDIEGE	FILLER.	БНУРННОН ЕРНИКС	LEHENPUMMHH SGEN	AAPASVEDGIEVLOAY	SKEV SKKLLUF VK	120
Q0D4Z3.2								
89MV4/8.1						.VASSEKDGLEVLQLY		
Q0IYC5.1						SSSSSVEDGIETLOLY		
A9NJ46.1						EAKEPHL DOSSVVOF Y		
090500.1						AVSSDODGIKILELYS		
Q2HRI7.1	25 NYSIMPPKORT	RDAVKNELIETLST	SVLTKR.	GTMSADEASAAAIO	LEDERFSVANA	.SSSTSNDN/TILEVY	SKEISKRMIETVK	110
09M7N3.1	28 SEKIMPPTORI	REAVVRREVETLTS.	SVLSKR.	GVIPEEDATSAARI	LEERFSVRSV, ASAAS	TGGRPEDENIEVEHIM	BOEIXORVVESAK	119
09M7N6.1						, TASDADDGIEILOVY		
09LE82.1						FONEPDGDGTSAVHVY		
09M651.2	13 SIKLMPPSLPT	REALIERITNNESS	TIFTEK	GSUTKODATENAKR	TEDIAFSTANO	FEREPOGOGGSAVOLY	KECSKLILEVEK	100
898748.1	48 SLSIMPPTORT	PDAVITELIETESS.	SVI SKR	GTTSHDERESOOPP	TEREPECVENT	ATSAEDOGLEILOLY	SKET SPEMI OTVX	133

DNA Sequence Analysis (e.g., Genomic sequencing of SARS-CoV-2)



Social Network Analysis

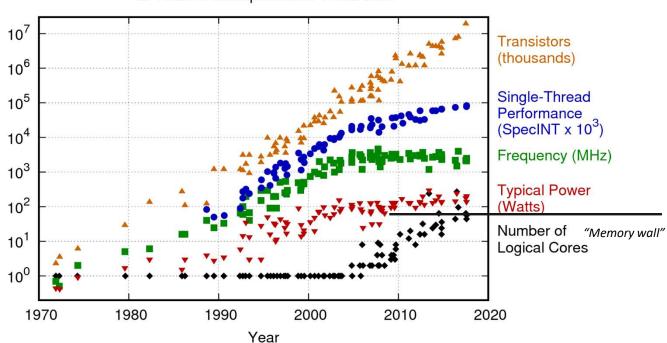


Recommendation Engines

2



Problem 1: Practical Limitations



42 Years of Microprocessor Trend Data

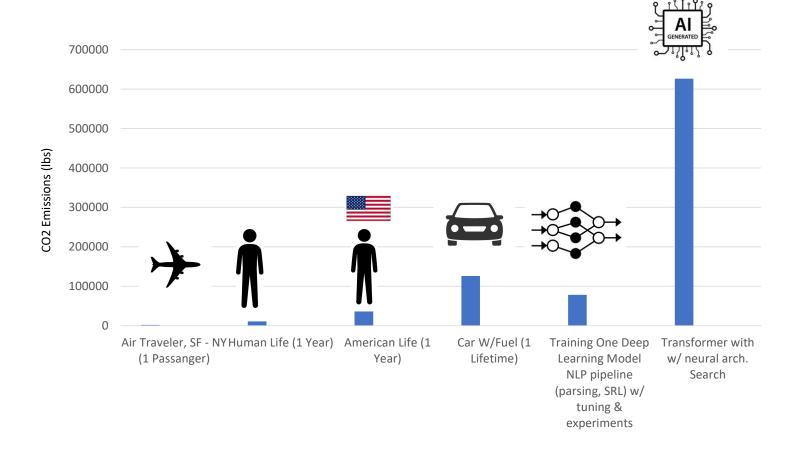
Original data up to the year 2010 collected and plotted by M. Horowitz, F. Labonte, O. Shacham, K. Olukotun, L. Hammond, and C. Batten New plot and data collected for 2010-2017 by K. Rupp

- #cores per chip doubles every 18 months instead of clock
- CPU-memory communication is becoming a bottleneck
- Too much heat is produced
- As transistors get smaller, power density increases because these do no scale with size anymore
- → practical limitations to
 processor frequency to around 4
 GHz since 2006





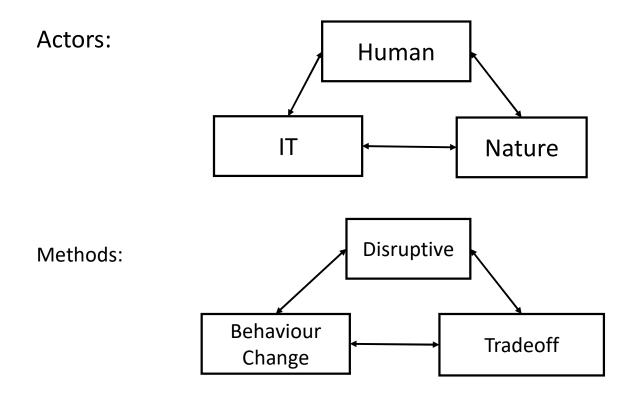
Problem 2: CO₂ Footprint of (generative) AI



Source: Emma Strubell, Ananya Ganesh, Andrew McCallum: Energy and Policy Considerations for Deep Learning in NLP, ACL (1) 2019: 3645-3650 Inspiration for Visualisation: Keren Bergman, Multicore World 2023 https://2019multicoreworld.files.wordpress.com/2023/02/bergman-keren-23.pdf

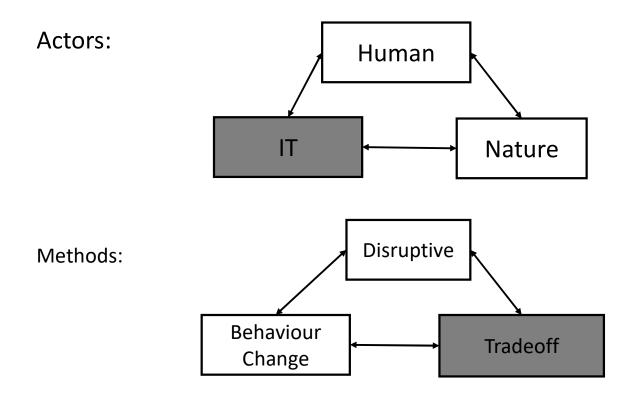


Computational Sustainability



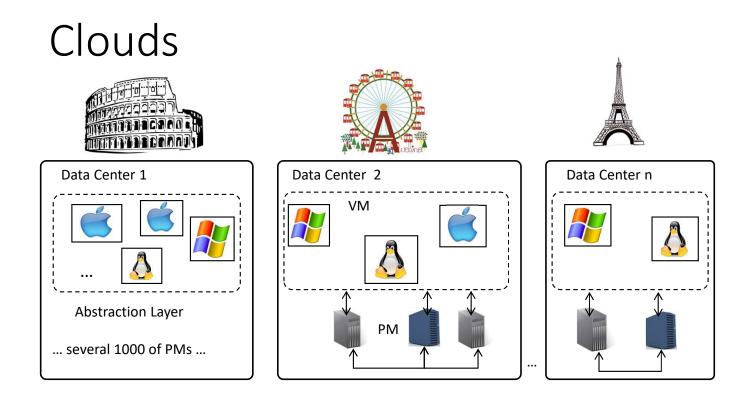


Computational Sustainability







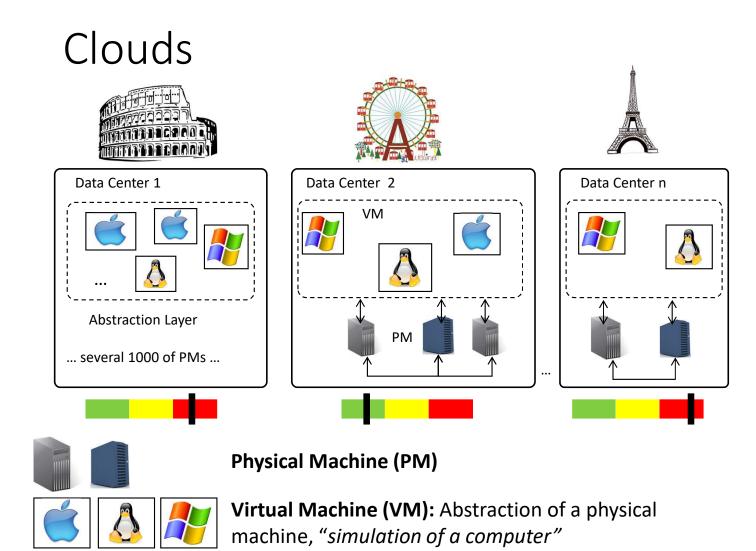


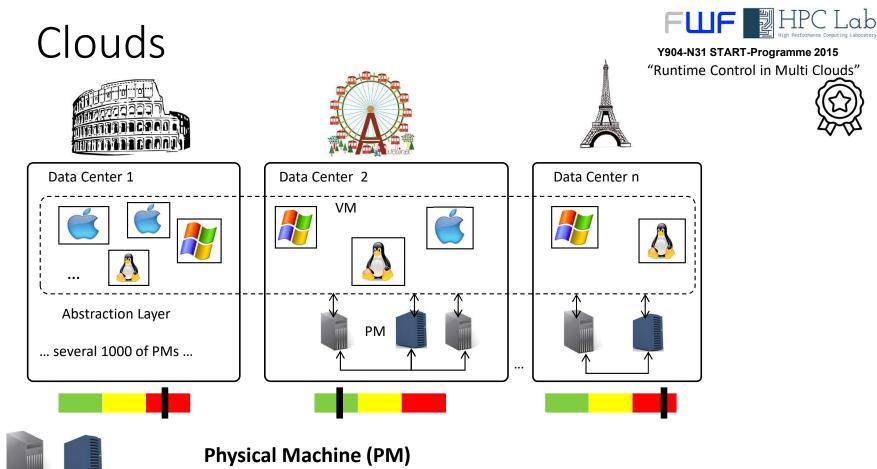


Physical Machine (PM)

Virtual Machine (VM): Abstraction of a physical machine, "simulation of a computer"









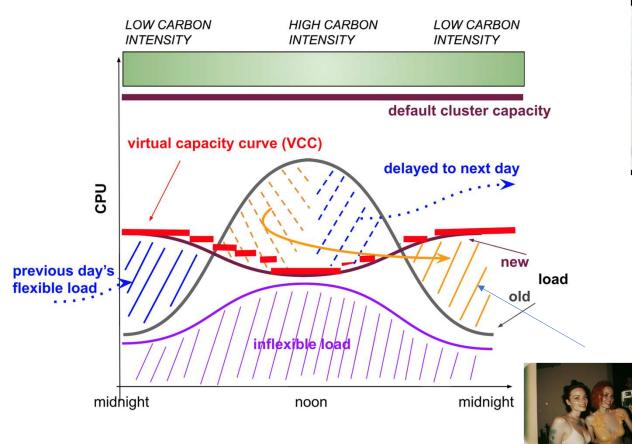
Virtual Machine (VM): Abstraction of a physical machine, "simulation of a computer"

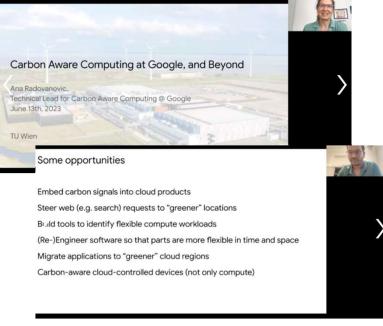
Cloud: economic and ecological data center solutions

Source: Damien Borgetto, Michael Maurer, Georges Da Costa, Jean-Marc Pierson, Ivona Brandic: Energyefficient and SLA-aware management of IaaS clouds. e-Energy 2012: 25



Workload Shift in Space and Time





Ana Radovanovic (Google) & Shashi Ilager (TU) Lecture at TU Wien: **"Data Intensive Computing"**



Source: Radovanović, Ana, et al. "Carbon-aware computing for datacenters." IEEE Transactions on Power Systems 38.2 (2022): 1270-1280.

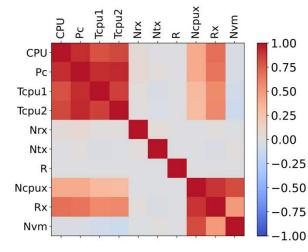
12

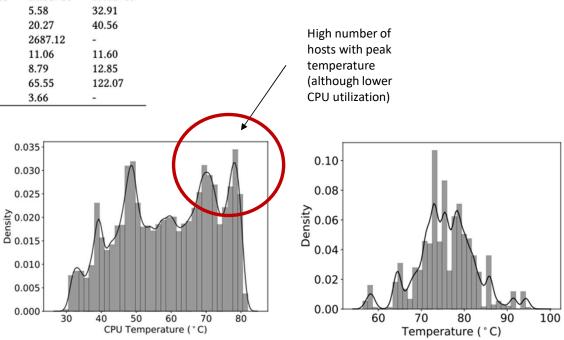


13

Thermal Intricacies in Data Center

Metrics	Min		Max		Mean		SD	
	D1	D2	D1	D2	D1	D2	D1	D2
CPU	0.0	0.0	68.36	86.31	10.35	22.29	14.21	18.51
R_x	0.49	1.44	64.42	191.41	35.60	46.77	14.12	35.40
N _{Rx}	0	0	10.65+e8	52.63+e8	28.75+e5	25.59+e6	17.71+e6	20.94+e7
N_{Tx}	0	0	11.24+e8	59.97+e8	22.86+e5	14.55 + e6	14.66+e6	19.41+e7
Num	0	0	54	261	9.6	9.01	5.58	32.91
NCPUx	0	0	128	320.00	55.93	39.81	20.27	40.56
$fs_1 - fs_4$	280	-	13941.66	-	8804	-	2687.12	-
T_{CPU1}	29.14	26.66	82.01	79.23	57.00	59.52	11.06	11.60
T_{CPU2}	25.46	25.58	77.95	81.40	48.29	59.76	8.79	12.85
$P^{$	55.86	260.00	448	806.00	205.58	546.44	65.55	122.07
Tin	4	-	25.75	-	17.73	-	3.66	-





MONASH University

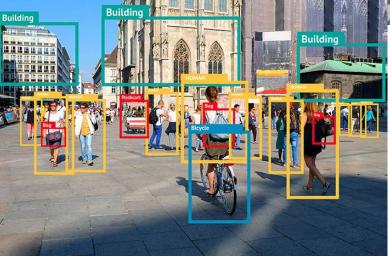
S. Ilager

Source: S. Ilager, A. N. Toosi, M. Raj Jha, I. Brandic, R. Buyya, "A Data-driven Analysis of a Cloud Data Center: Statistical Characterization of Workload, Energy and Temperature", In Proceedings of the 16th IEEE/ACM International Conference on Utility and Cloud Computing (UCC2023), Vancouver, Messina, Italy, December 4-7, 2023.

Slide: cortesy Shashi Ilager





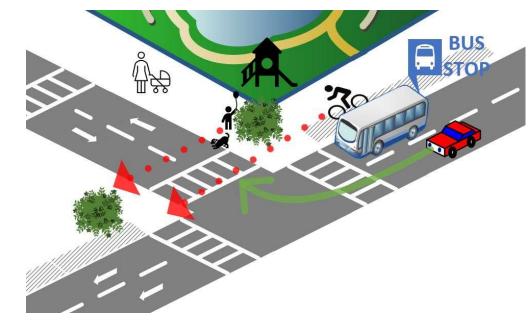


Symbolic Data Representation



Edge Computing in Action: Smart City

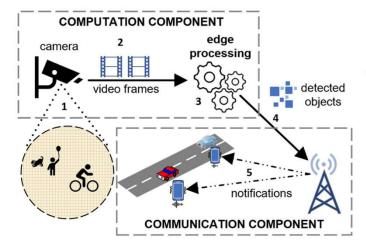
- Traffic accidents
- causing injuries and deaths
- Distractions, poor visibility (e.g., bad road and weather conditions), ...
- Drivers' brake reaction time
- 1500ms on average



Deaths among pedestrians and cyclists: 29% of all EU road deaths

ETSC (European Transport Safety Council) PIN Flash Report 38

Smart City





http://intrasafed.ec.tuwien.ac.at

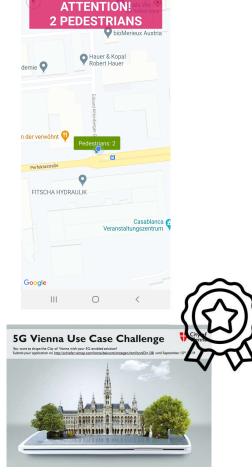


Slide: courtesy Ivan Lujic (Ericsson Nikola Tesla d.d.)



Source: Lujic, De Maio, Pollhammer, Bodrozic, Lasic, and Brandic, "Increasing Traffic Safety with Real-Time Edge Analytics and 5G," EdgeSys, pp. 19-24, 2021.

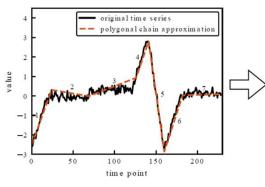




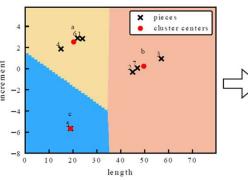
ଛ ♥ ፼...। 85% ∎

15:29 🚨 🖬 🌻

Adaptive and Online Symbolic Representation







ERICSSON S Ericsson Nikola Tesla d.d.



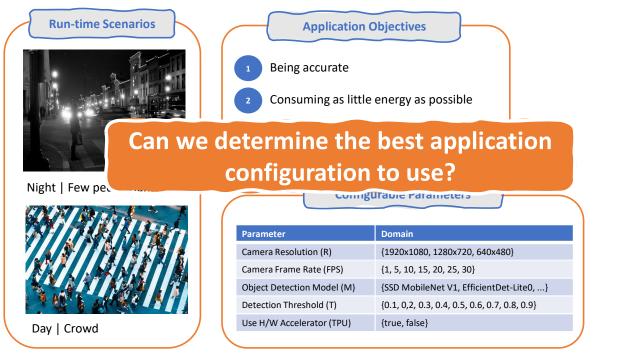
Daniel Hofstätter, Shashikant Ilager, Ivan Lujic, Ivona Brandic. **SymED: Adaptive and Online Symbolic Representation of Data on the Edge.** 29th International European Conference on Parallel and Distributed Computing, 28 August -1 September 2023 Limassol, Cyprus.







An Edge Computing Pedestrian Detection Scenario



This slide has been designed using images from Flaticon.com

ASE 2023 | Sep 11th - 15th | Luxembourg

Slides: Cortesy Alessandro Tundo (University Milano-Bicocca)

Source: Alessandro Tundo, Marco Mobilio, Shashikant Ilager, Ivong Brandic, Ezio Bartocci, Leonardo Mariani. An Energy-Aware Approach to Design Self-Adaptive Albased Applications on the Edge. 38th IEEE/ACM International Conference on Automated Software Engineering (ASE2023), Sep 11, 2023 - Sep 15, 2023, Luxembourg.

When is an H/W accelerator required?

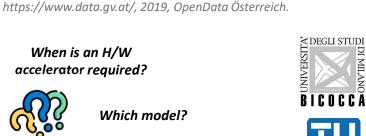
Source: Vienna Municipal Department 33, "Traffic

lights with/without audible signal devices in Vienna,"

Which model?

How many frames

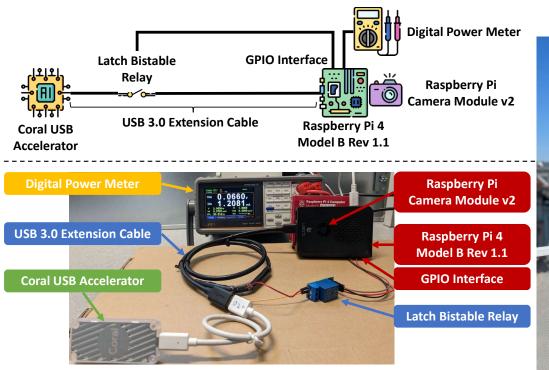
per second?



18



Evaluation Testbed



This slide has been designed using images from Flaticon.com

<image>

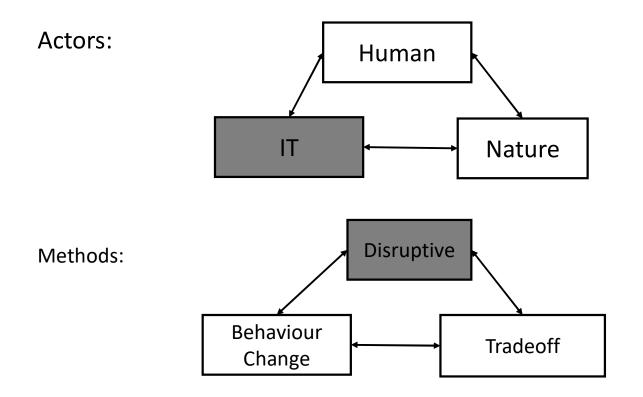
ASE 2023 | Sep 11th = 15th | Luxembourg

Slides: Cortesy Alessandro Tundo (University Milano-Bicocca)

Source: Alessandro Tundo, Marco Mobilio, Shashikant Ilager, Ivona Brandic, Ezio Bartocci, Leonardo Mariani. An Energy-Aware Approach to Design Self-Adaptive Albased Applications on the Edge. 38th IEEE/ACM International Conference on Automated Software Engineering (ASE2023), Sep 11, 2023 - Sep 15, 2023, Luxembourg.



Computational Sustainability









Gordon Moore: Moore's Law (1929 - 2023)



Data volumes are growing faster than the processing power

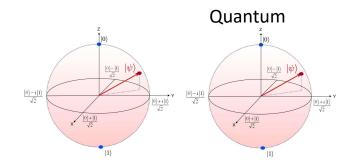


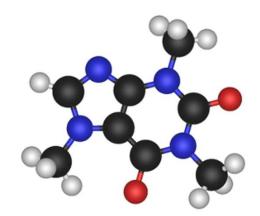
Alternatives:

- Neuromorphic Computing
- Quantum Computing



Beyond 0 and 1





Von Neumann



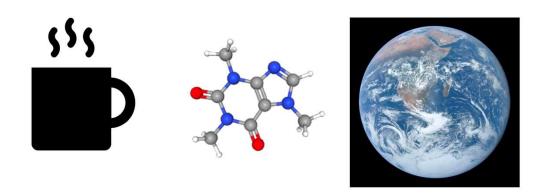
Bottom up approach

- Variational Quantum Linear Solver (VQLS)
- Quantum Eigenvalues -> Native 3d modeling of scientific applications

Problem: Currently quantum systems can be used by quantum researchers only!



A cup of coffee?



- Representing the energy configuration of a single caffeine molecule at a single instant requires approximately 10⁴⁸ bits in a classical computer
- Can be done using 160 logical qubits on a quantum machine

"Every time you add a qubit, you double your possible outcomes, With 20 qubits there are a million possible outcomes. With 100 qubits, you have more possibilities than there are bits in all the hard drives in the world. With 300 qubits—that's more possibilities than there are particles in the universe."

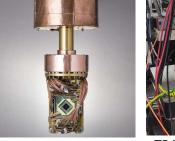
https://quantum.duke.edu/2020/10/1 6/more-possibilities-than-there-areparticles-in-the-universe/

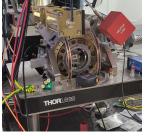
Courtesy Nico Einsiedler (IBM), TU Vienna Quantum Fall Fest 2022



Known Quantum Speedup

- Grover's algorithm (unstructured search): $O(\sqrt{n})$ vs O(n), developed 1996
- Shor's algorithm (finding the prime factors in integer): Polynomial vs Exponential, developed 1994
- Quantum ML
 - Bayesian Inference: quadratic
 - SVM: exponential
 - Reinforcement Learning: quadratic
- In reality
 - Lack of standardization
 - Data transformation / quantum state preparation
 - Decoherence
 - Noise





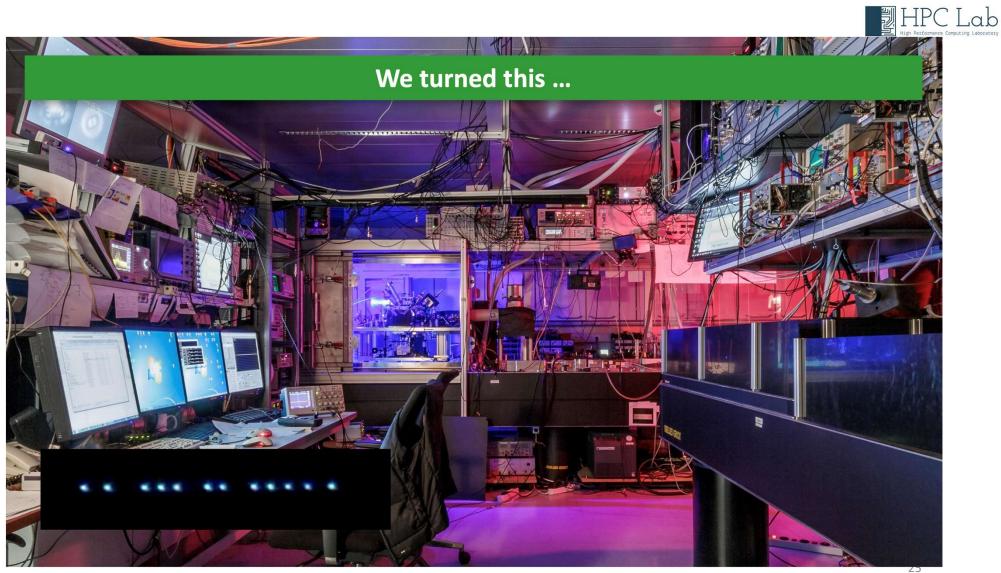
D-WAVE

TRAPPED ION



SUPERCONDUCTING

Source: Machine Learning: Quantum vs Classical, Tariq M. Khan et al., IEEE Access, November 2020



Slide: courtesy Thomas Monz, Uni Innsbruck & AQT





... into this

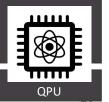


AQT DEMONSTRATED:

- 50+ ion-qubits
- 24-qubit entanglement
- Shor's algorithm
- Quantum Error Correction
- Fault-tolerant performance
- Demo'd finance applications
- Demo'd security applications
- Demo'd chemistry applications
- ...

WITH OUR SYSTEM BEING:

- Rack-mounted
- Cloud-accessible
- Data-center compatible



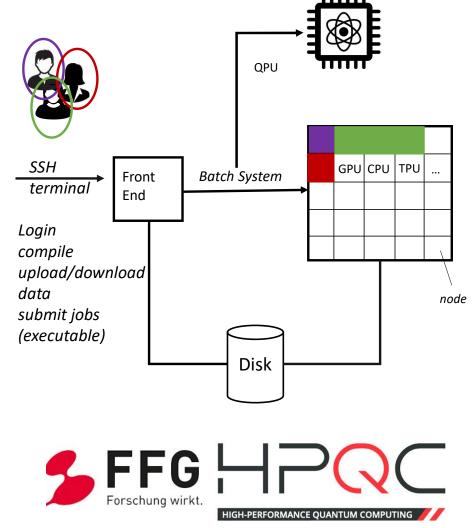
Slide: courtesy Thomas Monz, Uni Innsbruck & AQT



HPC Cluster



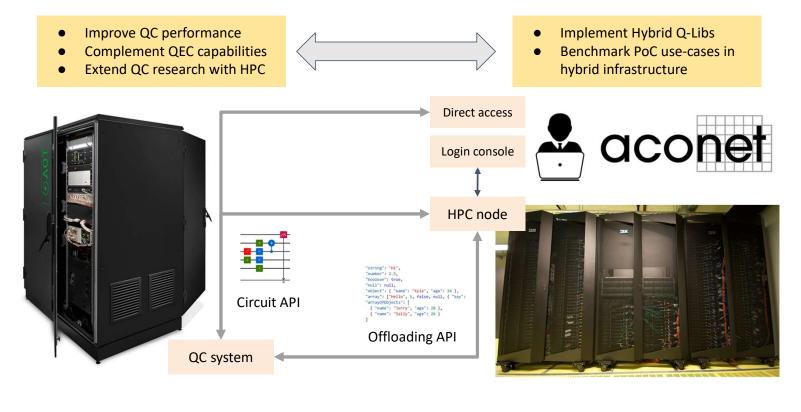
- Each "node" has its own operating system
- Nodes are interconnected with a network cable
- Higher performance demand more processors
- Accessed via front-end node/computer
- Shared with may users



27



HPQC Cluster







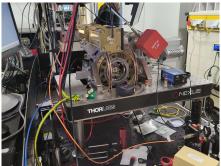
Installation of HPQC



- Move from Ca+ to Ba+New system with
- stage 1: 10x higher T₁ stage 2: infinit. Higher T₁
- 2q error rate: legacy: < 10⁻² target: < 10⁻³
- Init error
 legacy: < 10⁻³
 target: < 10⁻⁴
- Readout error legacy: < 10⁻³ -> ~ 10⁻⁴ target: < 10⁻⁵

Courtesy Experimentalphysik, Univ. Innsbruck



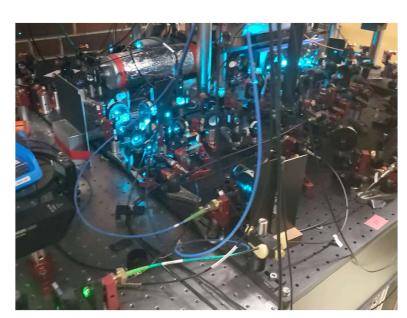


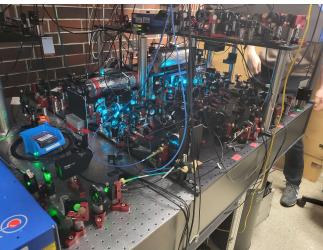
lon-trap quantum computer

universität innsbruck

Courtesy Experimentalphysik, Univ. Innsbruck Pictures, video: courtesy Vincenzo de Maio





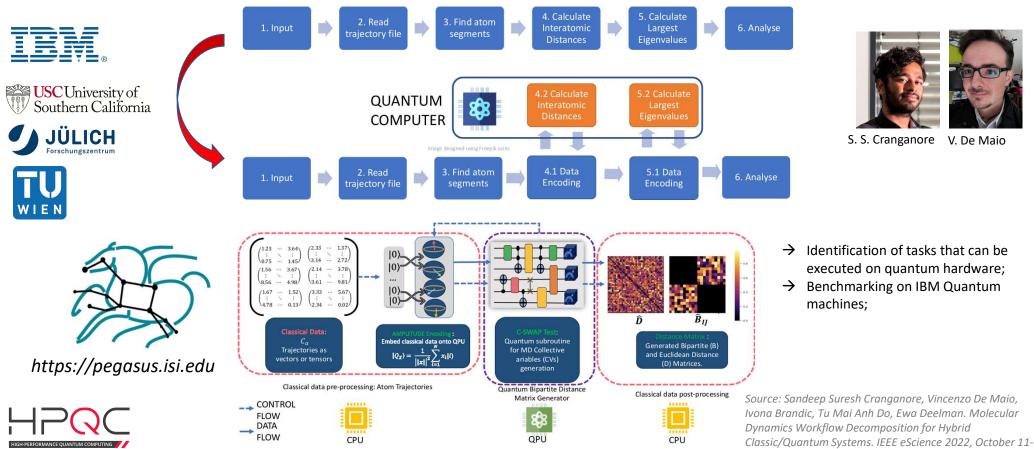




30



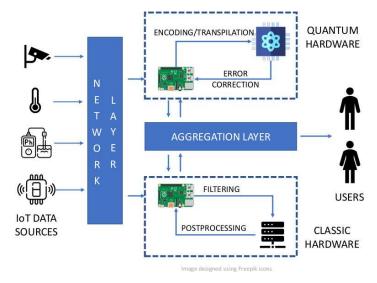
Benchmarking Molecular Dynamics Application

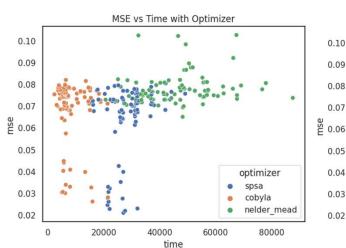


Classic/Quantum Systems. IEEE eScience 20 14, 2022 Salt Lake City, Utah, USA.



Benchmarking Quantum Machine Learning

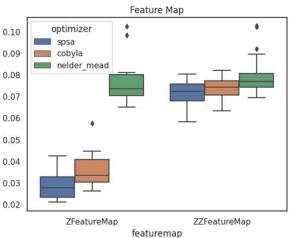






S. Herbst

V. De Maio



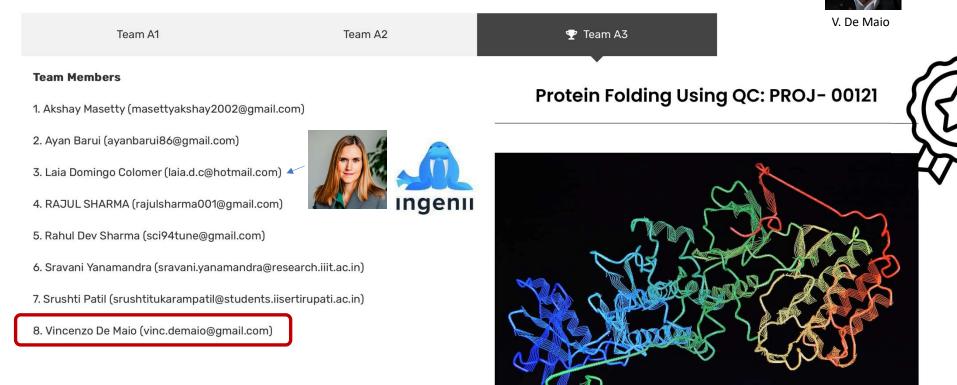
- → Benchmarking of Quantum Regression on typical IoT dataset
- \rightarrow Conceptual design of a Edge-enhanced pipeline for QML
- \rightarrow Preliminary results on IBM machines



S. Herbst, V. De Maio, I. Brandic, "Streaming IoT Data and the Quantum Edge: A Classic Quantum-Machine Learning Use Case", International Workshop on Urgent Analytics for Distributed Computing, co-located with 32 EuroPar 2023

What's next: Protein Folding

Submission



Source: https://quantumopenai.com/protein-folding-using-qc-proj-00161/







Outreach and Teaching







Co-organisation/sponsorship by CS TU Wien, Physik TU Wien, AQT





F. Zilk

M. Kanatbekova



N. Friis

T. Guggemos V. De Maio

To our knowledge first **joint lecture** on 'Hybrid classical-quantum systems' with a focus on applications currently attended by about 30 students Organised by **TU CS**, TU Physics, Uni Wien Photonics

U V V L L



Conclusion

- Trade off
 - Multiple dimensions: accuracy, maintainability, modularity, energy consumption, ...
- QPU for very specific operations
 - Chemistry
 - ML
- Challenge of integrating hybrid systems
- Mindset and education
- Limited hardware availability
 - Importance of simulators for teaching and engineering
 - Importance of benchmarking on real machines
- Focus on telescope technology



Thanks to funding agencies and my team



2010





Horizon 2020 European Union funding for Research & Innovation





chist-era













2021

FШF

FFG

Forschung wirkt



2017

86% of my group are third party funded – Thank you!



Quantum Hardware

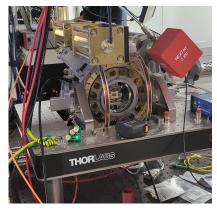


D-WAVE

SUPERCONDUCTING

- No "best" technology at the moment
- No standards
- Every machine different architecture
- Integration \rightarrow

FFG Flagship Project High Performance Integrated **Quantum Computing (HPQC)**



TRAPPED ION*

*Courtesy of University of Innsbruck, department of experimental physics



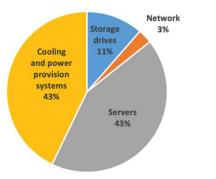


Cloud DC - Temperature Evolution



Understanding Thermal Behaviour in DC

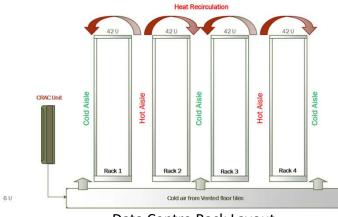
Features	Definition		
CPU	CPU Load (%)		
R	RAM- Random Access Memory (mb)		
R_x	RAM in usage (mb)		
NCPU	Number of CPU cores		
N _{CPUx}	Number of CPU cores in use		
N _{Rx}	Network inbound traffic (Kbps)		
N_{Tx}	Network outbound traffic (Kbps)		
P_c	Power consumed by host (Watts)		
T_{CPU1}	CPU 1 temperature (°C)		
T _{CPU2}	CPU 2 temperature (℃)		
fs_1	fan1 speed (RPM)		
fs_2	fan2 speed (RPM)		
fs3	fan3 speed (RPM)		
fs_4	fan4 speed (RPM)		
Tin	Inlet temperature (°C)		
N _{vm}	Number of VMs running on host		





S. Ilager

Figure 1. Fraction of U.S. data center electricity use in 2014, by end use. Source: Shehabi 2016.



Data Centre Rack Layout

Source: S. Ilager, K.Ramamohanarao, and R. Buyya, Thermal Prediction for Efficient Energy Management of Clouds using Machine Learning, IEEE TPDS, Volume 32, No. 5, Pages: 1044-1056, USA, May 2021.

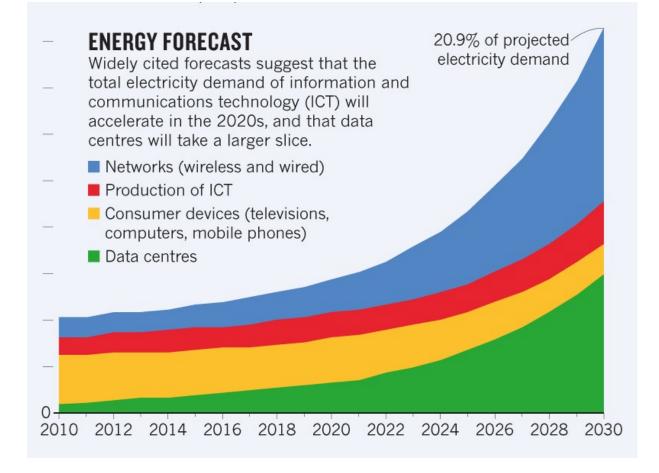
https://energyinnovation.org/2020/03/17/how-much-energy-do-data-centers-really-use/

Slide: cortesy Shashi Ilager

38



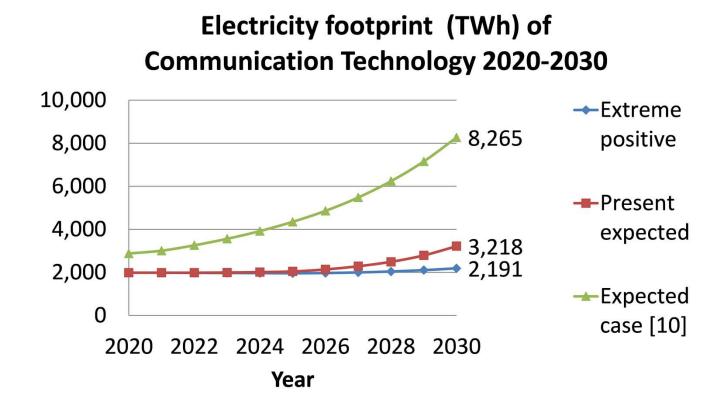
Problem 2: Explosion of Energy Demands



Nicola Jones. How to stop data centres from gobbling up the world's electricity, Nature, 12. Sep, 2018.



Problem 2: Explosion of Increasing Energy Demands



Source: SG Andrae, Anders. "New perspectives on internet electricity use in 2030." Engineering and Applied Science Letter 3.2 (2020): 19-31.