

Introduction to Computational Sustainability

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TU Wien

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European
Commission

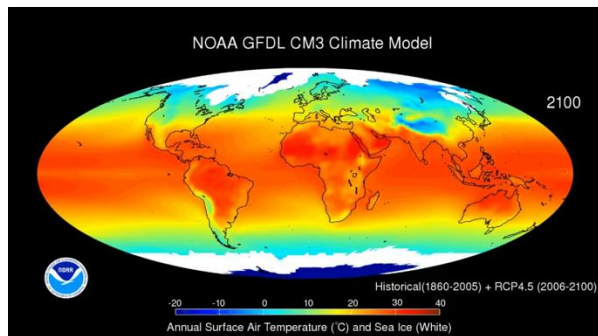


Computational Power

Simulation

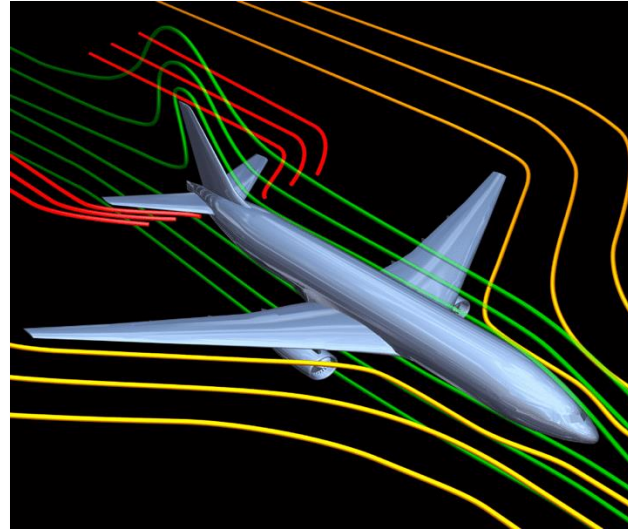


Mechanical Structure Simulation

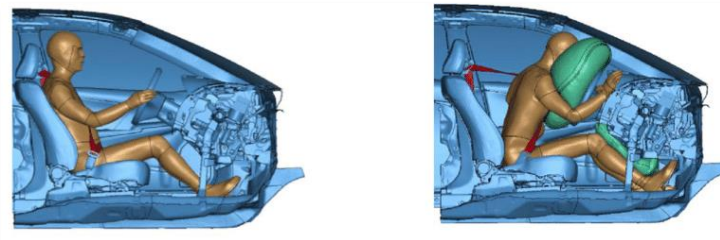


Climate Prediction

Optimization



Airflow Optimization



Finite Element Simulation
Hyper Parameter Optimization

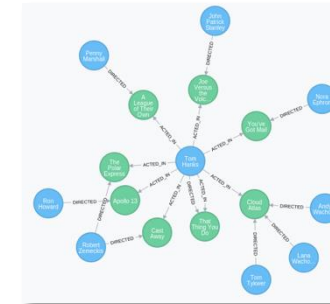
Today: Analytics, AI, LLMs,...

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ASASC3,1 14 SIKLUPPSOTLLEVERHANNLST..PSIFTRN..YVSEKTEAPENKQIEEACSTANO.....HYKEPCDCGSAVOLYKKECKSLIEVLK 101
B4F917,1 13 SIKLUPPSSTRIHVOAMNHLST..EETFSRN..YLLGKDEAHENKATTEELCALADE.....HFREEDPCGSSAVOLYKKECKSLIEVLK 100
P9S1V2,1 23 YKLUUPPSSTREYDORMLKSS..NCTES..FADIELKONCHAKKATEEVEGRIOL.....ROSGDCTESANMYYAKHAKKLETLK 109
B9GSH7,1 13 SIKLUPPSSTRIHVOAMNHLST..EETFSRN..YLLGKDEAHENKATTEELCALADE.....HFREEDPCGSSAVOLYKKECKSLIEVLK 100
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Q0D423,2 44 SFSIUPPTORTKOWHVLVETLST..PSILSKR..YVWPEACGAAHAAVEAHAYTES..SSAAAPBSVEGDEVLQYKKECKSLIEVLK 135
B9MVU8,1 56 SFSIUPPTORTKOWHVLVETLST..PSILSKR..YVWPEACGAAHAAVEAHAYTES..SSAAAPBSVEGDEVLQYKKECKSLIEVLK 141
Q0TY05,1 29 SFSIUPPTORTKOWHVLVETLST..PSILSKR..YVWPEACGAAHAAVEAHAYTES..SSAAAPBSVEGDEVLQYKKECKSLIEVLK 121
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Q9H7H3,1 28 SFSIUPPTORTKOWHVLVETLST..PSILSKR..YVWPEACGAAHAAVEAHAYTES..SSAAAPBSVEGDEVLQYKKECKSLIEVLK 119
Q9H7H6,1 29 SFSIUPPTORTKOWHVLVETLST..PSILSKR..YVWPEACGAAHAAVEAHAYTES..SSAAAPBSVEGDEVLQYKKECKSLIEVLK 110
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O9H601,2 13 SIKLUPPSSTRIHVOAMNHLST..EETFSRN..YLLGKDEAHENKATTEELCALADE.....HFREEDPCGSSAVOLYKKECKSLIEVLK 100
B9K740,1 40 SFSIUPPTORTKOWHVLVETLST..PSILSKR..YVWPEACGAAHAAVEAHAYTES..SSAAAPBSVEGDEVLQYKKECKSLIEVLK 133
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DNA Sequence Analysis
(e.g., Genomic sequencing of SARS-CoV-2)

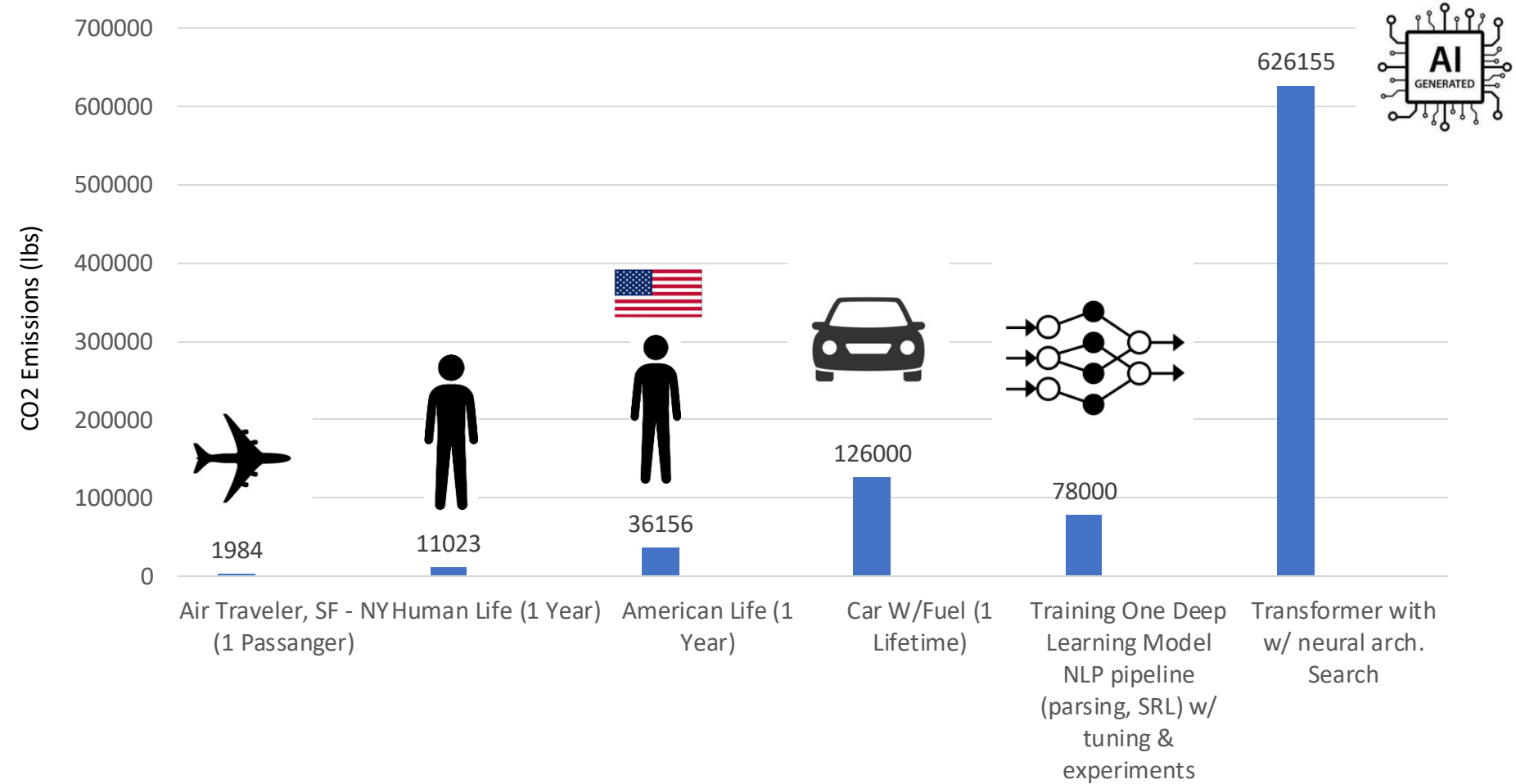


Large Language Models



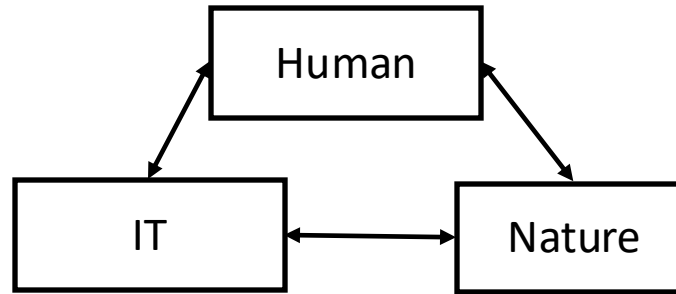
Recommendation Engines

Problem: CO₂ Footprint of (generative) AI

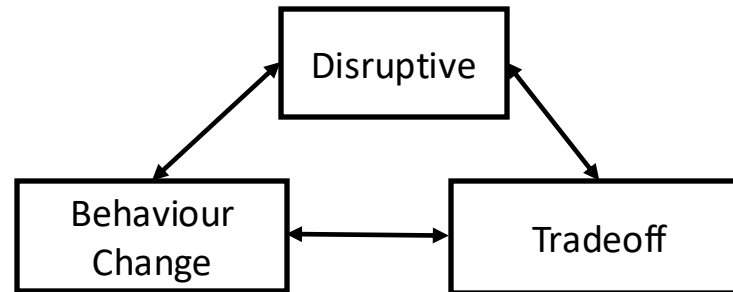


Computational Sustainability

Actors:

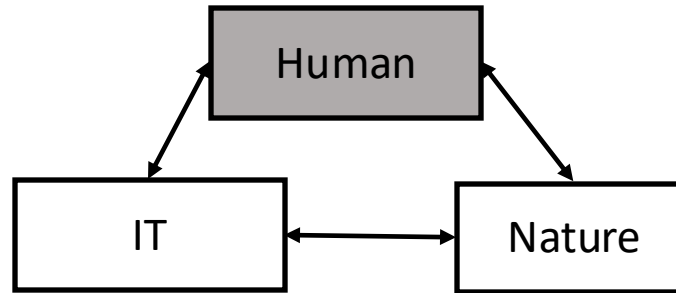


Methods:

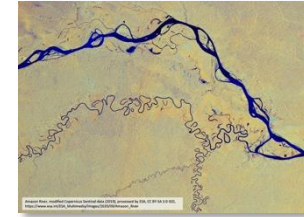


Computational Sustainability

Actors:



Sustainable Watershed Management Through IoT-Driven Artificial Intelligence

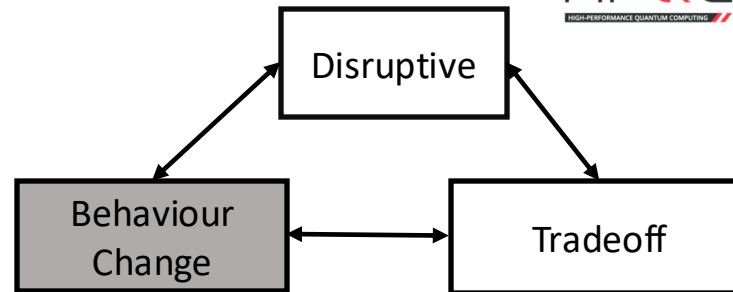


Virtual Shepherd

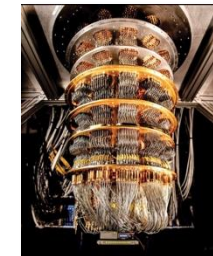


TRUST! Explaining the increasing energy consumption of IT infrastructures

Methods:



High Performance Integrated Quantum Computing



Quantum Algorithm Engineering COMET Module



Bundesministerium
Frauen, Wissenschaft
und Forschung



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AKADEMIE DER
WISSENSCHAFTEN



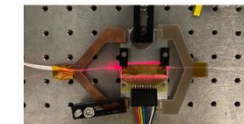
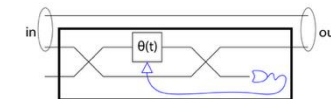
WWTF: AI-supported Holographic Environmental Water Monitoring



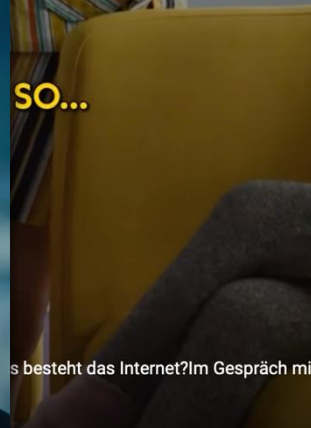
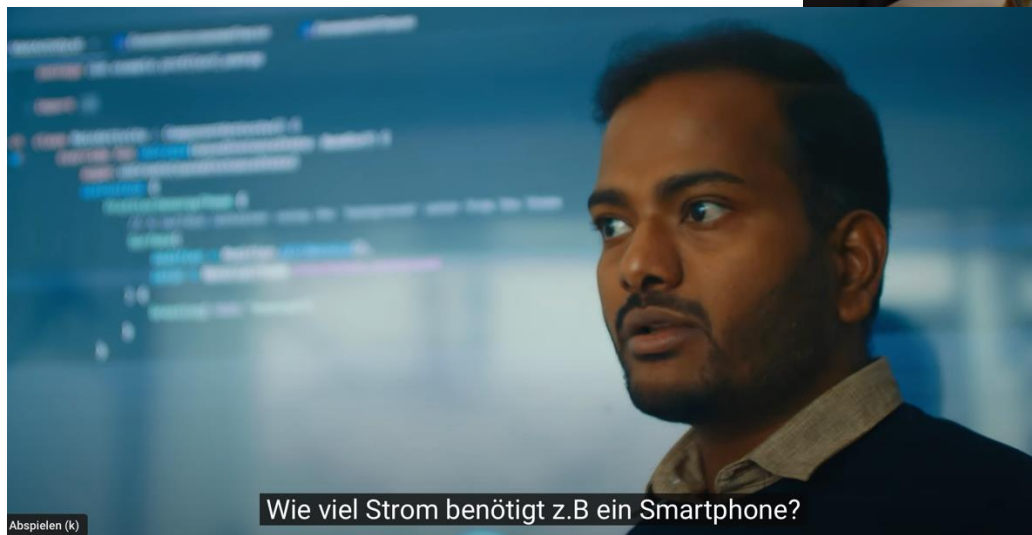
Themis: Trustworthy and Sustainable Code Offloading



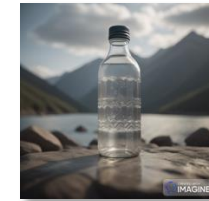
Triton: Transprecise Edge



WWTF: Advancing learning paradigms for photonic quantum processors



Asking ChatGPT Qs
with 5-50 prompts¹ ≈



500ml of water

Search on LLM vs
Google Search¹ ≈



2.9 Wh/ Search or
up to 20x
expensive

A text-to-image
generation² (1000
inferences) ≈



3 kWh or
3hr of
dishwasher

1 min video (e.g.,
midjourney) ≈



5,4 kWh or
36hr playing
Playstation 4.0

Costs of saying "thank you"

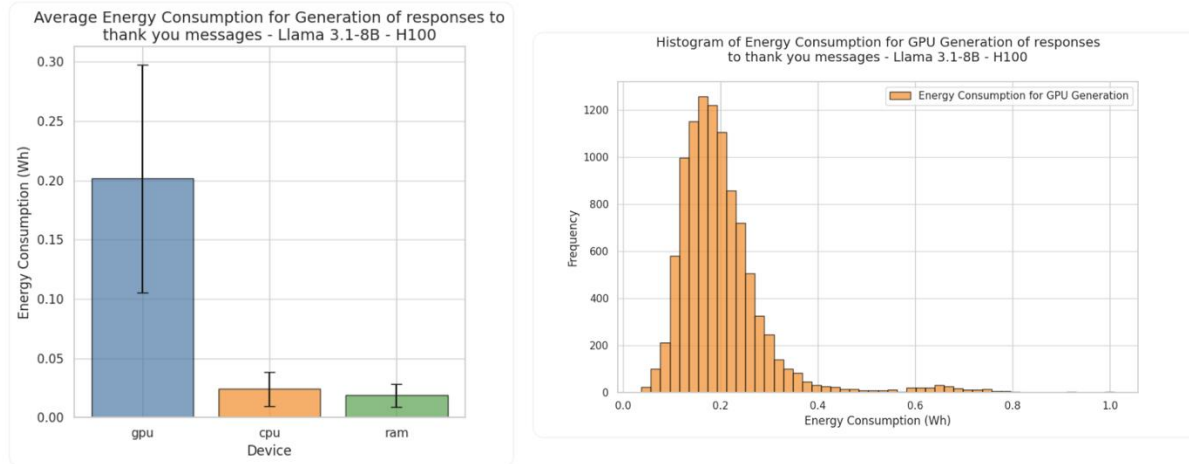


Figure 2: Left — Mean Energy consumption by device (GPU, CPU, RAM) for one response. Right — Histogram of GPU energy to respond to a thank you message.

Source: Saying Thank You to a LLM Isn't Free — Measuring the Energy Cost of Politeness
<https://huggingface.co/blog/jdelavande/thank-you-energy>



Wie viel kostet uns die Höflichkeit mit ChatGPT, ORF1 ZIB Interview.

Human-AI Interaction for Sustainability (HAI4S): Green Prompting Strategies and Eco-Feedback for Efficient Use of Large Language Models

Hosting Institution: LIT Robopsychology Lab **JKU** JKU

Supervisors:



Martina Mara

JKU JKU



Ivona Brandic

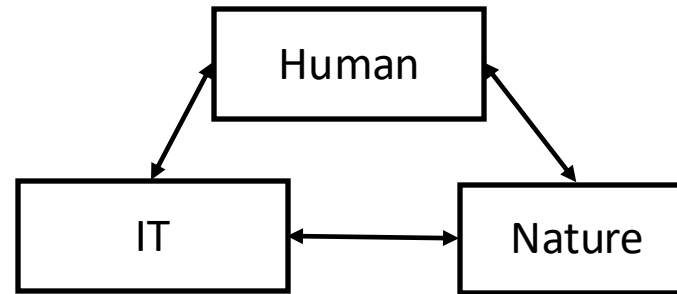
TU TU Wien



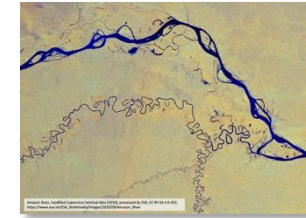
**Doctoral College
Computational Sustainability**

Computational Sustainability

Actors:



Sustainable Watershed Management Through IoT-Driven Artificial Intelligence

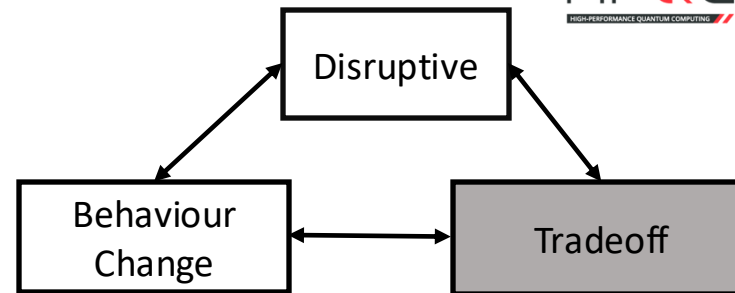


Virtual Shepherd

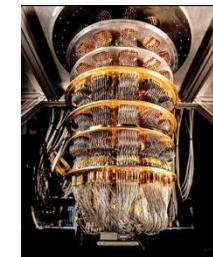


TRUST! Explaining the increasing energy consumption of IT infrastructures

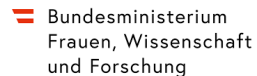
Methods:



High Performance Integrated Quantum Computing



Quantum Algorithm Engineering COMET Module



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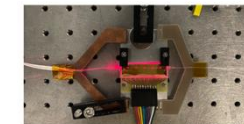
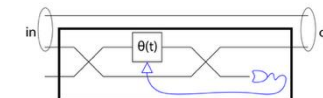
WWTF: AI-supported Holographic Environmental Water Monitoring



Themis: Trustworthy and Sustainable Code Offloading



Triton: Transprecise Edge

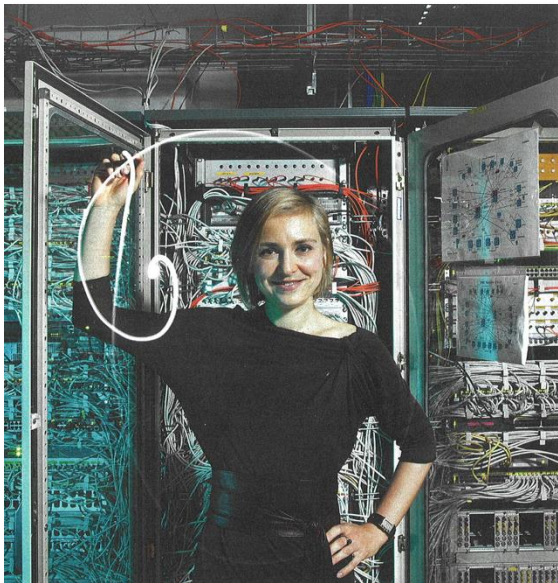


WWTF: Advancing learning paradigms for photonic quantum processors

Data Centers once...



INRIA Rennes, 2012



TU Wien (VCS), 2011 ("Der Falter")



**RÉPUBLIQUE
FRANÇAISE**

*Liberté
Égalité
Fraternité*



Institut de Recherche
en Informatique de Toulouse
CNRS - INP - UT3 - UT1 - UT2J



Leibniz-Rechenzentrum
der Bayerischen Akademie der Wissenschaften

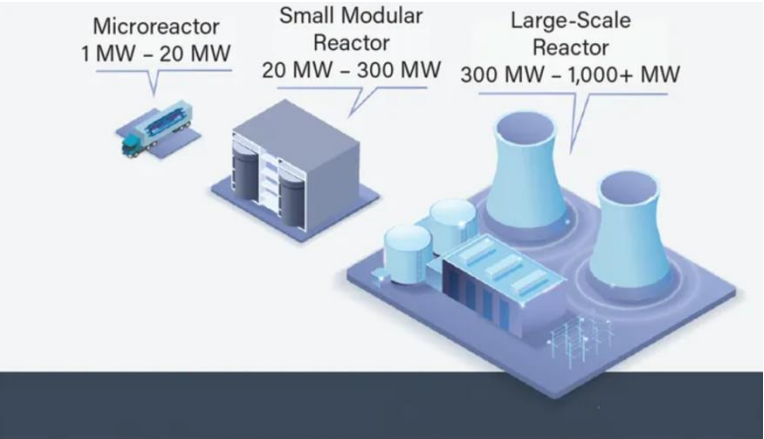
SuperMUC Munich, 2013



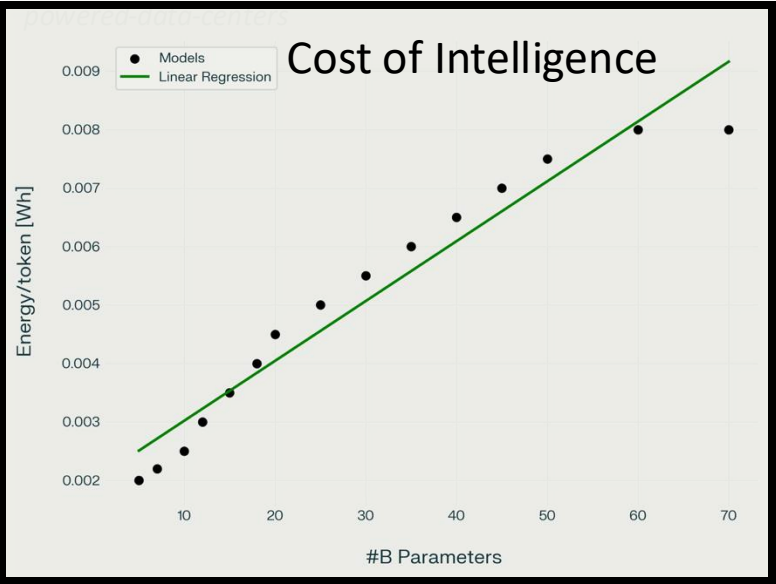
TU Wien, 2018



... Data Centers Now



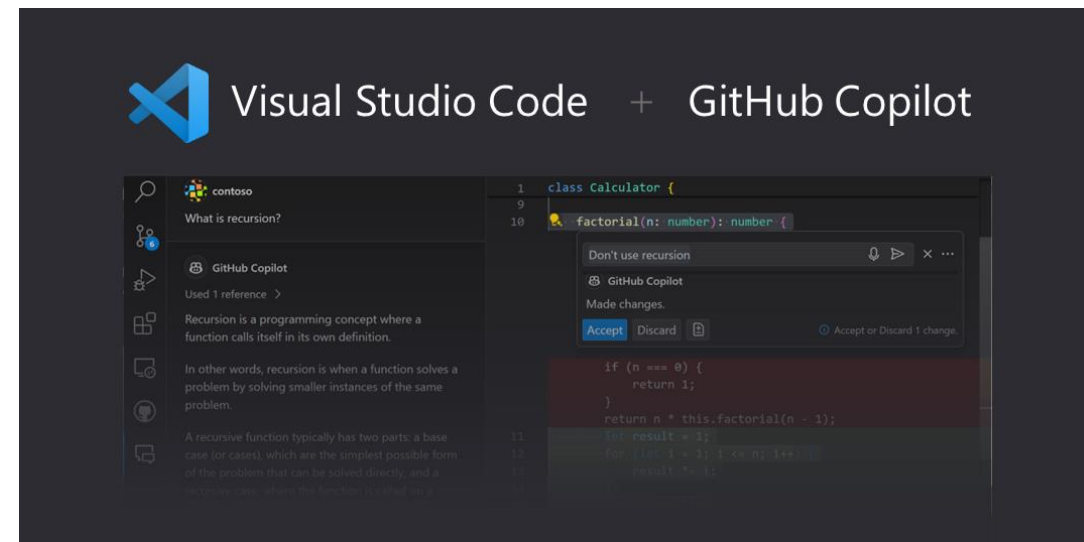
Source: [https://www.datacenterknowledge.com/energy-power-supply/going-nuclear-a-guide-to-smrs-and-nuclear-](https://www.datacenterknowledge.com/energy-power-supply/going-nuclear-a-guide-to-smrs-and-nuclear-powered-data-centers/)



Source: <https://www.datacenterdynamics.com/en/news/amazon-signs-deals-to-invest-in-nuclear-smrs-to-power-data-centers/>

Code generation

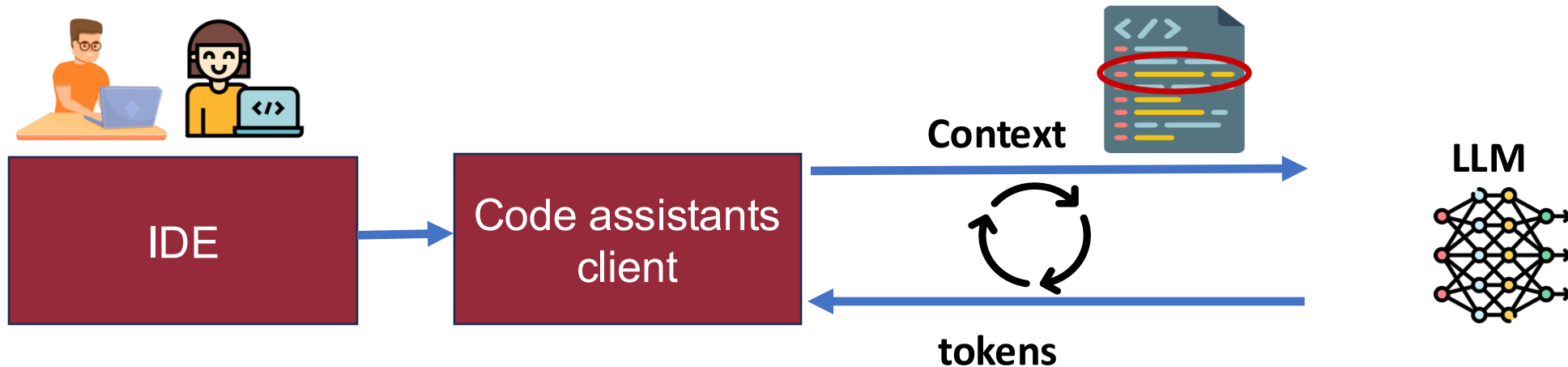
- AI-native IDEs are seeing rapid adoption, tailored for LLM-based code assistants.
- Code assistants are widely viewed as the future of software development.
- Studies report that 30–40% of code in large organizations is now AI-generated¹.



1. McKinsey & Company, *The Economic Potential of Generative AI*, 2023

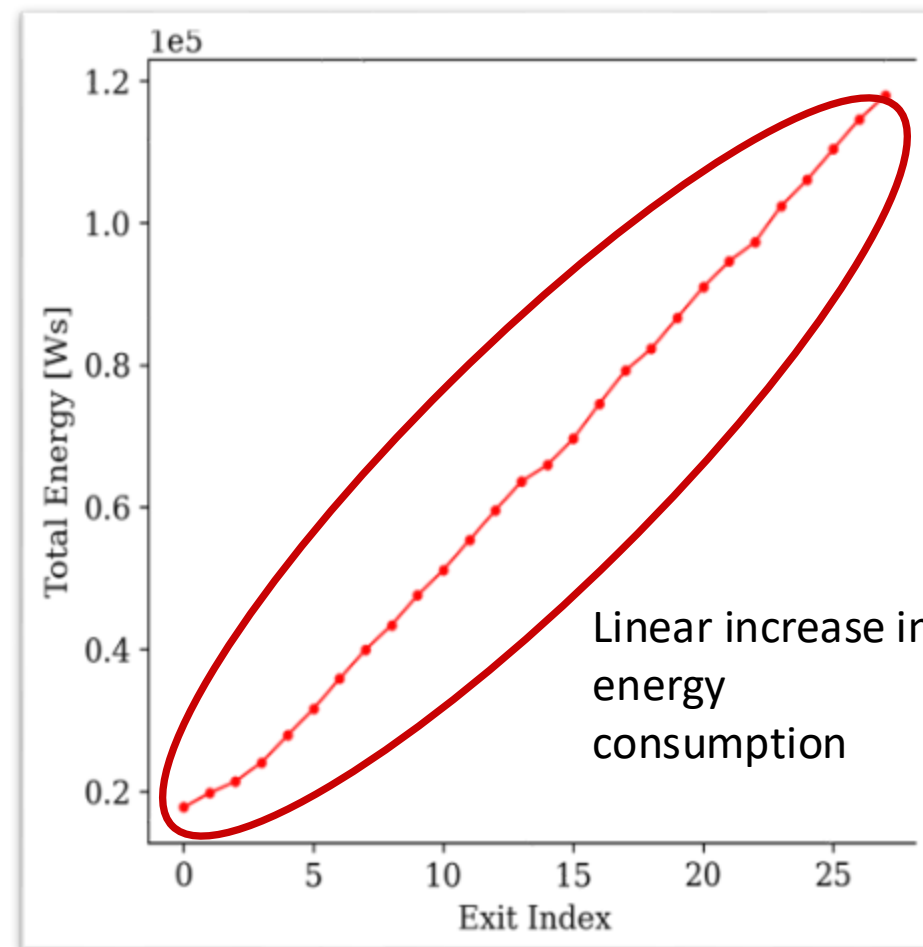
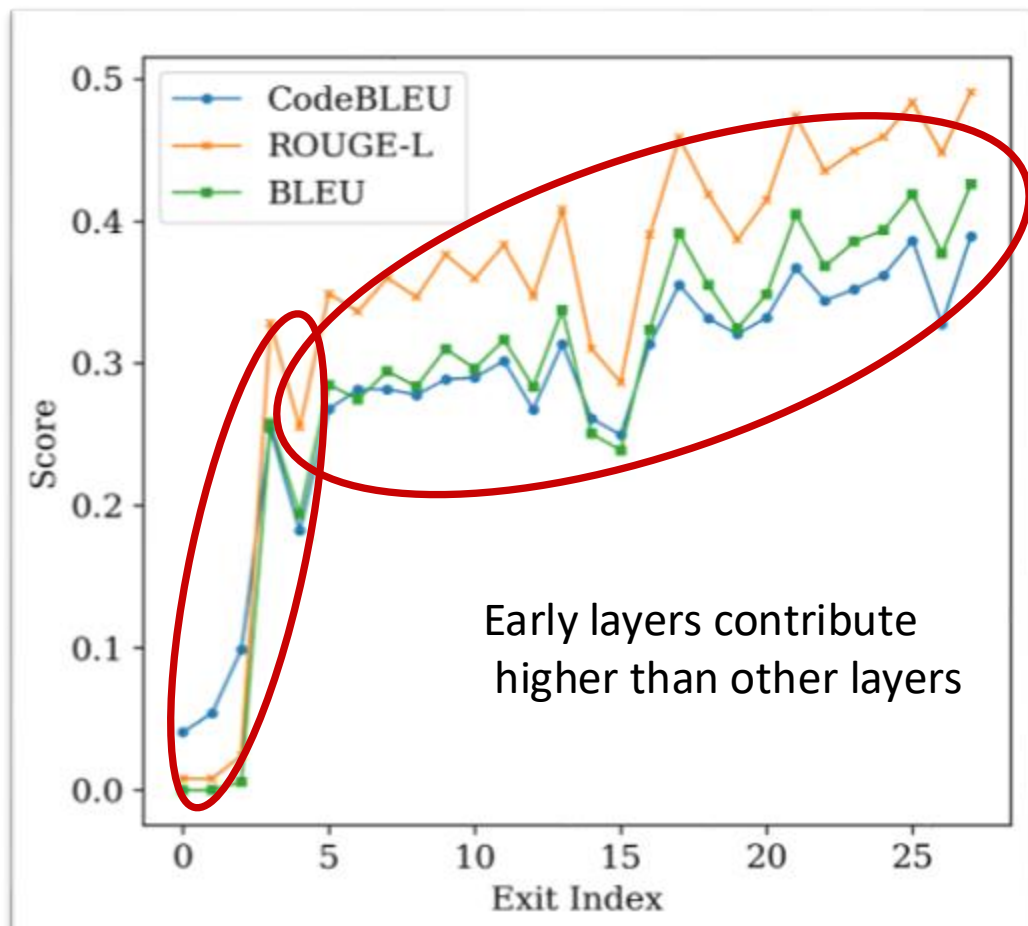
Resource-efficient inference in LLMs

- Knowledge Distillation
 - Quantization
 - Compression
- }
- Introduces loss of accuracy
 - Non-adaptivity in runtime
- Code generation is highly **dynamic**, with a different degree of complexity in each line of code
 - Need to understand the **input context** (code file) and choose the corresponding inference load dynamically



Slide: courtesy Shashi Ilager,
University of Amsterdam

Exploiting LLM's Linear Increase in Energy Consumption



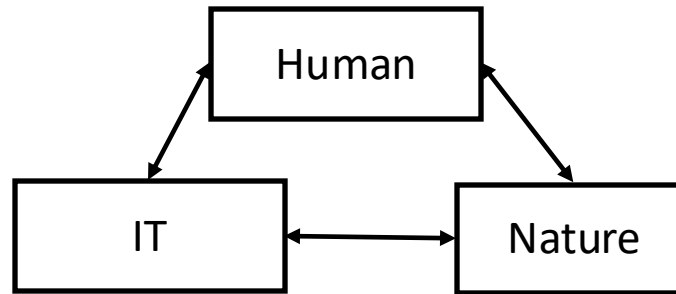
Performance vs energy-consumption Llama3.2-3B/Java, fixed exits

Triton: Transprecise Edge
Grant-DOI: 10.55776/
P36870

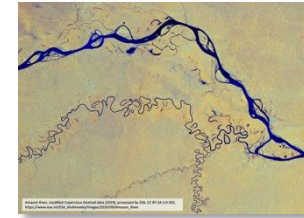
Themis: Trustworthy and
Sustainable Code
Offloading
Grant-DOI:
10.55776/PAT1668223

Computational Sustainability

Actors:



*Sustainable Watershed Management
Through IoT-Driven Artificial Intelligence*

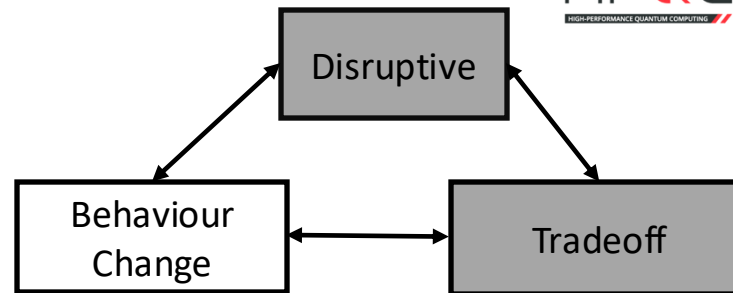


Virtual Shepherd

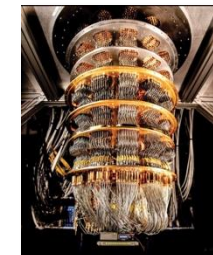


*TRUST! Explaining the
increasing energy
consumption of IT
infrastructures*

Methods:



*High Performance Integrated
Quantum Computing*



*Quantum Algorithm Engineering
COMET Module*

**Doctoral College
Computational Sustainability**

 **Bundesministerium
Frauen, Wissenschaft
und Forschung**

ÖAW

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WISSENSCHAFTEN

WWTF



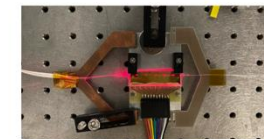
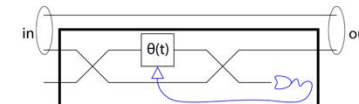
*WWTF: AI-supported Holographic Environmental
Water Monitoring*

FWF

*Themis: Trustworthy and
Sustainable Code Offload*

FWF

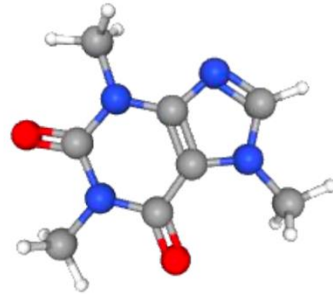
Triton: Transprecise Edge



WWTF

*WWTF: Advancing learning paradigms for
photonic quantum processors*

A cup of coffee?



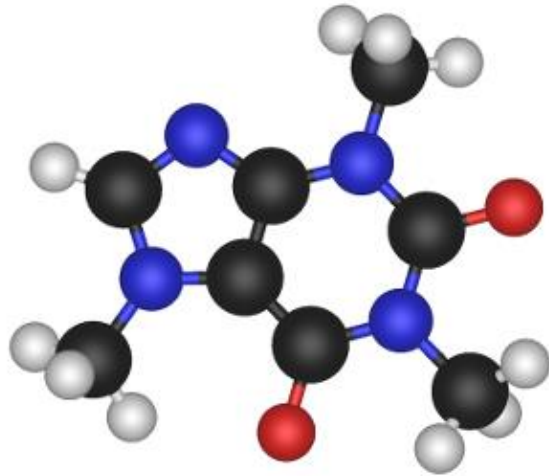
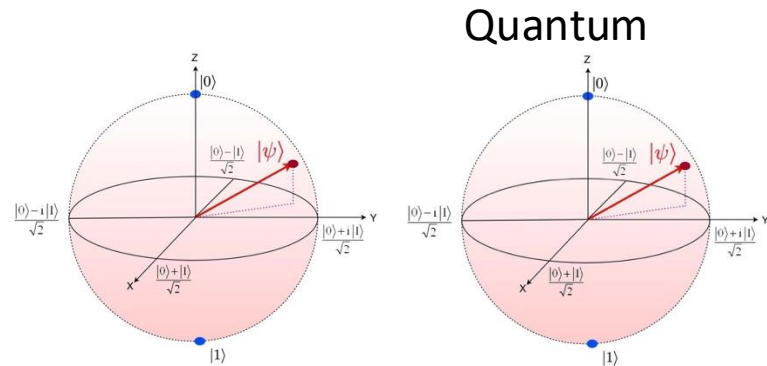
“Every time you add a qubit, you double your possible outcomes, With 20 qubits there are a million possible outcomes. With 100 qubits, you have more possibilities than there are bits in all the hard drives in the world. With 300 qubits—that’s more possibilities than there are particles in the universe.”

- Representing the energy configuration of a single caffeine molecule at a single instant requires approximately 10^{48} bits in a classical computer
- Can be done using 160 logical qubits on a quantum machine

1 000 000 000 000 000 000 000 000 000 000 000 000 000 000 000 000 vs. 160

Beyond 0 and 1

Von Neumann



Bottom up approach

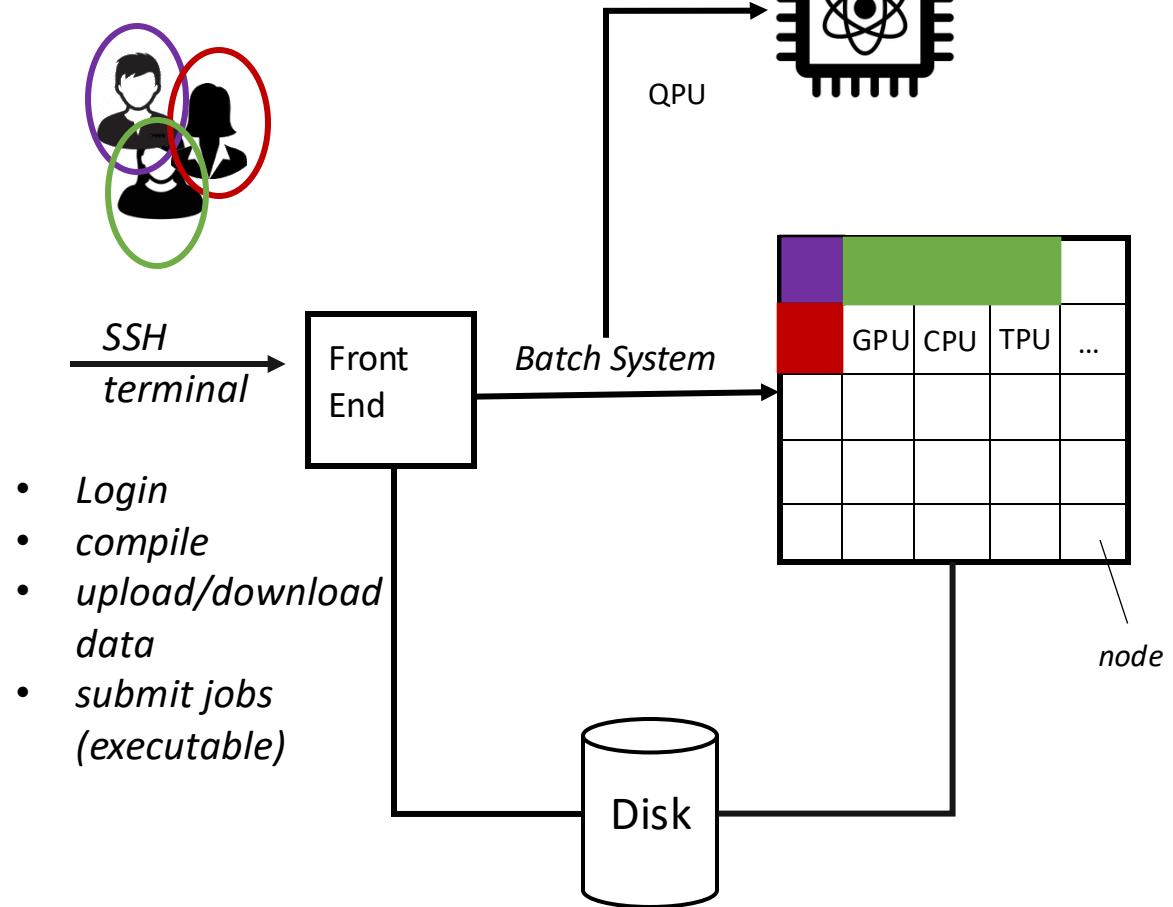
- Variational Quantum Linear Solver (VQLS)
- Quantum Eigenvalues → **Native 3d modeling of scientific applications**

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

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 many users



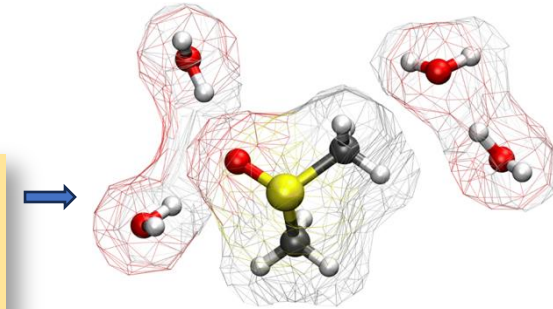
HPQC Cluster

FFG High Performance Integrated
Quantum Computing
Grant ID: #4562957

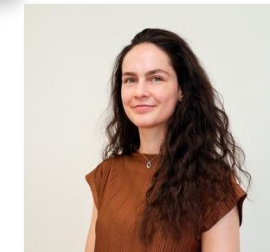


- Improve QC performance
- Complement QEC capabilities
- Extend QC research with HPC

- Implement Hybrid Q-Libs
- **Benchmark PoC use-cases in hybrid infrastructure (HPC TU Wien)**



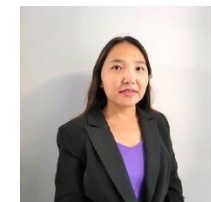
QML



S. Herbst



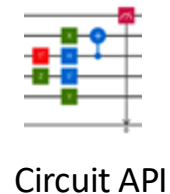
V. De Maio



M. Kanatbekova



F. Zilk



```
"string": "Hi",
"number": 2.5,
"boolean": true,
"null": null,
"object": { "name": "Kyle", "age": 24 },
"array": ["hello", 5, false, null, { "key":
  { "name": "Jerry", "age": 20 },
  { "name": "Sally", "age": 26 }
}]
```

Offloading API

QC system

Direct access

Login console

HPC node



aconet



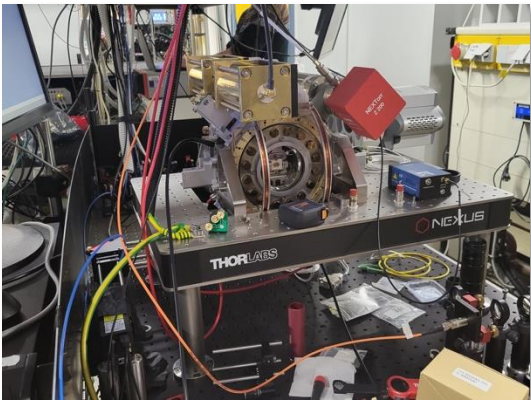
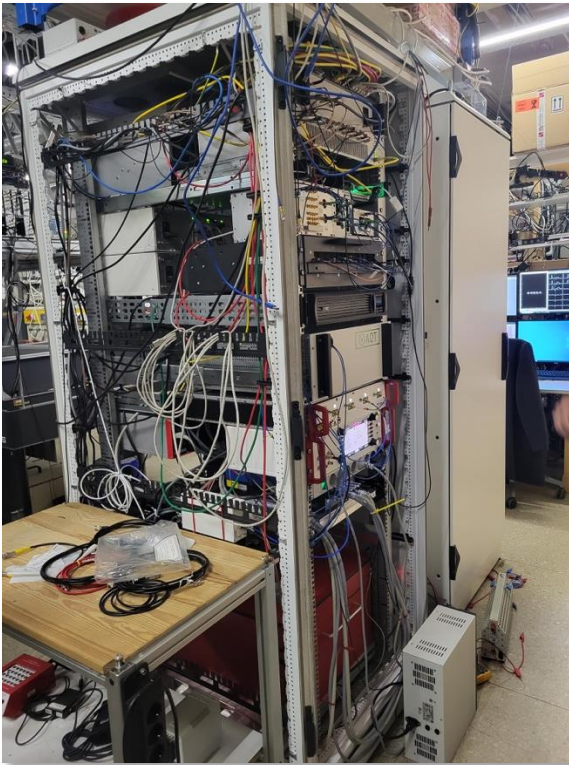
- Best Paper Award at ISVLSI 2025
- FGCS2024 Fall Editor's Choice Award
- Dieberger-Skalicky Prize 2025
- NetIdee Scholarship 2024
- DISCOVER US Scholarship 2026



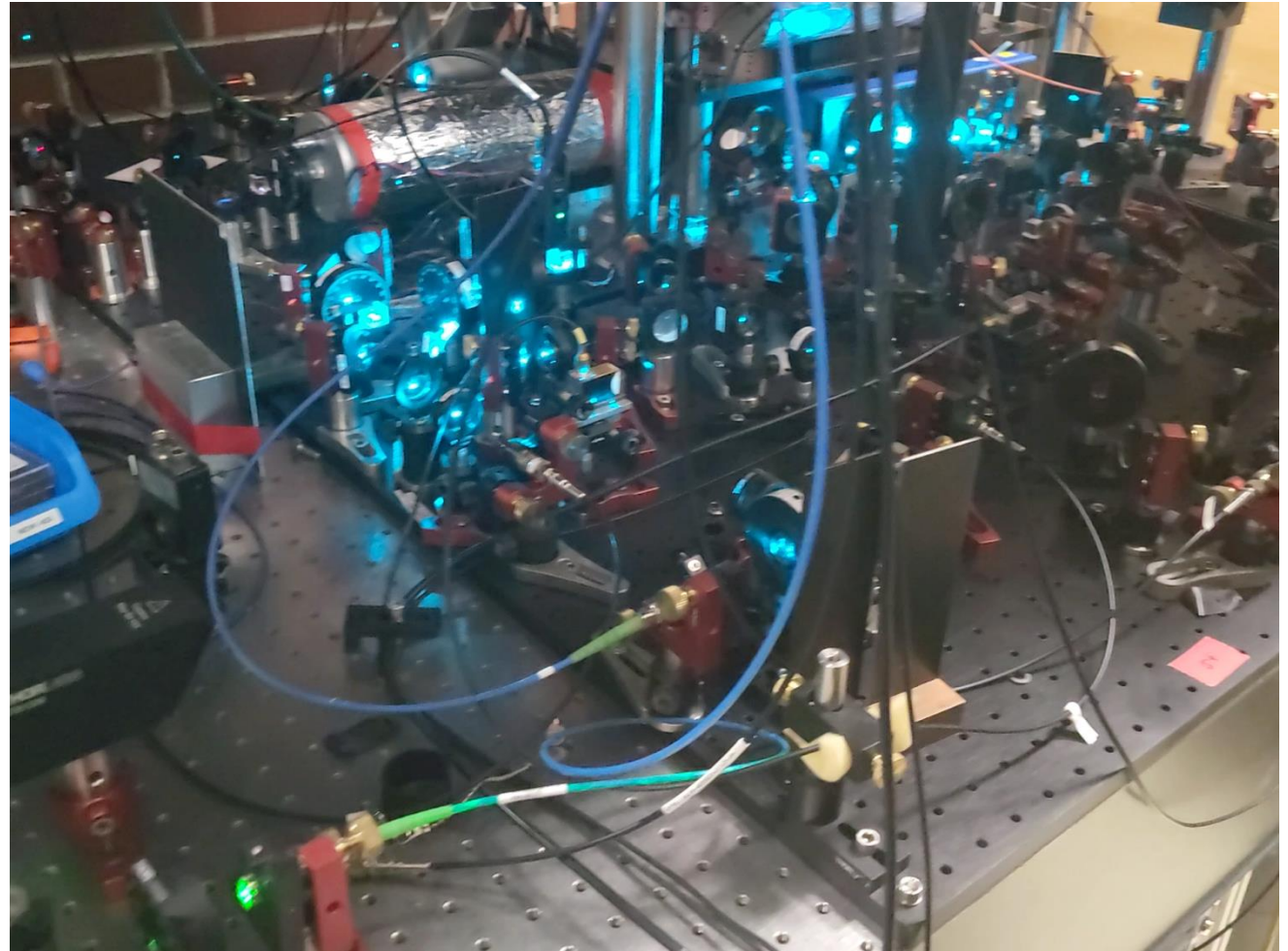
Assembly of a Quantum Processing Unit (QPU)



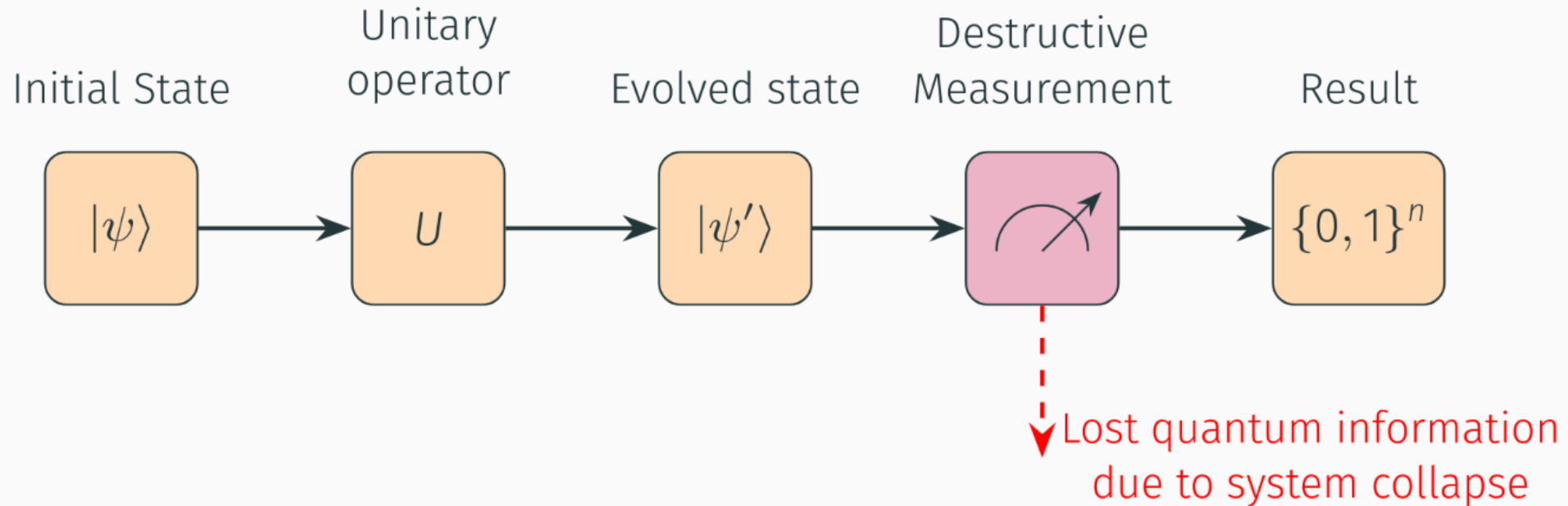
- Move from Ca+ to Ba+
- New system with
stage 1: 10x higher T_1
stage 2: infinit. Higher T_1
- 2q error rate:
legacy: $< 10^{-2}$
target: $< 10^{-3}$
- Init error
legacy: $< 10^{-3}$
target: $< 10^{-4}$
- Readout error
legacy: $< 10^{-3} \rightarrow \sim 10^{-4}$
target: $< 10^{-5}$



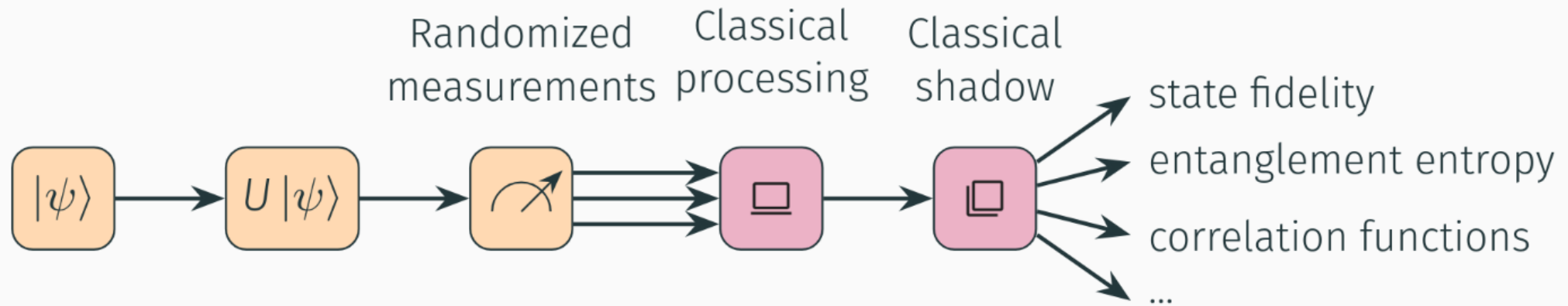
Ion-trap
quantum
computer



Destructive Measurements



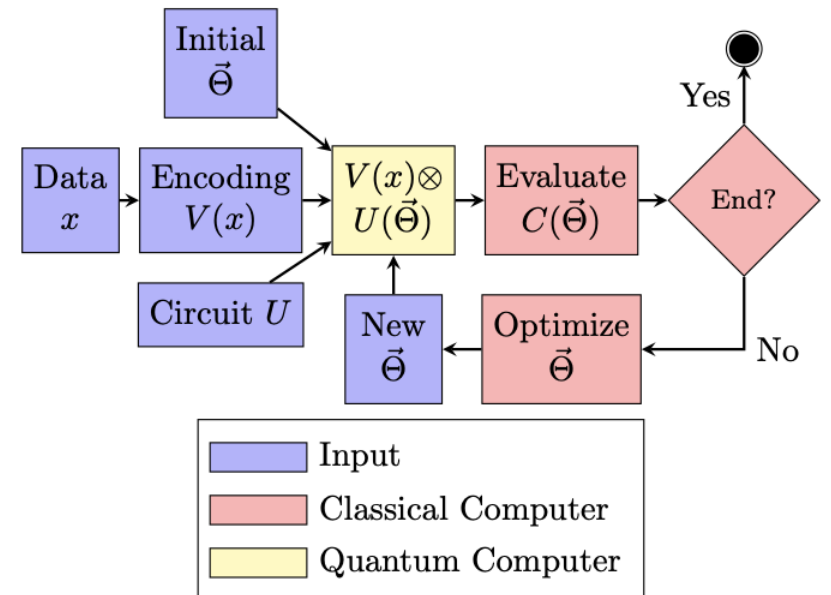
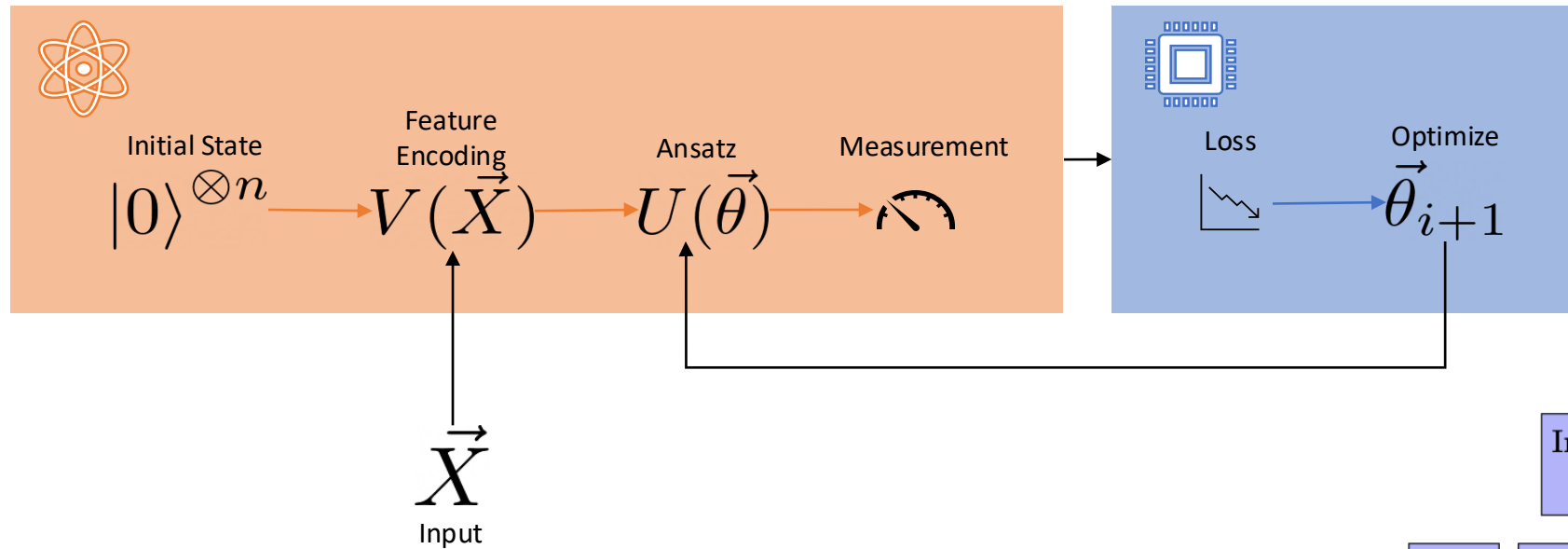
Classical Shadows



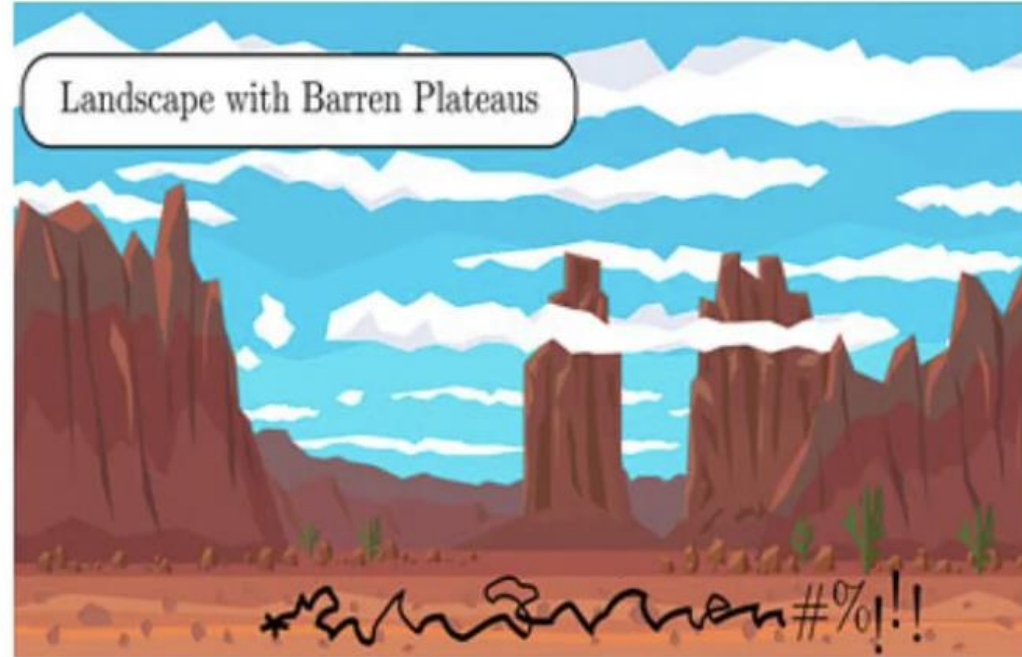
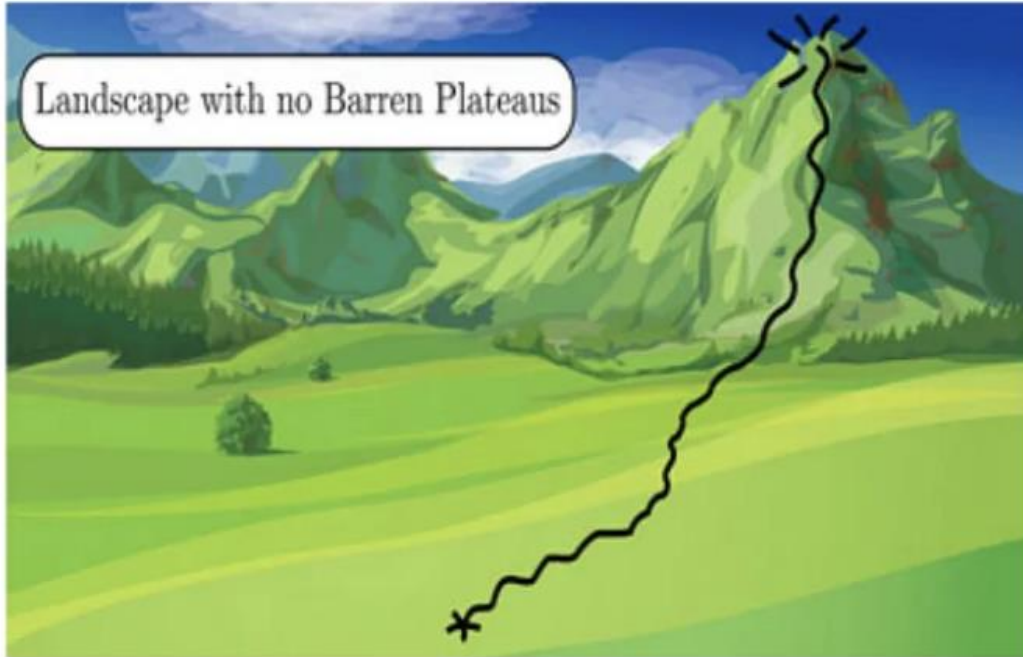
Source: Huang, H.Y., Kueng, R. & Preskill, J. Predicting many properties of a quantum system from very few measurements. *Nat. Phys.* 16, 1050–1057 (2020).

Slide courtesy: S. Herbst

Quantum Neural Network (QNN)



Barren Plateau



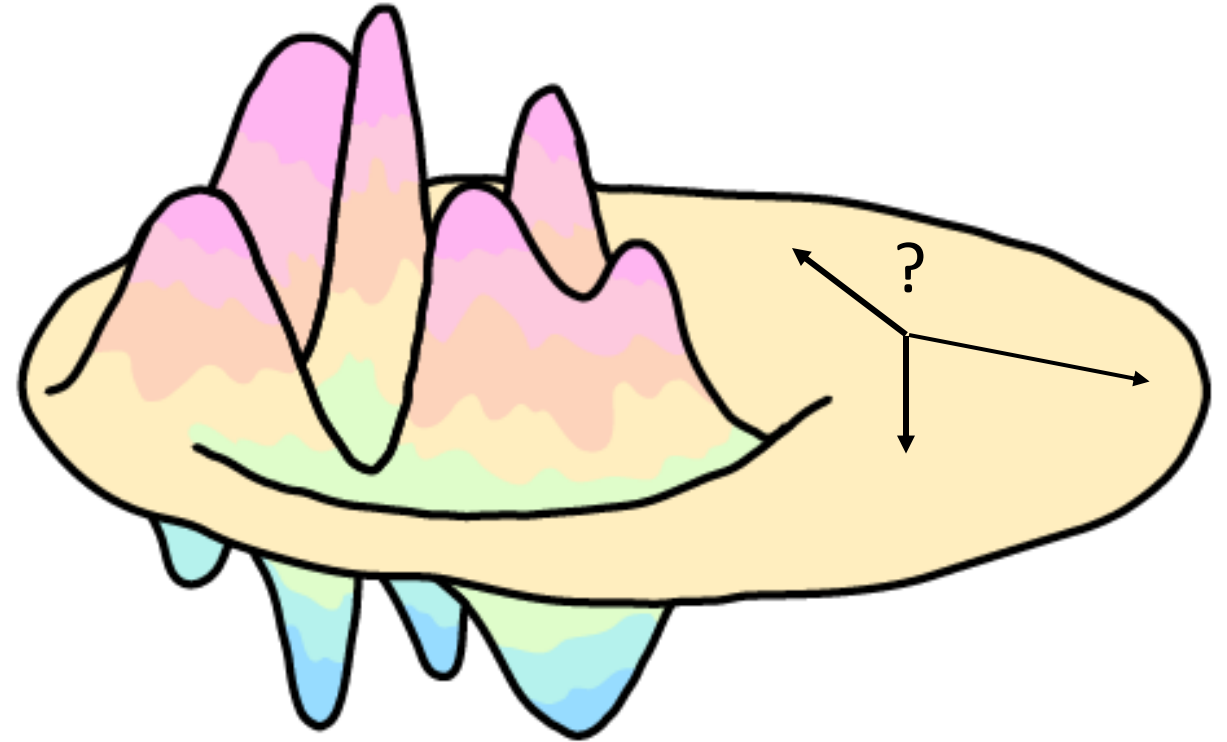
Source: <https://medium.com/@deltorobarba/can-quantum-computing-accelerate-generative-ai-b1df36398ac5>

optimization landscape

- gradient of the cost function with respect to the circuit parameters becomes **exponentially small** as the system size increases.
- **training the model extremely difficult**, as gradient-based optimization essentially stalls—similar to being stuck on a flat, featureless desert with no clear direction to go.

Implications for QML

- **Training fails:** Gradient descent becomes ineffective, especially as the number of qubits increases.
- **Wasted resources:** Running quantum circuits that don't contribute to learning wastes both time and quantum compute resources.
- **Scalability issues:** Makes it hard to train larger QML models.



Unveiling the Resource Footprint of Hybrid Classic-Quantum Systems

Hosting Institution: *Institute of Computer Engineering*  TU Wien

Supervisors:



Ivona Brandic

 TU Wien



Richard Kueng

 JKU



Vincenzo De Maio

 TU Wien

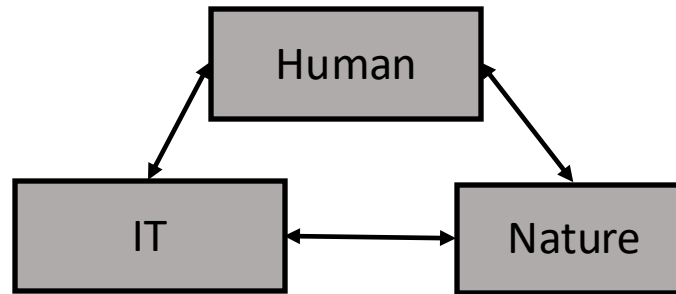
 Uni. Leicester



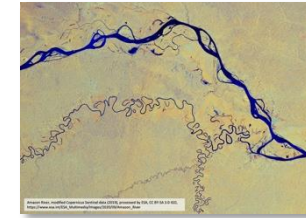
**Doctoral College
Computational Sustainability**

Computational Sustainability

Actors:



Sustainable Watershed Management Through IoT-Driven Artificial Intelligence

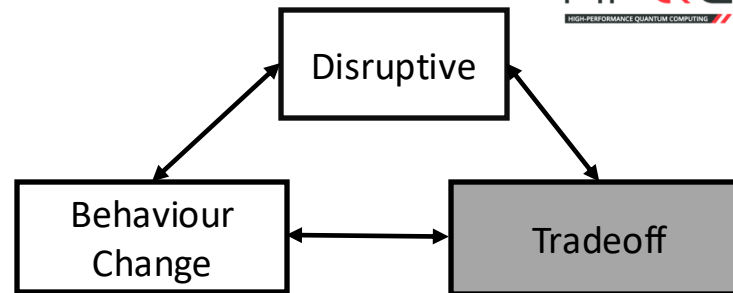


Virtual Shepherd

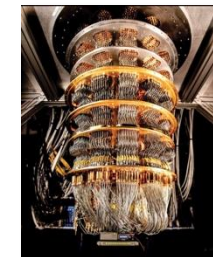


TRUST! Explaining the increasing energy consumption of IT infrastructures

Methods:



High Performance Integrated Quantum Computing



Quantum Algorithm Engineering COMET Module

Doctoral College Computational Sustainability

ÖAW

ÖSTERREICHISCHE
AKADEMIE DER
WISSENSCHAFTEN



WWTF: AI-supported Holographic Environmental Water Monitoring

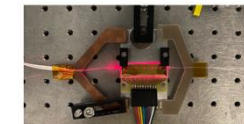
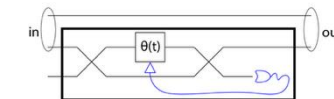


FWF

Themis: Trustworthy and Sustainable Code Offloading

FWF

Triton: Transprecise Edge



WWTF: Advancing learning paradigms for photonic quantum processors

Siberia's gateway to the underworld



Source: <https://www.science.org/content/article/siberia-s-gateway-underworld-grows-record-heat-wave-thaws-permafrost>

Batagaika crater



Ambarnaya river, Norilsk, Siberia

Source: <https://www.brusselstimes.com/news/115566/u-s-offers-to-help-russia-following-arctic-oil-spill/>



universität
wien



SYKE

Finnish Environment Institute

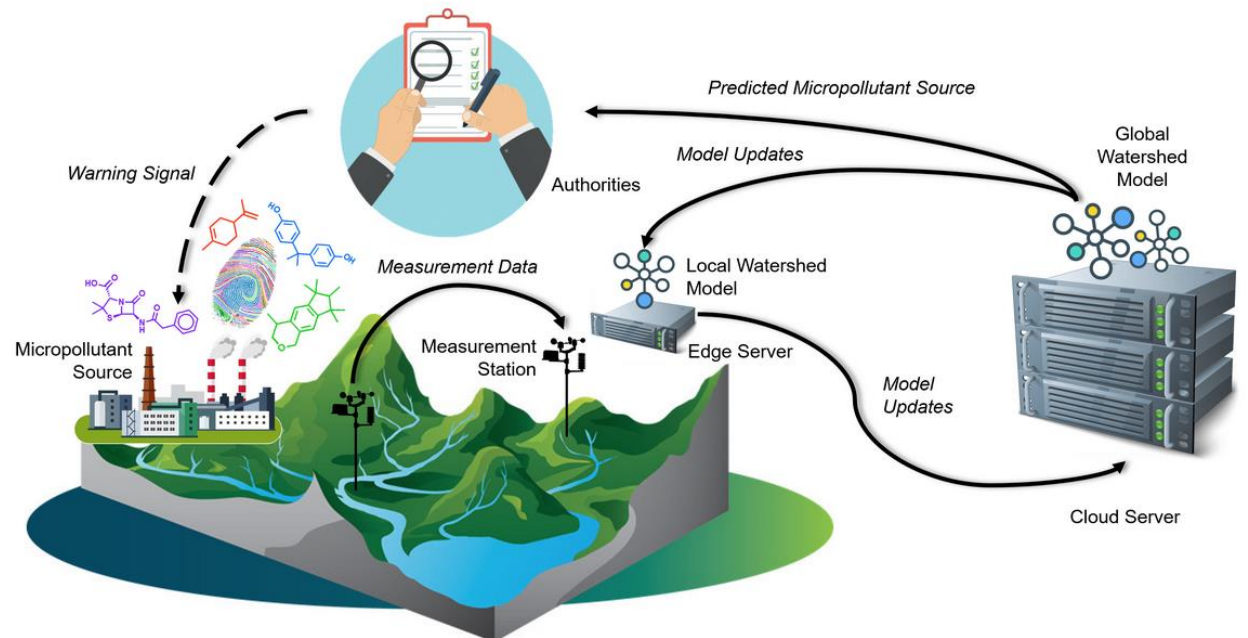
Source: S. Ahmad, H. Uyanik, T. Ovatman, M. T. Sandikkaya, V. De Maio, I. Brandic, A. Aral. Sustainable Environmental Monitoring via Energy and Information Efficient Multi-Node Placement. IEEE Internet of Things Journal

SWAIN - Sustainable Watershed Management Through IoT-Driven Artificial Intelligence

chist-era
CHIST-ERA-19-CES-005

FWF

- In-silico digital representation of the river
- viscoelasticity
- **Data from IoT sources + expert knowledge = digital model of the river**

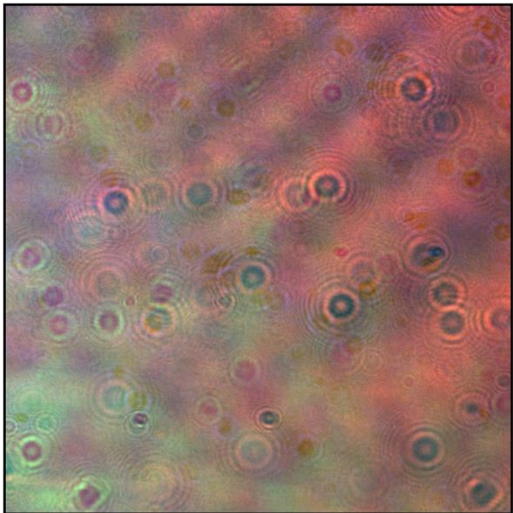
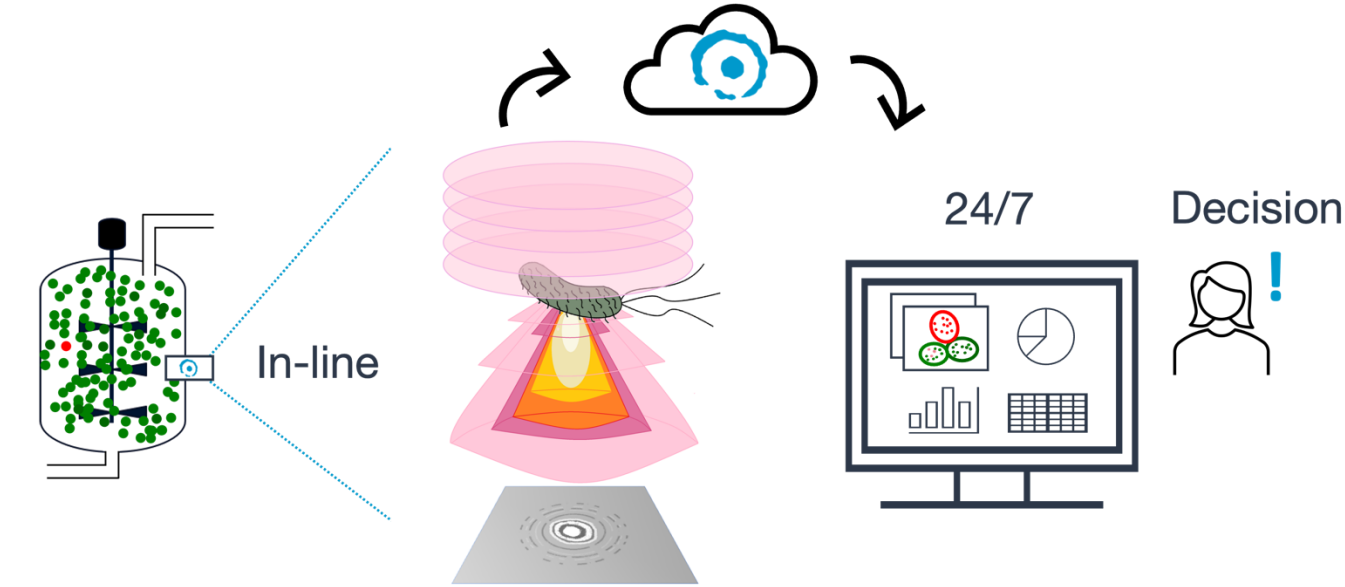




Wasserversorgung der Stadt Wien



Holographic microscopy



ESR24-053
AI-supported Holographic
Environmental Water
Monitoring

- Our device has the size of a shoebox
- It collects liquids and shines light through them
- Some light is scattered by any particles inside
- The illuminating and scattered light interfere forming holograms that are recorded by a camera
- The preprocessed images are sent to the cloud
- The light in the volume is numerically reconstructed to ~500 images in a stack
- The total volume of view is ~1M times that of a conventional microscope
- Proprietary algorithms boost the contrast and resolution
- The whole platform provides real time microscopic monitoring

Slide: courtesy Peter van Oostrum, Holooid, BOKU Wien



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Commission

Horizon 2020
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 Bundesministerium
Frauen, Wissenschaft
und Forschung

