

# Can generative AI run on thin air?

Generative AI impact on data centers

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# RISE ICE Data Center

A full-scale research datacenter and test environment with the objective to increase knowledge and strengthen the global AI & Data Center ecosystems

2000 physical servers  
250 kW  
10 petabyte storage  
240 GPUs  
12,5 petaflops  
HDFS clusters  
OpenStack ECC  
Kubernetes cluster  
OCP servers  
Edge modules

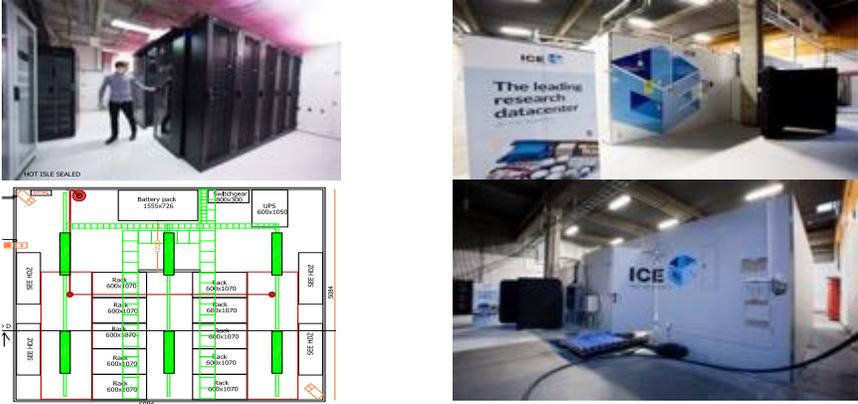
- 30+ projects, from the ground to the cloud
- 20+ employees
- 4+ MEUR turnover
- Established 2016
- 100+ project partners

Partners: Ericsson, ABB, Vattenfall, Meta, LTU, Region North, Vertiv, BP Castrol, Intel



# RISE ICE Datacenter test environment

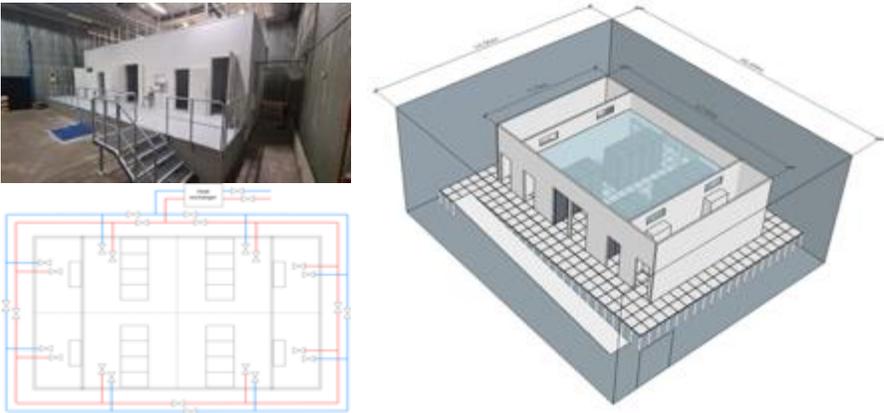
Module 1 & 2 Compute clusters



Module 3 OCP Lab + Climate box + Heat box



Module 4 Facility hardware test lab

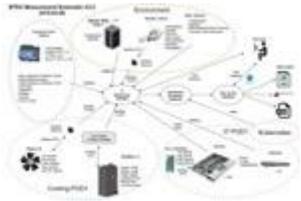


Edge + wind tunnels + liquid cooling test bed



# RISE ICE Datacenter public funded projects (examples)

BTDC



An EU project. The goal was the Most Energy and Cost efficient data center in the world. Holistic cooling was used to reach PUE = 1,015.

ArctiqDC



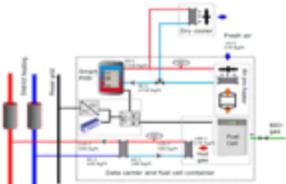
An EU Interreg Nord project. The goal was to study challenges with data centers in the arctic region. Show solutions for heat reuse & operations.

AutoDC



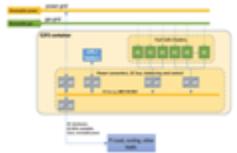
An ITEA3 project. The goal was lower OPEX by making the data center module autonomous, self-healing, self-optimizing and robust using AI.

Wedistrict



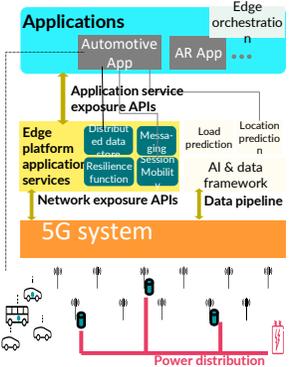
An EU project. The goal is to show a recovery process using liquid cooling and a fuel cell and use the electricity from the fuel cell for the datacenter

E2P2



An EU project. The goal is to define the fuel cell prime power concept for EU data centres. Demonstrate and validate a proof of concept.

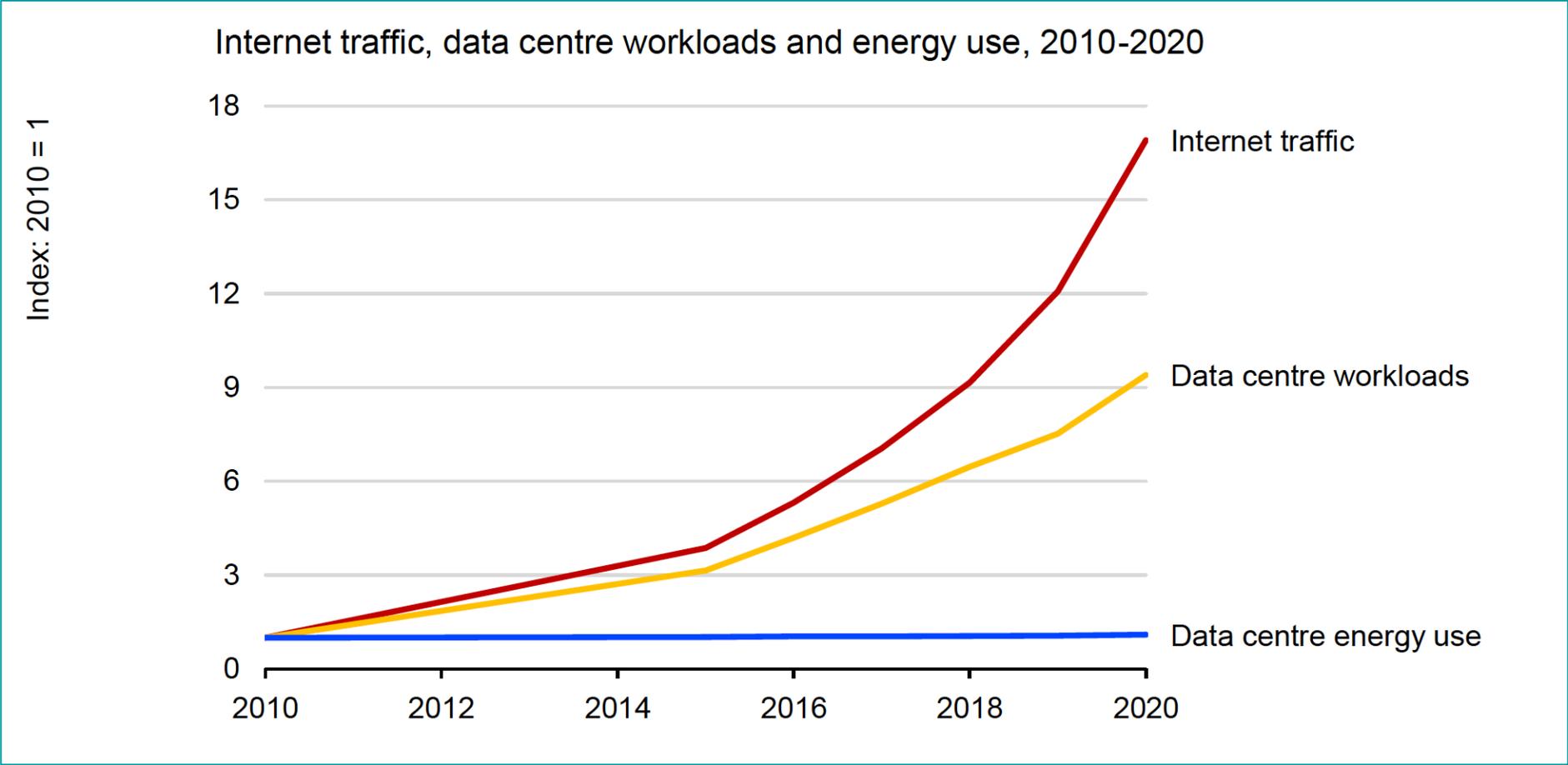
Aniara



A Celtic+ project. The goal is to develop automation of a network of edge nodes availability, robustness, optimization and reliability.

# Use of GPT-4 running on XPU

# Global data center energy use trends

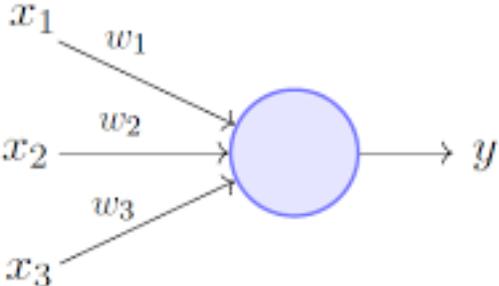


Data center energy use almost flat compared to internet traffic or data center workloads

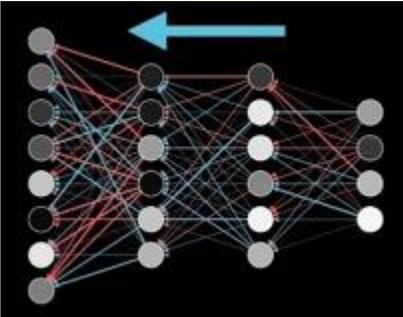


Source: Global data center energy use trends (IEA, 2021)

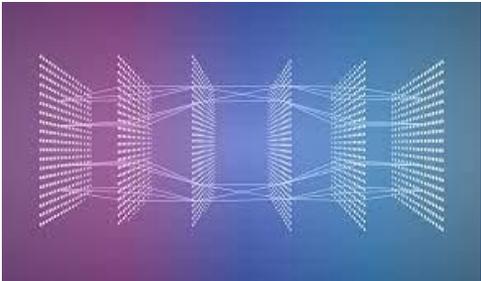
# History of AI by @Torshammer



Preceptron 1950-60  
(one layer, few parameters)



Backpropagation 1980-1990  
(many layers, many parameters)



Deeplearning 2010-20  
(many layers, billion parameters)



Simple AI - Many humans  
Few control parametr  
Laptop computer



Machine learning - Few experts  
Many control parameters  
Advanced GPU server



Generative AI - No single human (not all tasks)  
Billion control parameters  
Super computer cluster



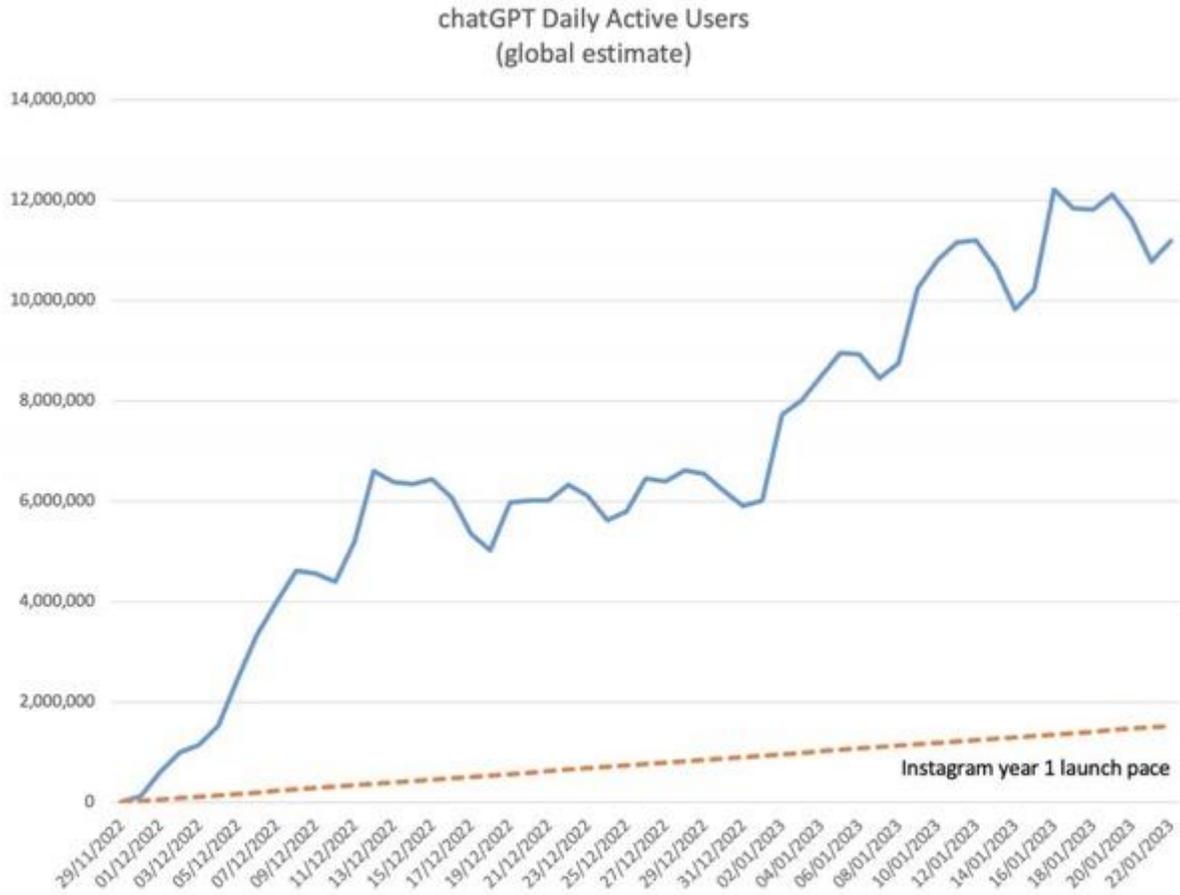
# Generative AI (for ex. ChatGPT)

 D Tell me what you can do.

 As an AI language model, I have a wide range of capabilities, including:  

1. Language translation: I can translate text from one language to another.
2. Text summarization: I can summarize long pieces of text into shorter, more digestible summaries.
3. Text completion: I can generate text based on a given prompt or context.
4. Question answering: I can answer questions based on information available in my database.
5. Sentiment analysis: I can analyze the sentiment of a piece of text and determine whether it is positive, negative, or neutral.
6. Language generation: I can generate text in a specific style, tone, or voice.
7. Chatbot capabilities: I can engage in a conversation with a user, answering questions, providing information, or just having a chat.
8. Content creation: I can help generate content for a wide range of purposes, such as writing articles, social media posts, product descriptions, and more.
9. Personalization: I can personalize my responses based on user preferences, interests, and history.
10. Data analysis: I can analyze data and generate insights based on the patterns and trends I detect.

# Growth rate ChatGPT

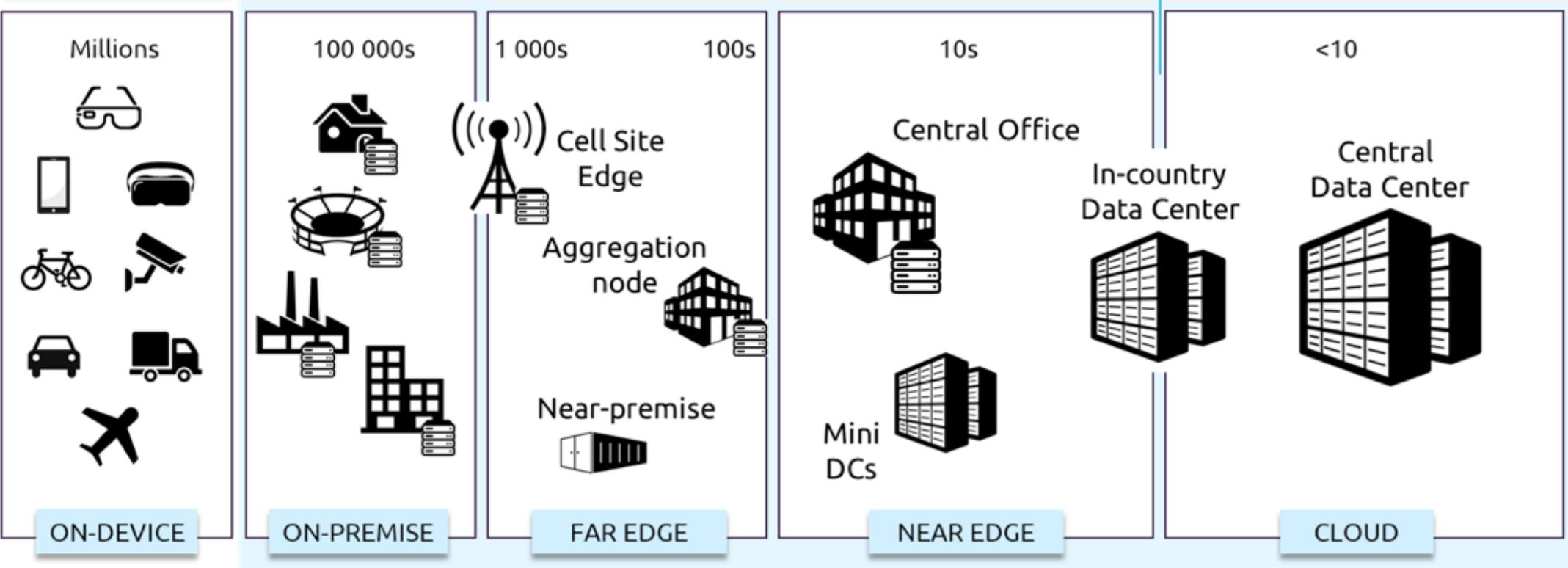


Source: similarWeb, ARK Invest



# High-end efficient and heavy AI compute

Where will it happen?



Inference (use phase)

Model learning (training phase)

# Training GPT-4

1,7 trillion model parameters & 13T tokens for training and 6 Floating operations (FLOPs) per parameter for one token

⇒ 133G petaFLOPs

OpenAI trained on 25 000 NVIDIA A100 GPUs

⇒ Training time >100 days

⇒ Cost \$100 million

⇒ Energy 50 GWh (2% of Swedish DC energy)

peta =  $10^{15}$  = 1 000 000 000 000 000



# Using (inference) GPT-4

1,7 trillion model parameters & max 32k tokens and 2 Floating operations (FLOPs) per parameter for one token (assume *1k* tokens/request)

⇒ 3,4 petaFLOPs per request -> 1,2 Wh

GPT-3 – 0,13 Wh, Google search – 0,28 Wh

OpenAI use 128 NVIDIA A100 GPUs for inference  
100 million active users and *2 request/day*

⇒ 0,68G petaFLOPs per day -> 91 GWh/year  
(10MW)

⇒ 0,1 seconds and 200 clusters for GPT-4



1 Wh = 60 W light bulb on for 1 minute.

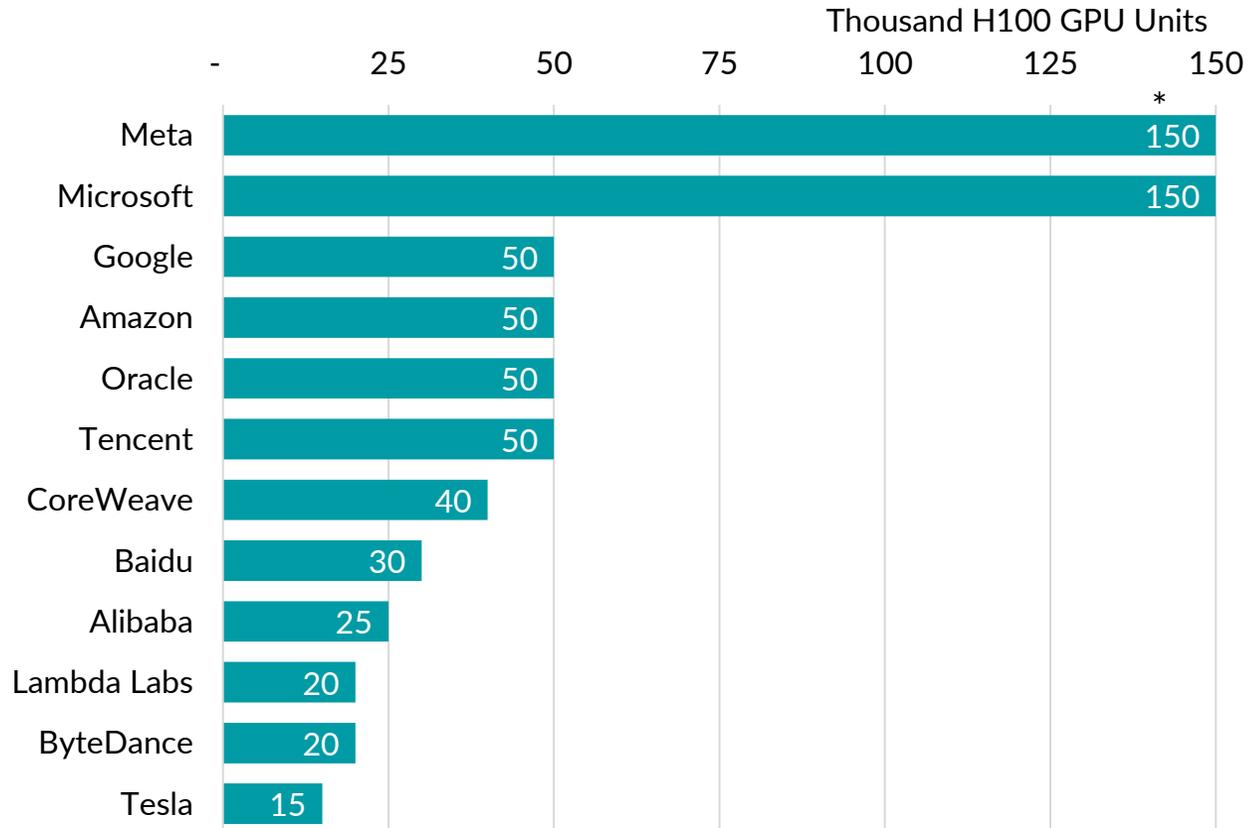


GPT-3 - 0,07G petaFLOPs/day  
and 9,5 GWh/year (1 MW)



# AI capacity buildout continues

H100 GPU orders expected to be fulfilled in 2023

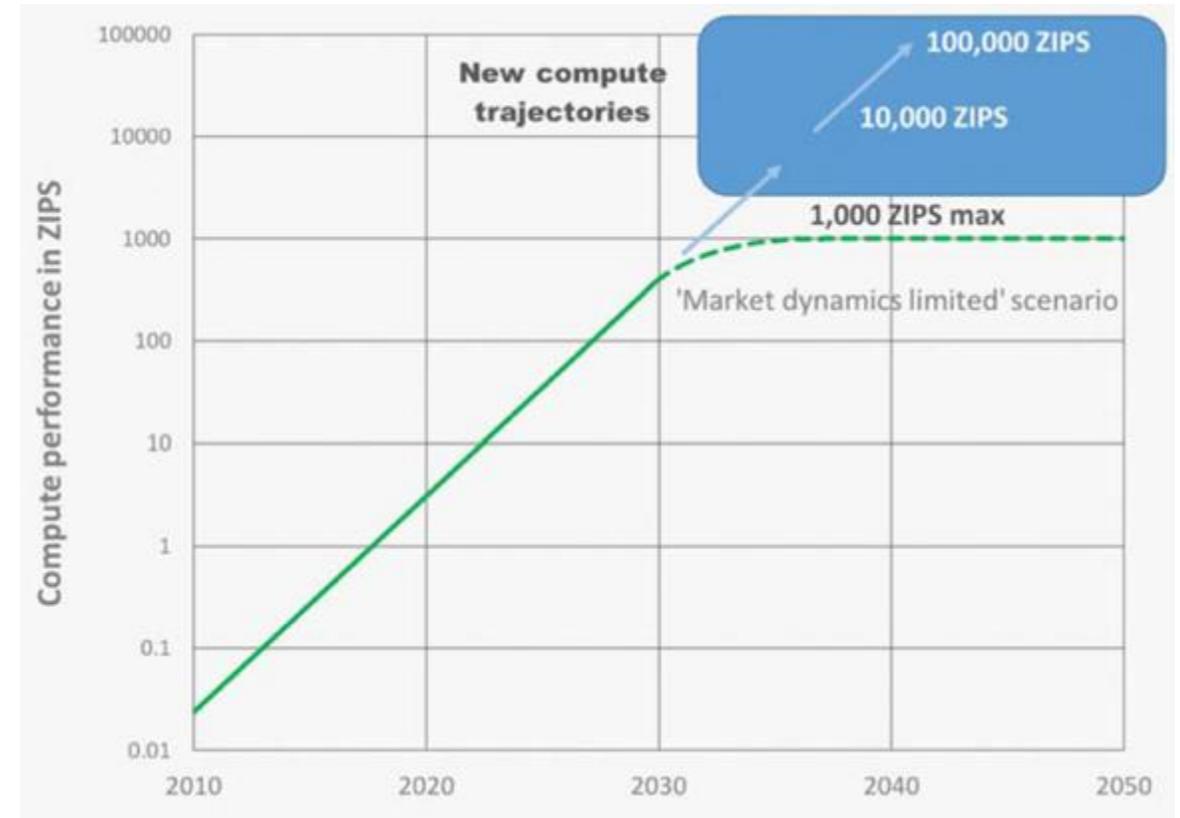
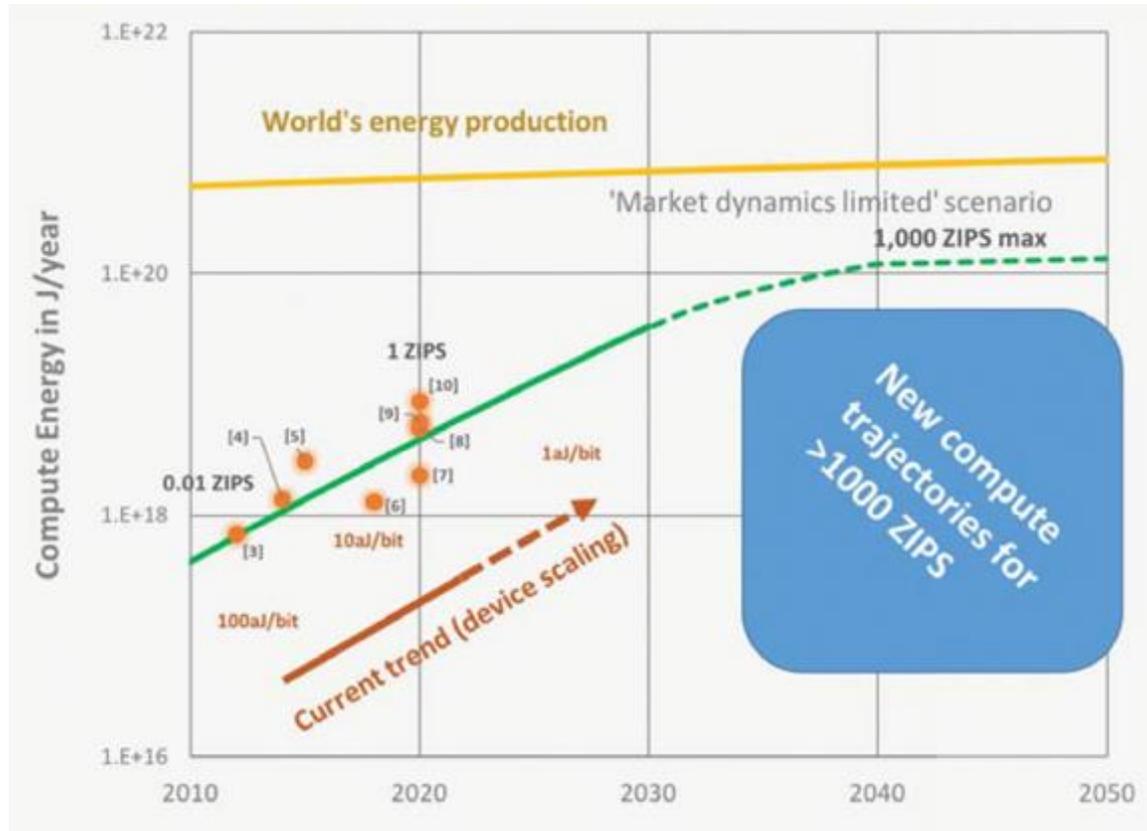


\*Some Meta order fulfilment might slip into 2024 due to a delay in data center go-live dates.

NVIDIA shipped nearly half a million H100 and A100 GPUs in 3Q23.

# Thermal management in future XPU

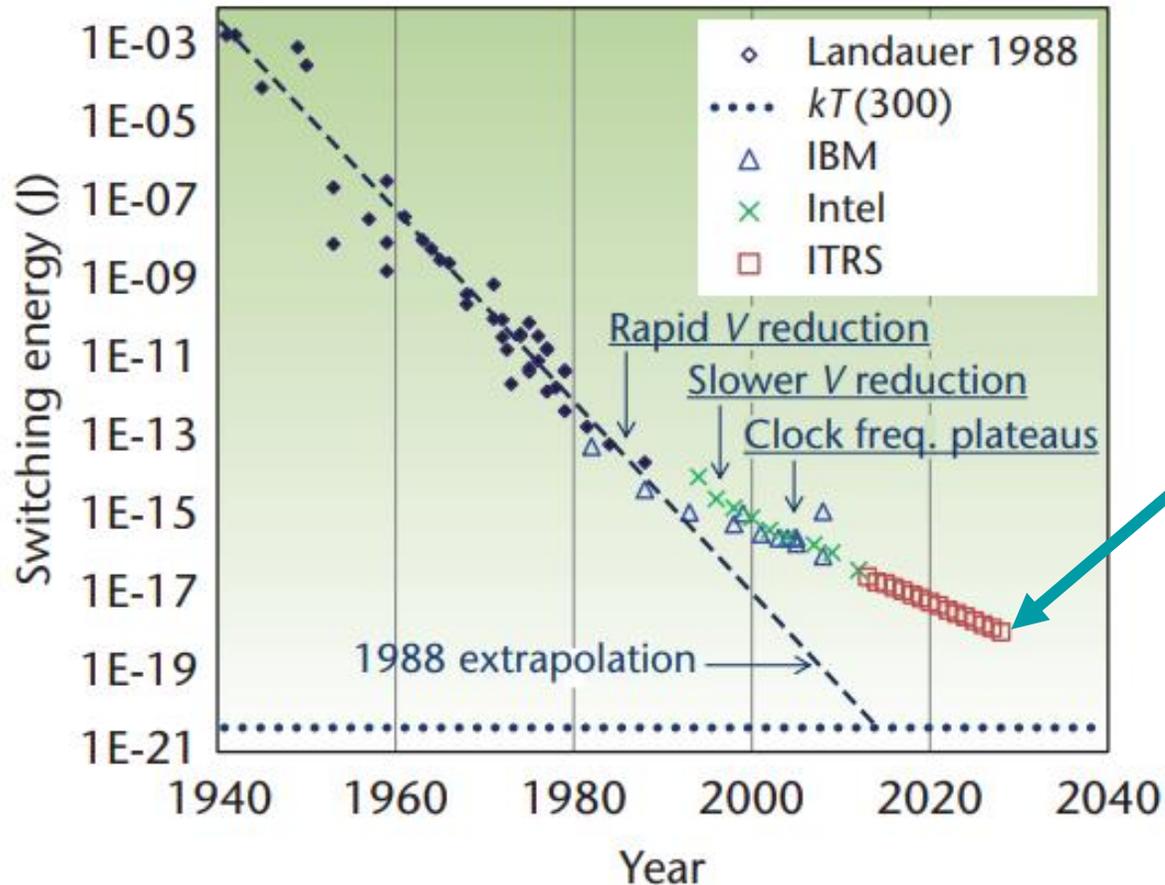
# Information processing predictions and energy requirements



Source: Chapter 5, The Decadal Plan for Semiconductors, January 2021. <https://www.src.org/about/decadal-plan/>

ZIPS = Zetta Instructions Per Second  
Zetta is  $10^{21}$

# Energy efficiency in processing information.



When	Switching energy dissipation
Today	10 aJ
2030	1 aJ
Landauer limit at room temp	0.003 aJ

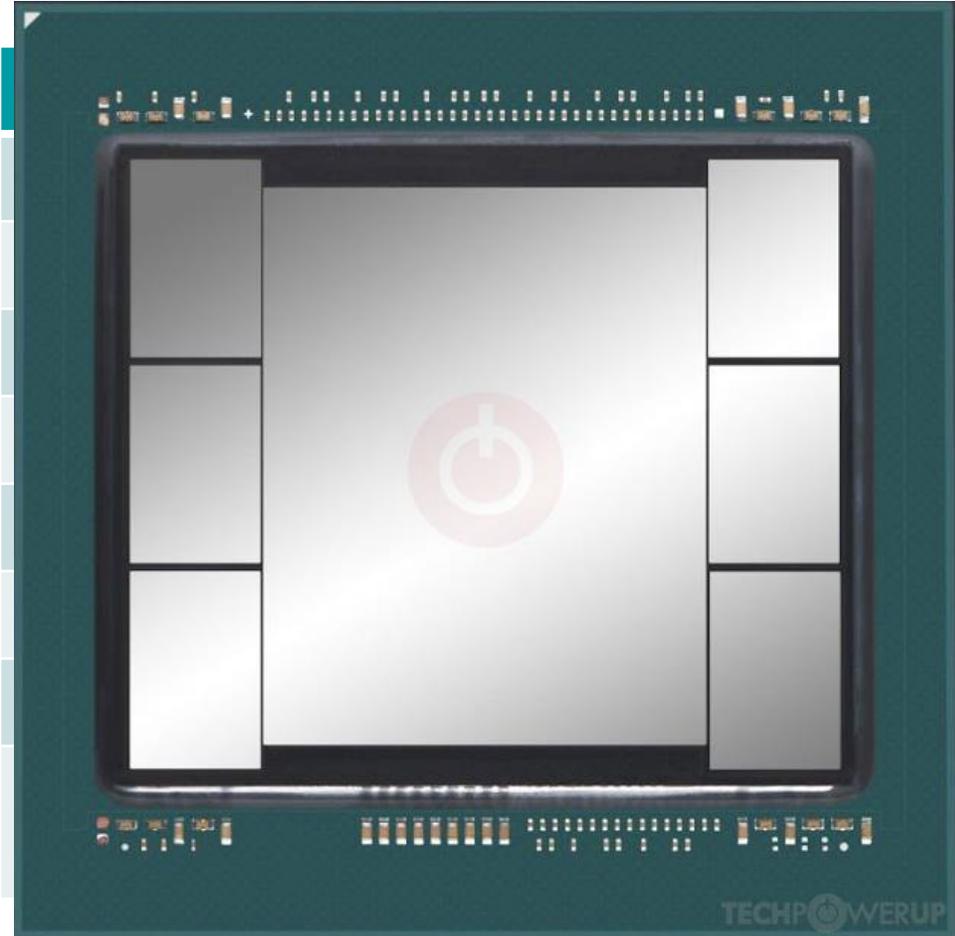
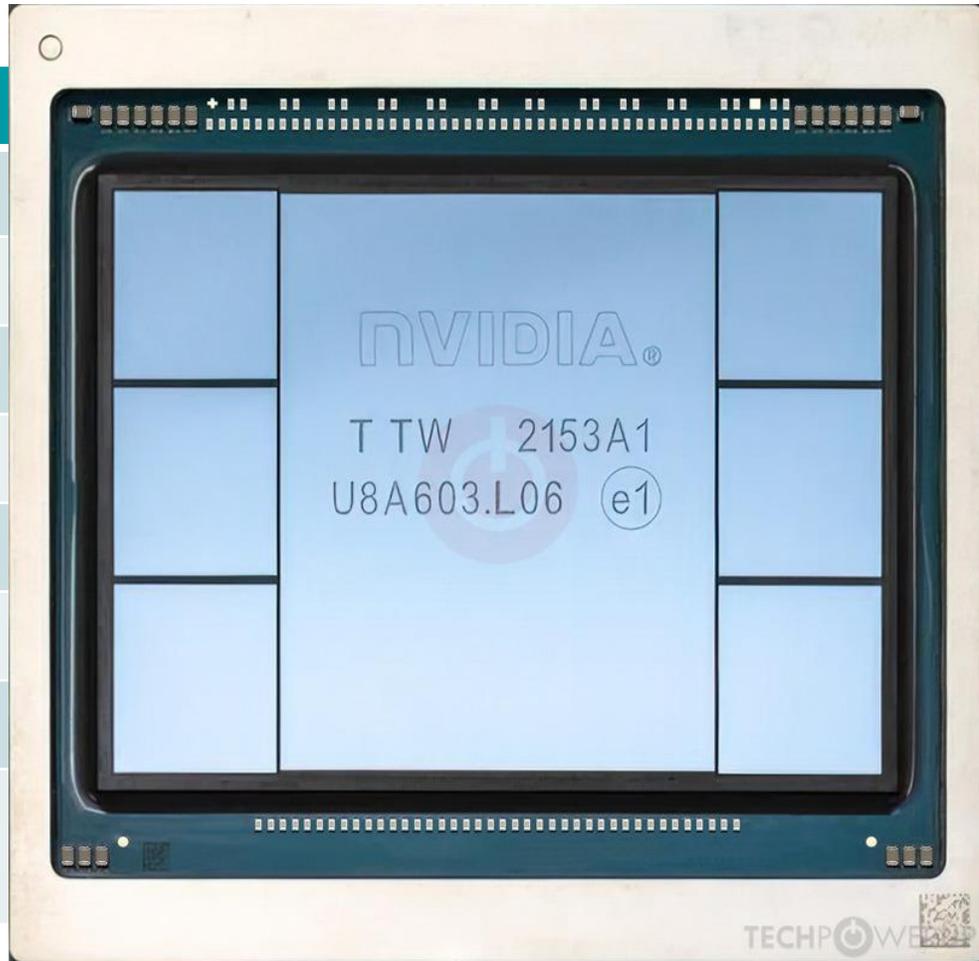
aJ is an attoJoule

1aJ = 0.00000000000000000001 J

Thomas N. **Theis**, H.-S. **Philip Wong**. "The End of Moore's Law: A New Beginning for Information Technology," *Computing in Science and Engineering*, IEEE CS and AIP, March/April, 2017

**Landauer**, R., 1988. Dissipation and noise immunity in computation and communication. *Nature*, 335(6193), pp.779-784.

# The thermal design power (TDP)



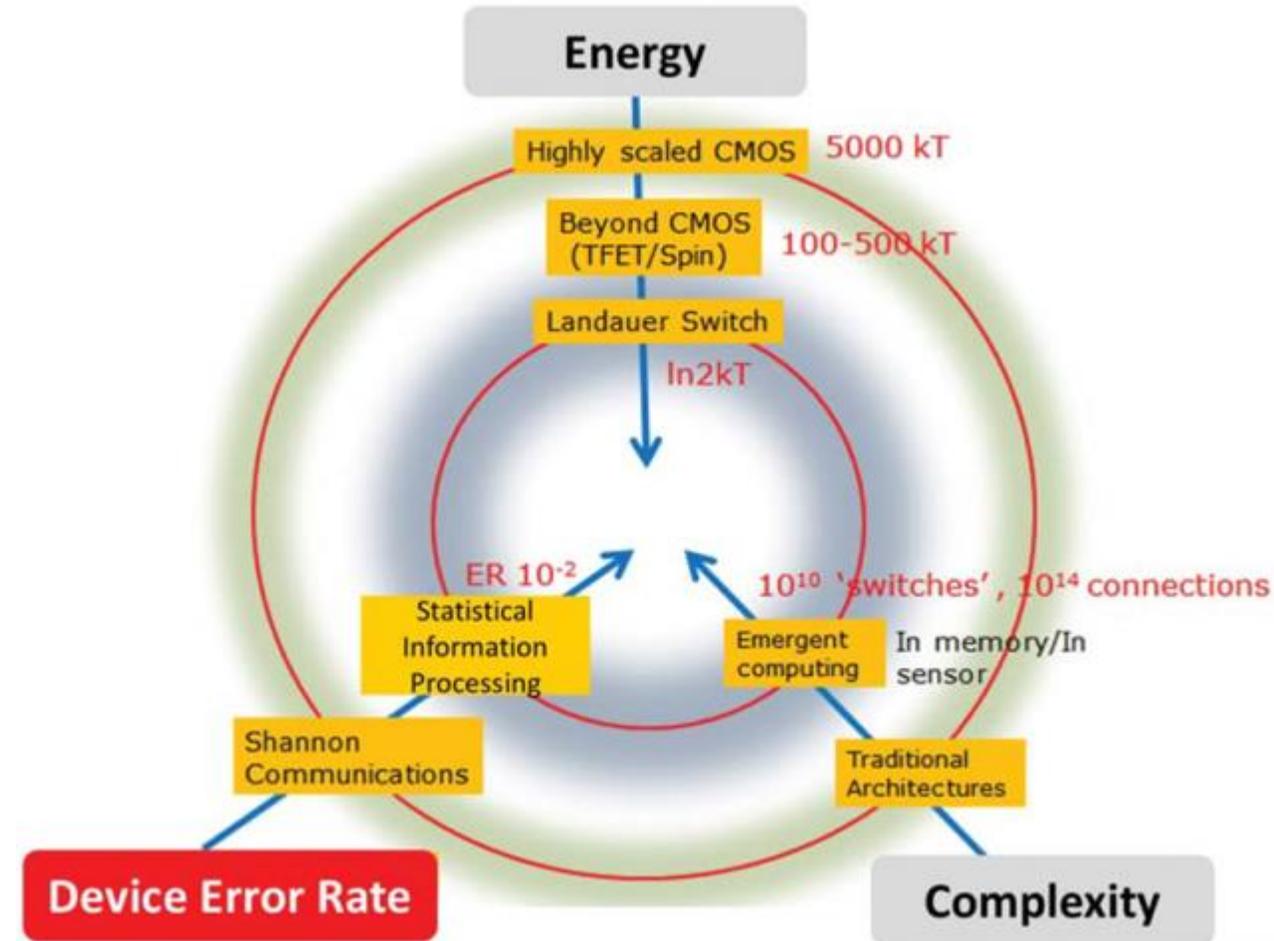
$$\text{Power (W)} = \text{Switch Energy (J)} \times \text{Switching Rate (s}^{-1}\text{)}$$

<https://www.techpowerup.com/gpu-specs/>

# Where are TDPs heading?

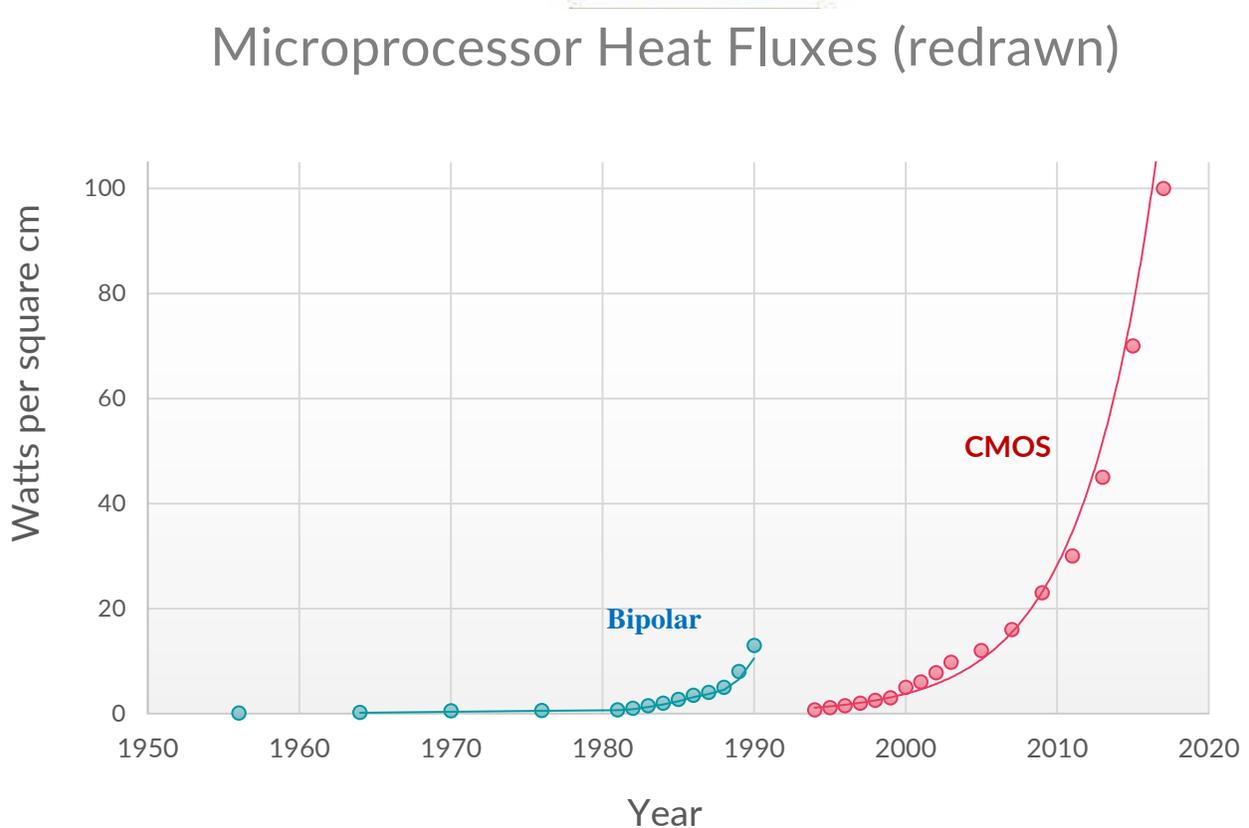
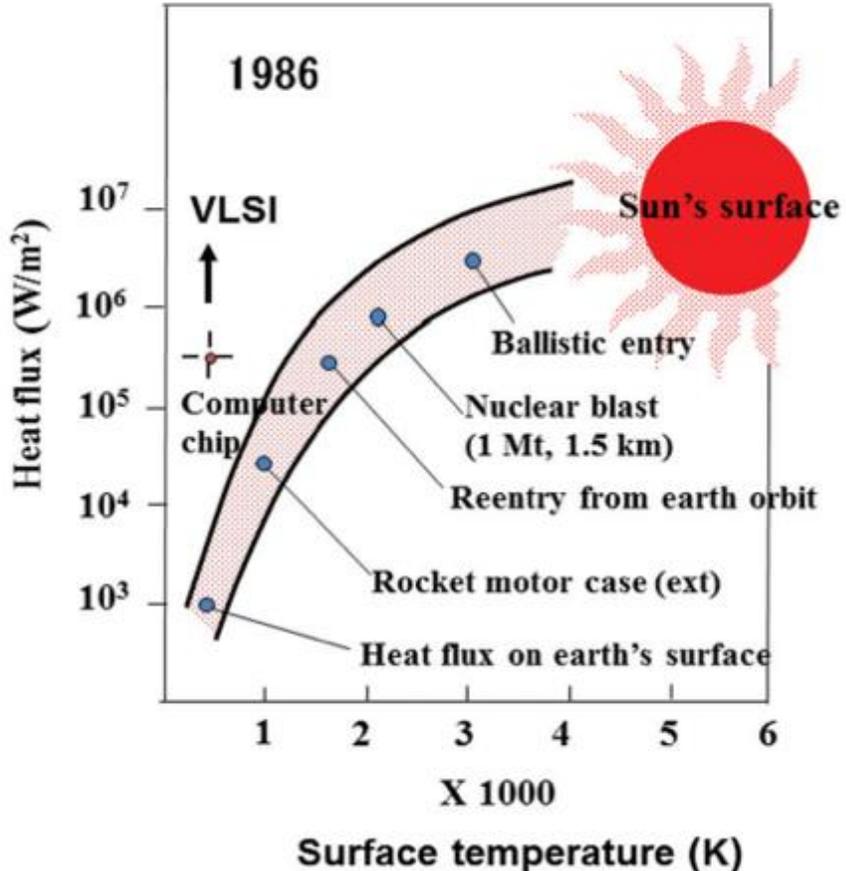
Intel microprocessor	2030
Energy dissipation (E)	1 aJ per transistor
#transistors (#T)	1 trillion
Switching frequency (f)	4 GHz (1.4 nm / A14)
Power=E x #T x f	4000 W
Dark silicon (assumption)	50%
TDP	2000 W
Die size	1250 mm <sup>2</sup>
Max heat flux HF= TDP/area	1.6 kW/m <sup>2</sup> T <sub>cmax</sub> < 60°C

Power (W)= Switch Energy (J) x Switching Rate (s<sup>-1</sup>)



Source: Chapter 5, The Decadal Plan for Semiconductors, January 2021. <https://www.src.org/about/decadal-plan/>

# Heat fluxes of microprocessors

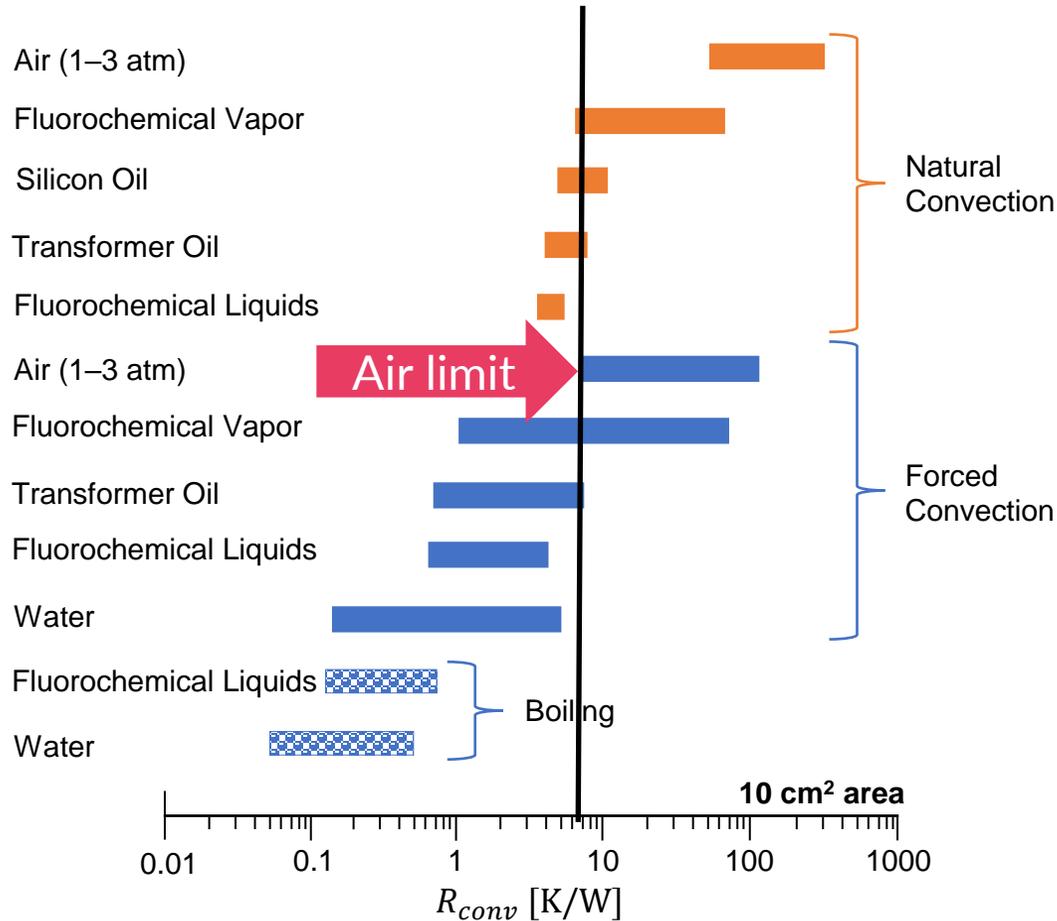


Oktaý, Sevgin, Robert Hannemann, and Avram Bar-Cohen. "High heat from a small package." *Mech. Eng.:(United States)* 108, no. 3 (1986).

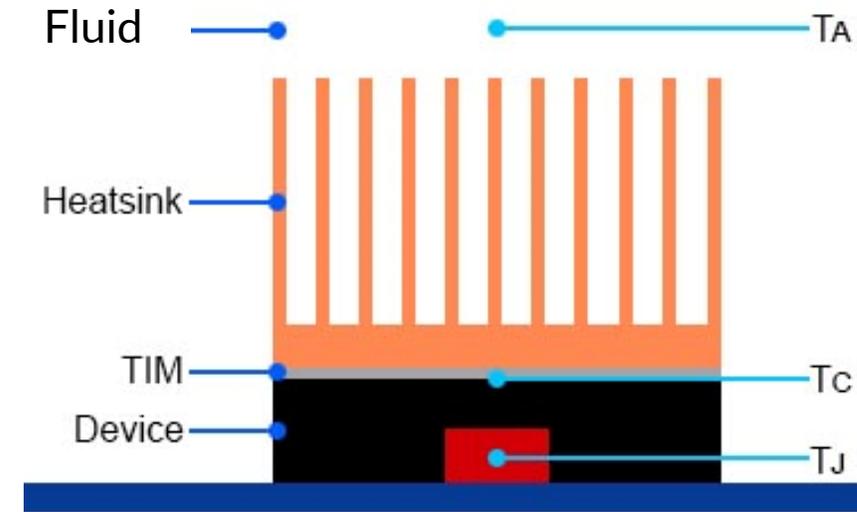
Roger Schmidt, Liquid Cooling is Back, *Electronic Cooling* August 2005. M.J.Ellsworth. Interpack '11 Tutorial



# Fluids are needed to remove the heat, but which?



Ahmed El-Neama based on Tummala, R.R. 2001.  
Fundamentals of Microsystems Packaging. McGraw-Hill.



700W spread over  $700\text{cm}^2 = 70 \times 10\text{cm}^2$   
 $1\text{W}/\text{cm}^2$  or  $10\text{W}/10\text{cm}^2$  from  $700\text{W}/70$   
 Temp difference between  $T_C$  and  $T_A$  is then  
 $10\text{W}/\text{cm}^2 \times R_{conv} \text{ } ^\circ\text{C}$

$$T_C - T_A = 10 R_{conv} \text{ } ^\circ\text{C} = \underline{55 \text{ } ^\circ\text{C}} \text{ for air } R_{conv} = 5.5$$

$$T_A = T_C - 55 = 85 - 55 = \underline{30 \text{ } ^\circ\text{C}} \text{ supply air.}$$

Could just and does run on “thin air”!

# Fluids are needed to remove the heat, but which?

Future 2000W XPU's with  $T_C$  of  $60^\circ\text{C}$  and  $T_A$  at  $20^\circ\text{C}$

2000W spread over  $1000\text{cm}^2$  say

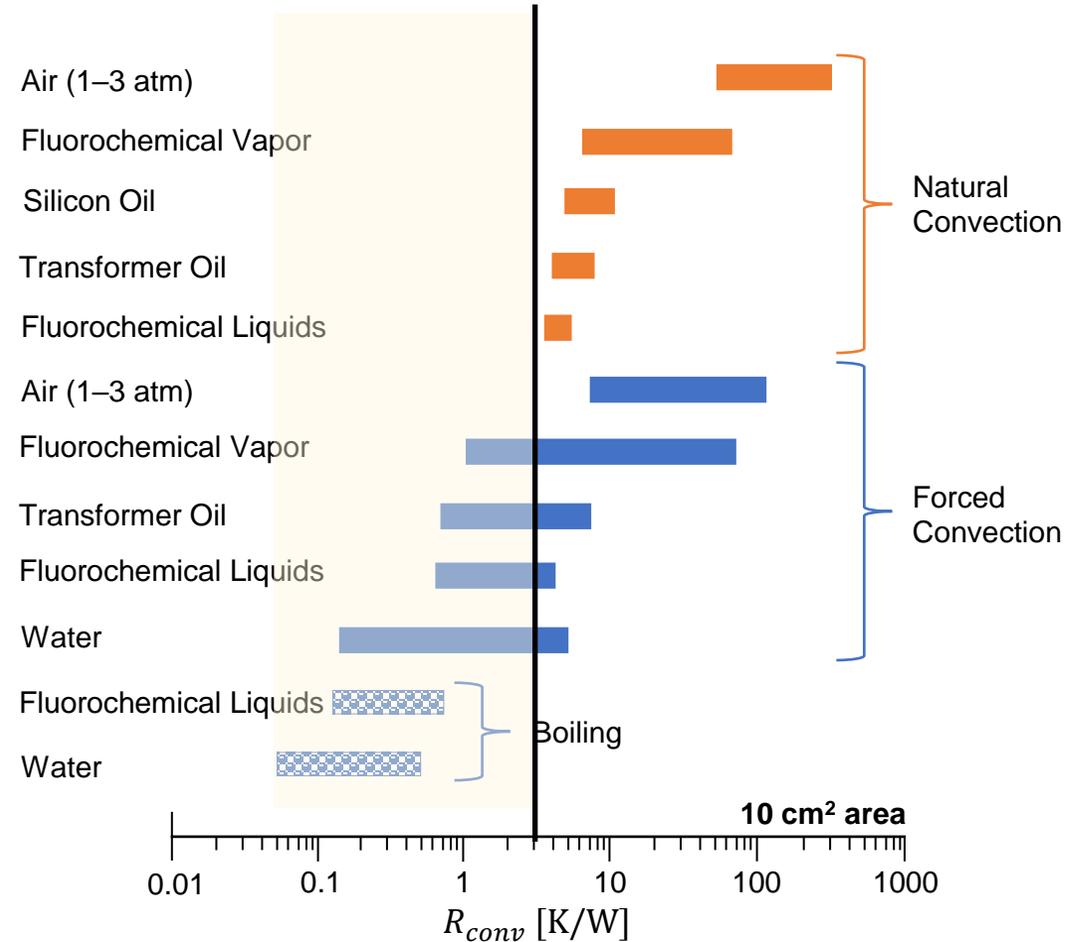
$$2000\text{W} / 100 = 20\text{W}/10\text{cm}^2$$

$$\text{Temp difference} = (20 R_{\text{conv}}) ^\circ\text{C}$$

$$T_A = T_C - 20 R_{\text{conv}} ^\circ\text{C}$$

With  $T_C$  at  $60^\circ\text{C}$   
Need  $R_{\text{conv}}$  of 3 or less.

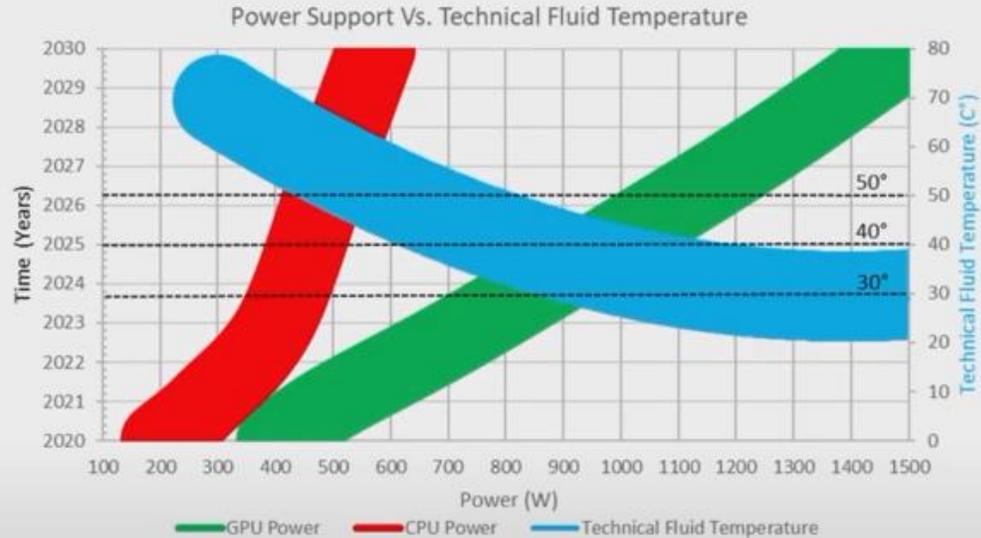
Cannot run on "thin air"!



# OCP 2023 presentations looking at the need for liquid cooling.

## GPU and CPU Power and Coolant Trends

Assumptions:  
1.5 lpm per 1KW



OCP REGIONAL SUMMIT | APRIL 19-20, 2023 | PRAGUE, CZ

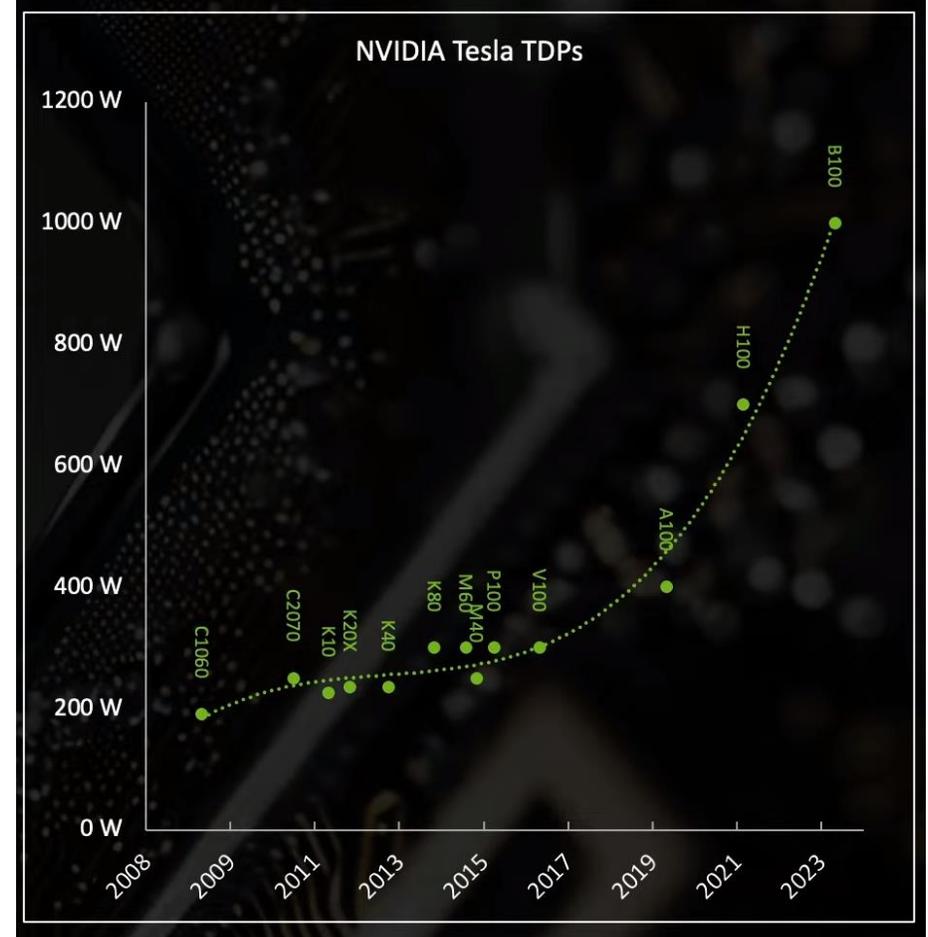
EMPOWERING OPEN.

Panel: Coolant Temperatures for Durable Data Center Designs  
<https://youtu.be/ToY2PBYK01Q>

## Igniting Change for the Future of Data Center Cooling

Rolf Brink, Promersion

<https://youtu.be/i5TSvB8Akz>

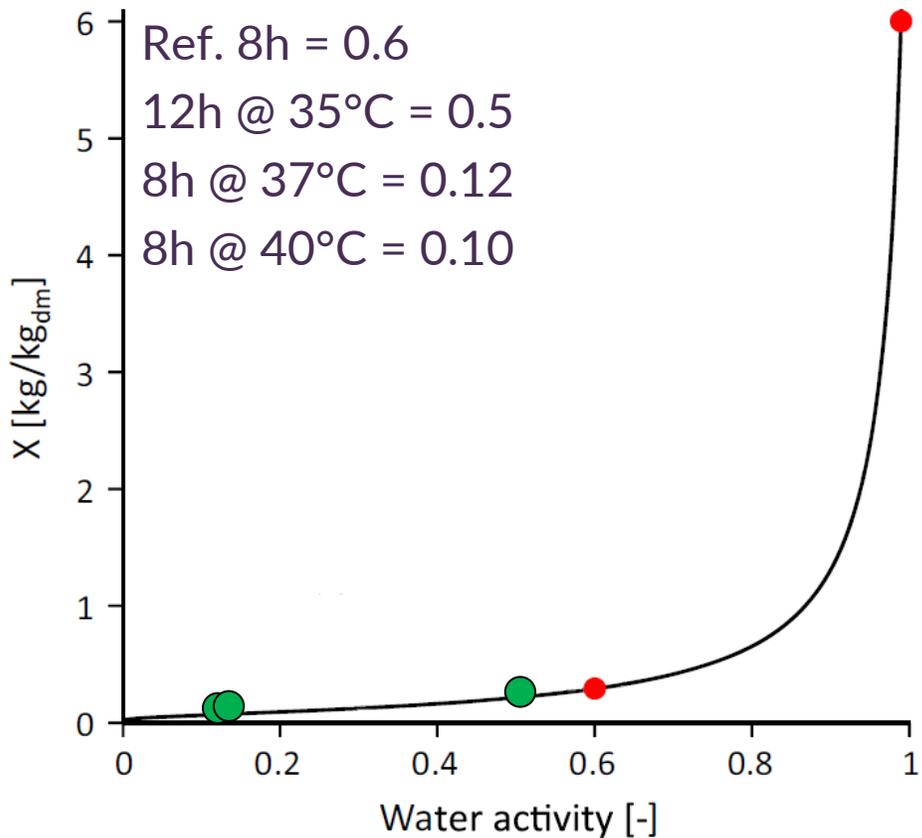


RISE

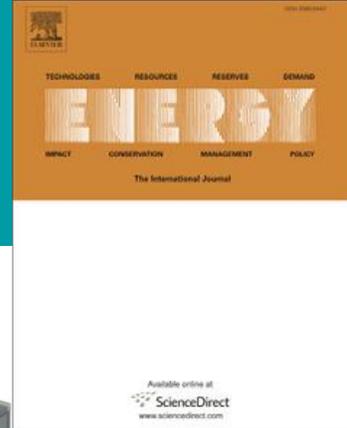
# Data Center Heat Reuse

# DC Excess Heat – Apple drying

Keep the berry harvest in the region for further processing, instead of going for export.



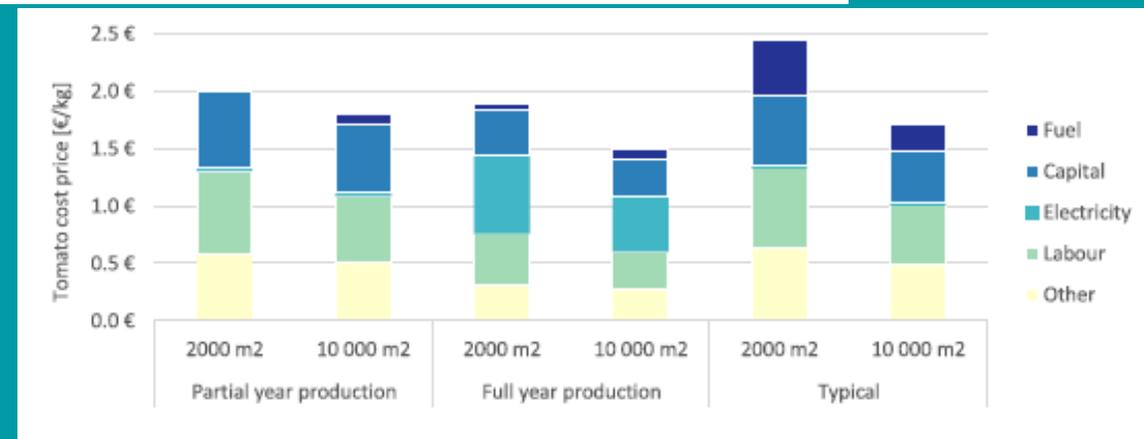
# DC Excess Heat – Green House



DC power 1 MW @ 27°C  
 Green House area 2´000, 10´000 m<sup>3</sup>  
 Partial and full year production

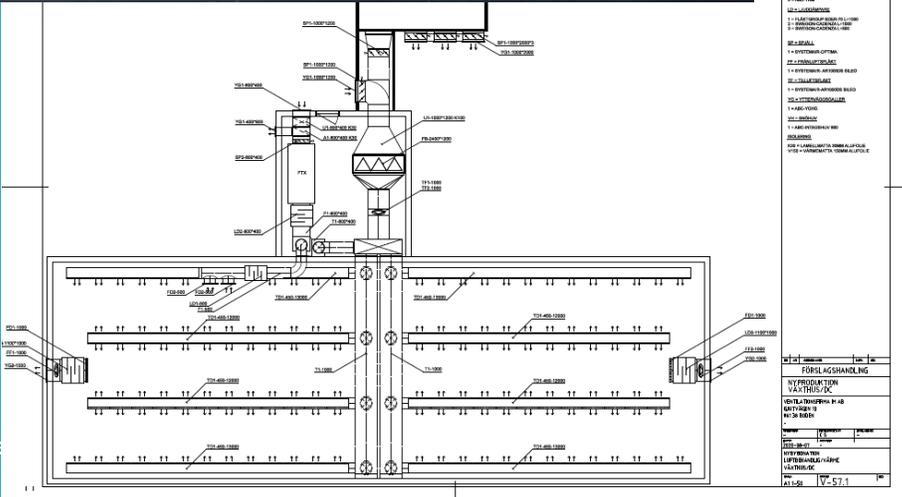


Conclusions  
 FYP and 10´000 m<sup>2</sup> most beneficial  
 Increase self-sufficiency ≤ 8%



Area [m <sup>2</sup> ]	2000		10 000	
	PYP	FYP	PYP	FYP
Production scenario				
Boiler peak heating demand [kW]	360	560	2560	3520
Boiler run time [h]	130	996	1908	4006
Excess heat ratio of total greenhouse heating demand	97.9%	89.7%	61.5%	50.0%
Excess heat recovered of the data centers total input	5.5%	14.0%	18.3%	30.5%

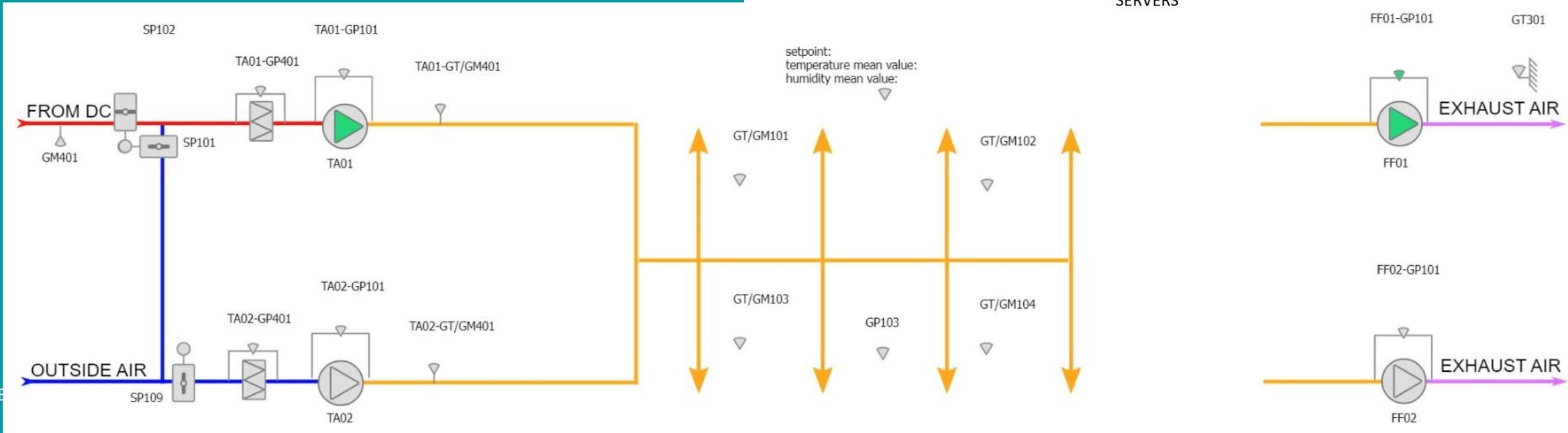
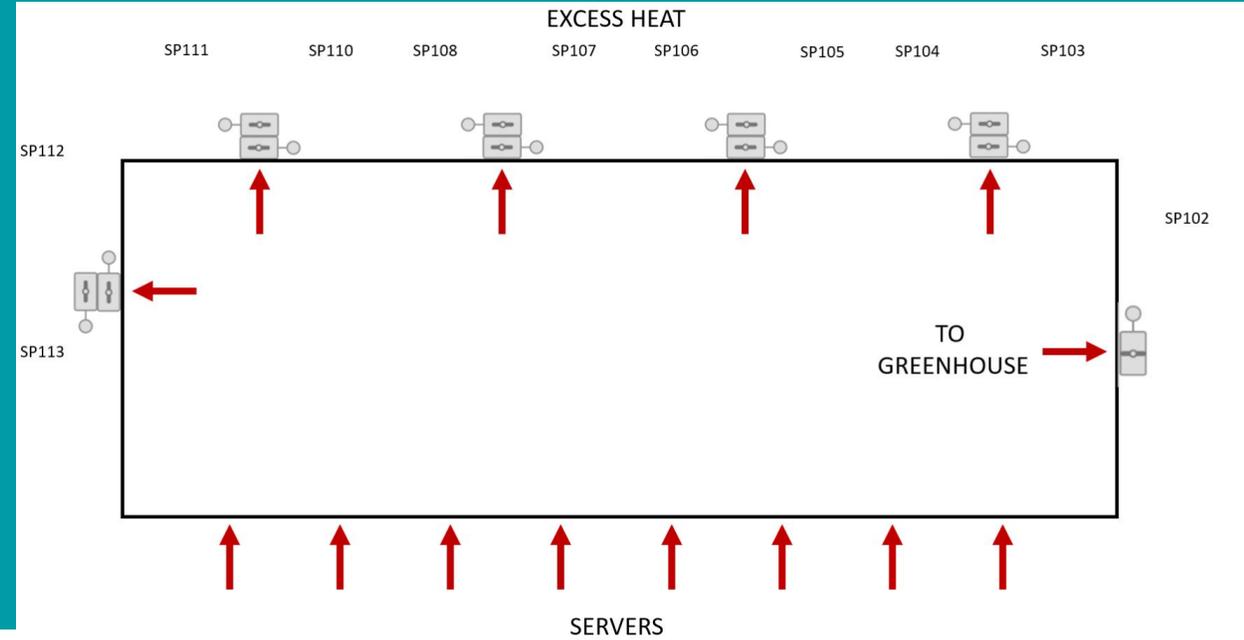
# DC Excess Heat – Green House



# DC Excess Heat – Green House

Data collection platform measuring

- Temperatures
- Flows
- Humidity
- Pressure
- Damper position



# DC Excess Heat – Mealworm Farming

10 kg of mealworms for 10 tests  
DC excess heat temp (20, 30°C)  
3 feeding schedules (basic, extra, tunned)  
With and without access to water

## Conclusions

Dominating parameters:

- DC excess heat temp
- Food
- Water

Reduction of production time 50%



# DC Excess Heat – Woodchip Drying

120 m<sup>3</sup> woodchips for 2 weeks and 10 tests

DC excess heat temp (35, 42°C)

Feeding rate (50, 75 100%)

Fan speed (50, 100%)

## Conclusions

Moister reduction  $\leq 10\%$

Energy Reuse Efficiency  $\leq 42\%$

Dominating parameters:

- DC excess heat temp
- Fan speed



# DC Excess Heat – Hot Tap Water

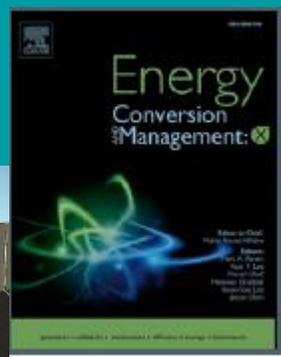
DC testbed for liquid cooled servers  
Connection of 2 parallel set-ups  
Part of the building hot tap water system

## Conclusions

Peak excess heat temp =>60 °C  
Hot tap water production without heat pump



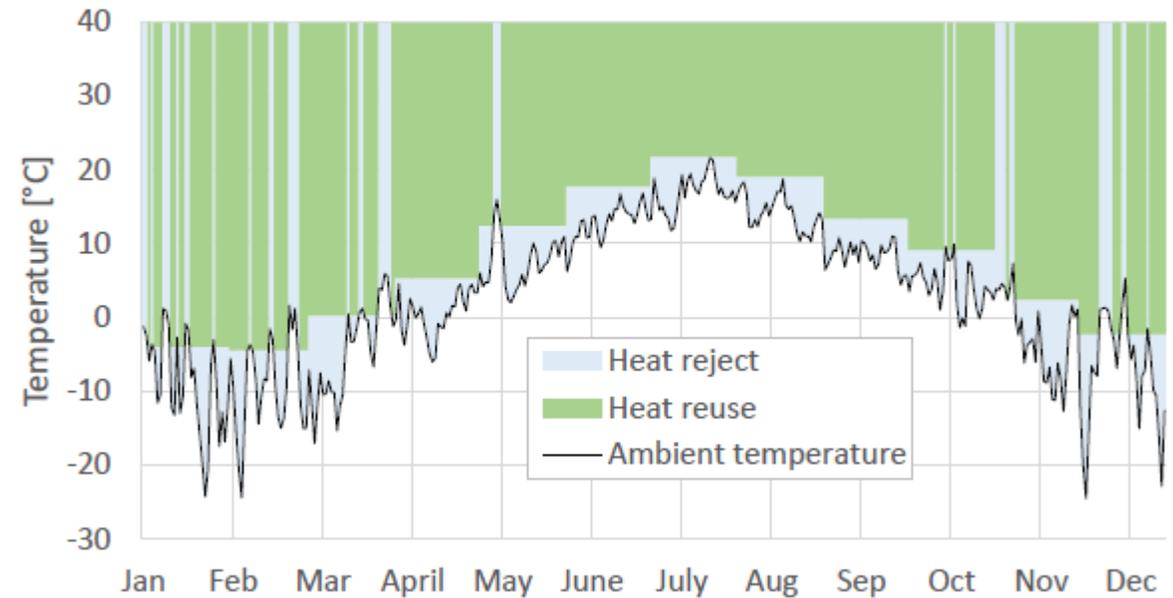
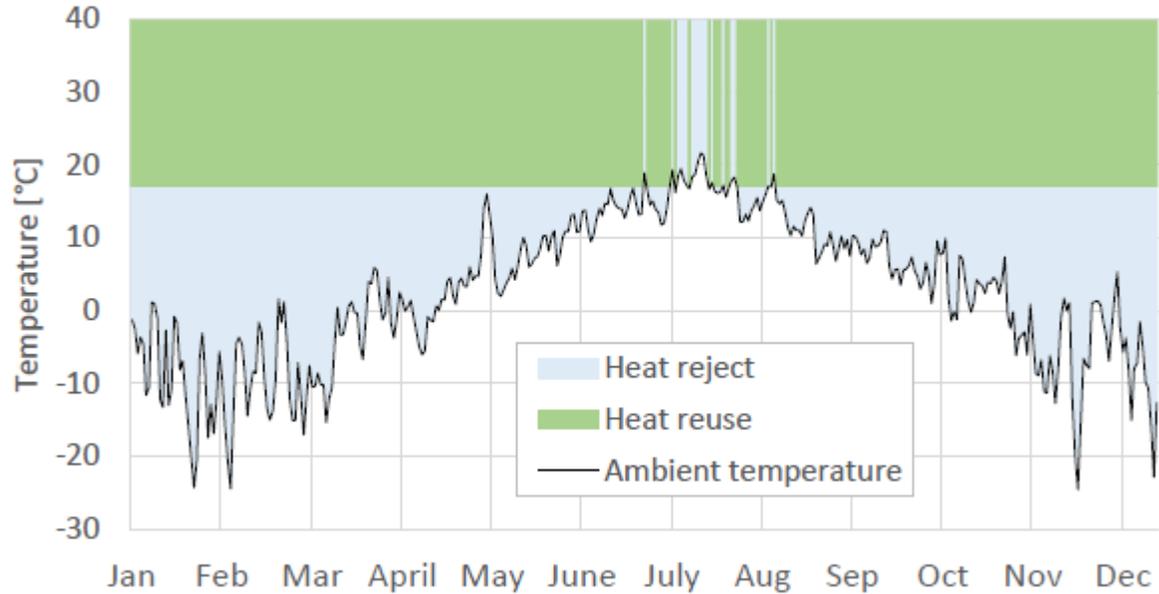
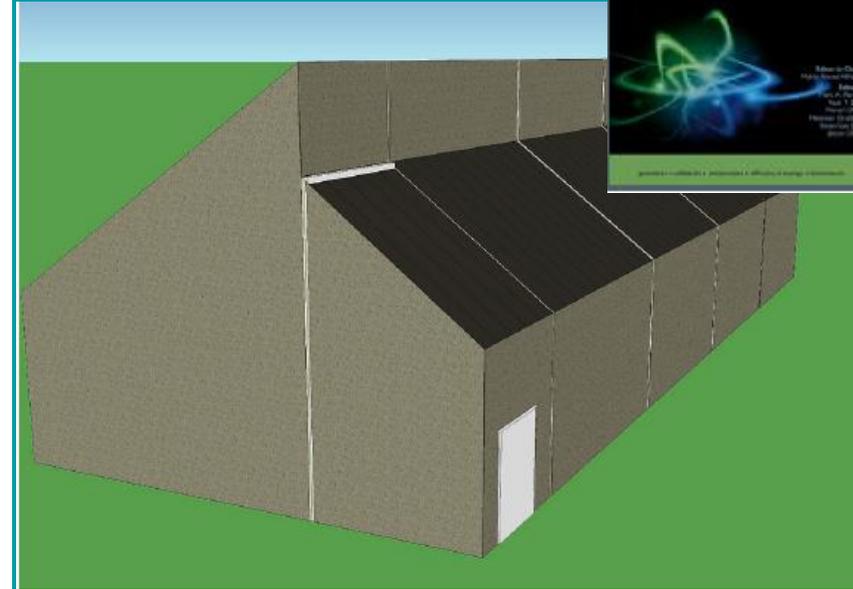
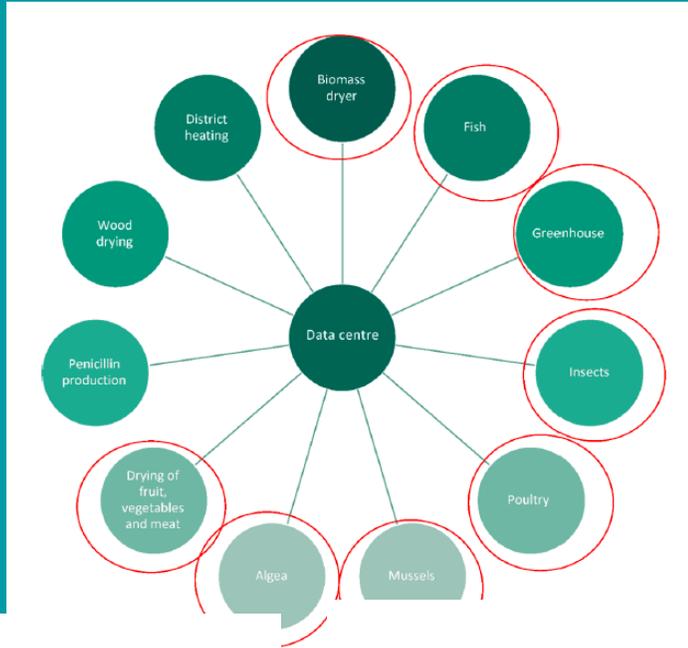
# DC Excess Heat – A realistic view



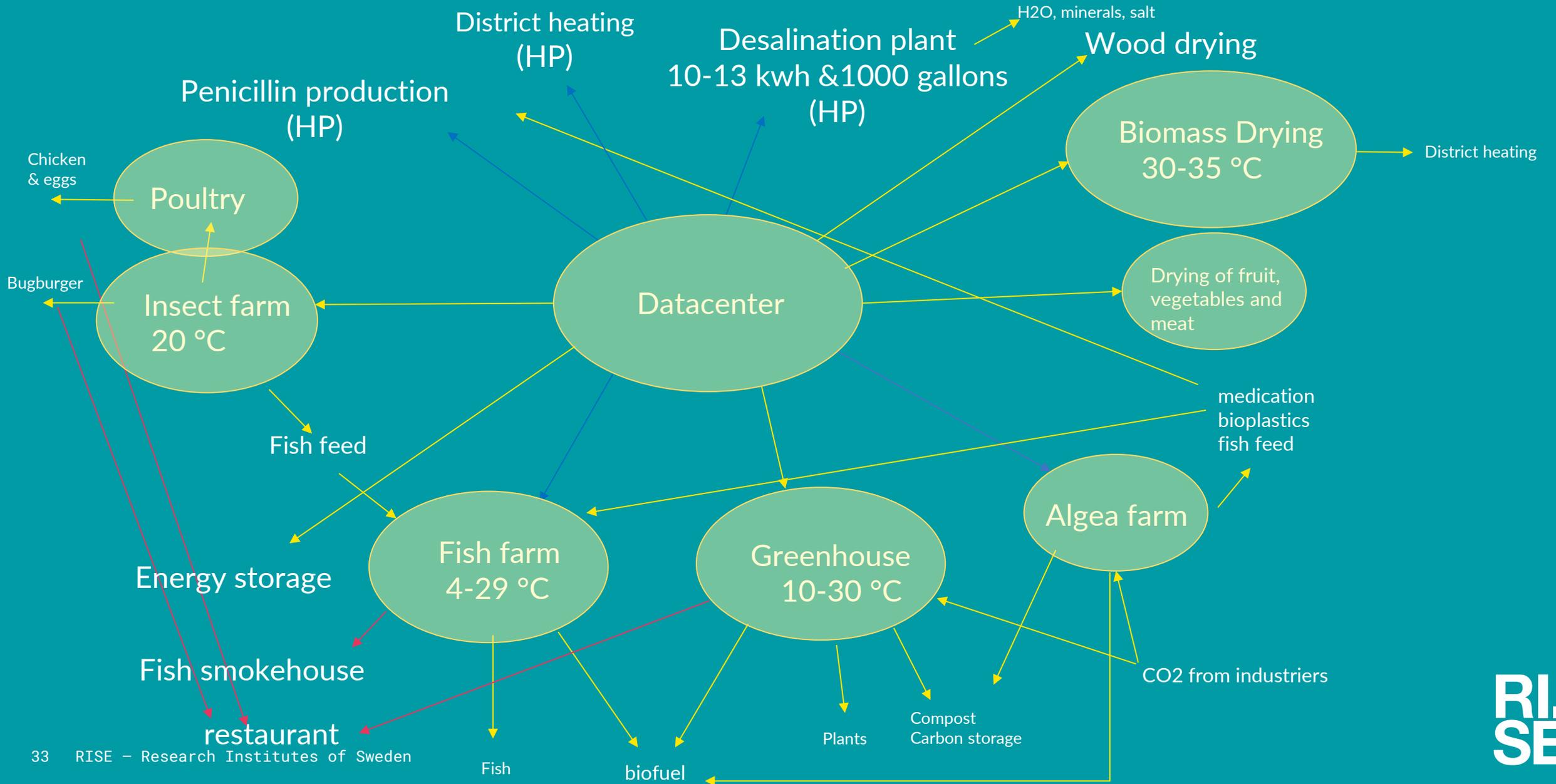
## Conclusions

ERE; 0.5, 0,59 and 0.66 for 30, 40 and 50°C respectively

11-31% better than 1 case application.



# DC Excess Heat – The IUS Matrix



RI.  
SE

Thank you!

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