

From seismic imaging to wind turbine modelling: benefits of the NEC SX-Aurora Vector Engine for the energy sector

Vincent Etienne, NEC Deutschland

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Introduction

HPC for the O&G

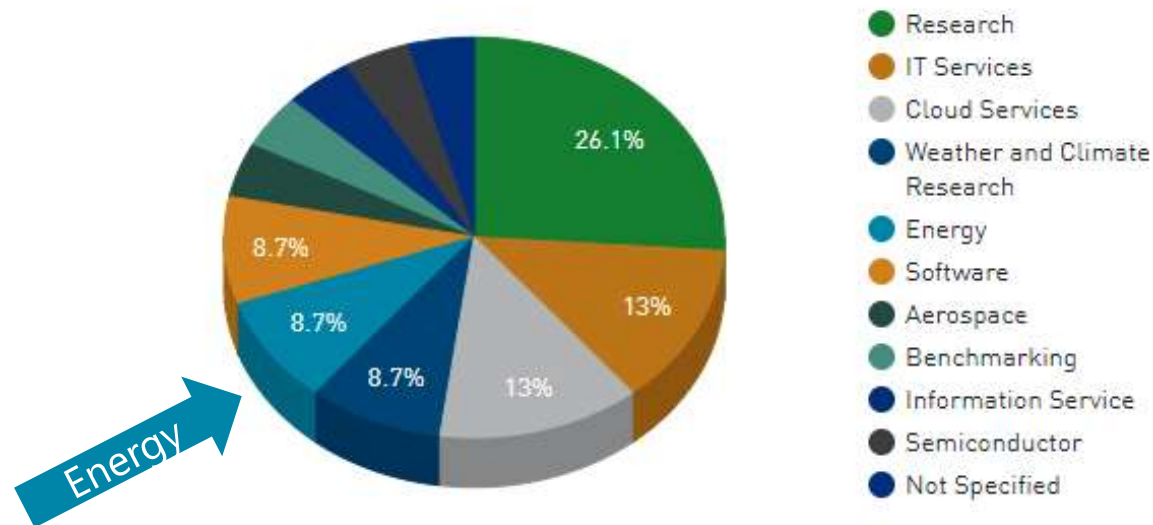
New energies, new HPC workloads

Introduction – HPC for the Energy sector



June 2022

Application Area System Share



Energy is #1 with Weather Forecast for number of HPC systems
In front of Aerospace, Car manufacturing & other industries

Introduction – HPC for the O&G



History of O&G systems in Top 10

- ◆ Nov 2019 None
- ◆ Jun 2020 #6 ENI → 1st industrial system in Top 10 is O&G
- ◆ Nov 2020 #8 ENI and #10 Saudi Aramco → 1st time 2 O&G systems
- ◆ Jun 2021 #9 ENI
- ◆ Nov 2021 #9 ENI
- ◆ June 2022 None

Today (June 2022) - O&G systems in Top 100

- ◆ #12 ENI (GPU)
- ◆ #18 Saudi Aramco (GPU)
- ◆ #28 Saudi Aramco (CPU)
- ◆ #33 TotalEnergies (GPU)
- ◆ #44 ENI (GPU)
- ◆ #60 Petrobras (GPU)
- ◆ #101 TotalEnergies (CPU)

Large investment of Oil majors in HPC
Significant increase over the past few years

Introduction – New energies, new HPC workloads

To reduce carbon footprint, renewable energies
are raising a lot of interests in our society
Especially in the O&G community



NEDO Green Japan,
Green Innovation

[Overview of the Green Innovation Fund Projects](#)

Working toward a carbon-neutral future.

The driving force behind Japan's future growth is the challenge of achieving carbon neutrality.

Now is the time for Japan-A technological superpower
One world-changing innovation after another.

Working together to create a carbon-neutral future.
A new Japan is waiting in 2050.



Sixth EAGE High Performance Computing Workshop

19-21 SEPTEMBER 2022 | MILAN, ITALY

[CALL FOR ABSTRACTS](#)

Sixth EAGE High Performance Computing Workshop

HPC: A Pathway to Sustainability

19-21 September 2022
Milan, Italy

Introduction – New energies, new HPC workloads

Most of oil majors are investing heavily in renewables, such as wind and solar, as they look to transition towards cleaner energy sources

Home > Our Company > Our expertise > Explore and produce > Renewable energies



Strengthening our presence in renewable energies

Electricity demand is set to rise faster than global demand for energy as a whole in the coming years. According to the International Energy Agency's Sustainable Development Scenario, renewable energies will represent more than 35% of the world's energy mix in 2040. To support this growth, our ambition is to achieve 100 gigawatts of gross installed renewable power generation capacity by 2030, through the development of our solar and wind energy businesses around the world.



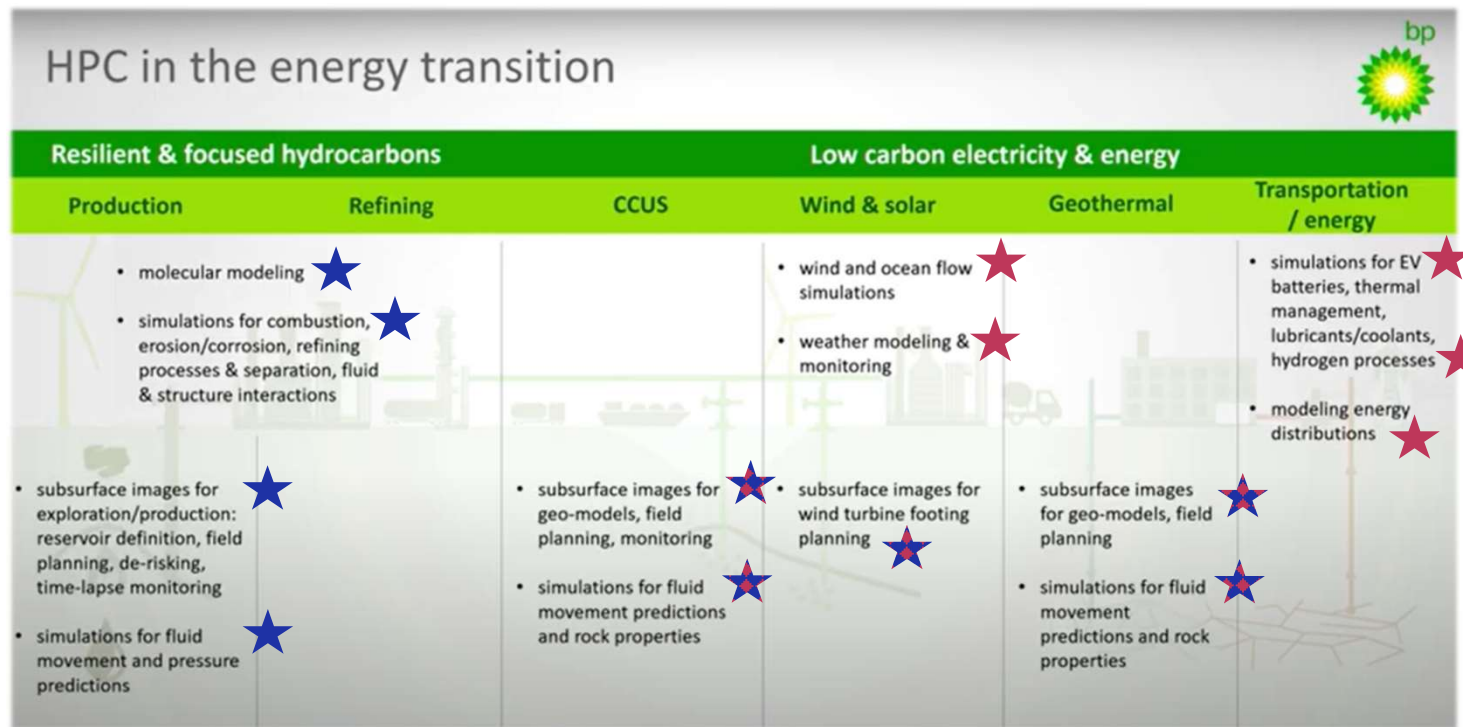
Solar and wind: Our ambition in renewable energies

Confident of the opportunities afforded by renewable energies, we aim to attain electricity production of 120 TWh by 2030, mainly through the development of solar and wind power.

From TotalEnergies website

Introduction – New energies, new HPC workloads

Besides traditional applications for O&G, new ones are expected to increase the HPC requirements



Source web (adapted): HPC and ML Aspects of the Energy Transition (RICE university)

★ Traditional application

★ Trad. with new use case

★ New application

**What are the compute footprints
of traditional applications for O&G?**

**How different it will be
for new kind of applications
involved in the energy transition?**

What are the benefits of NEC Vector Engine?

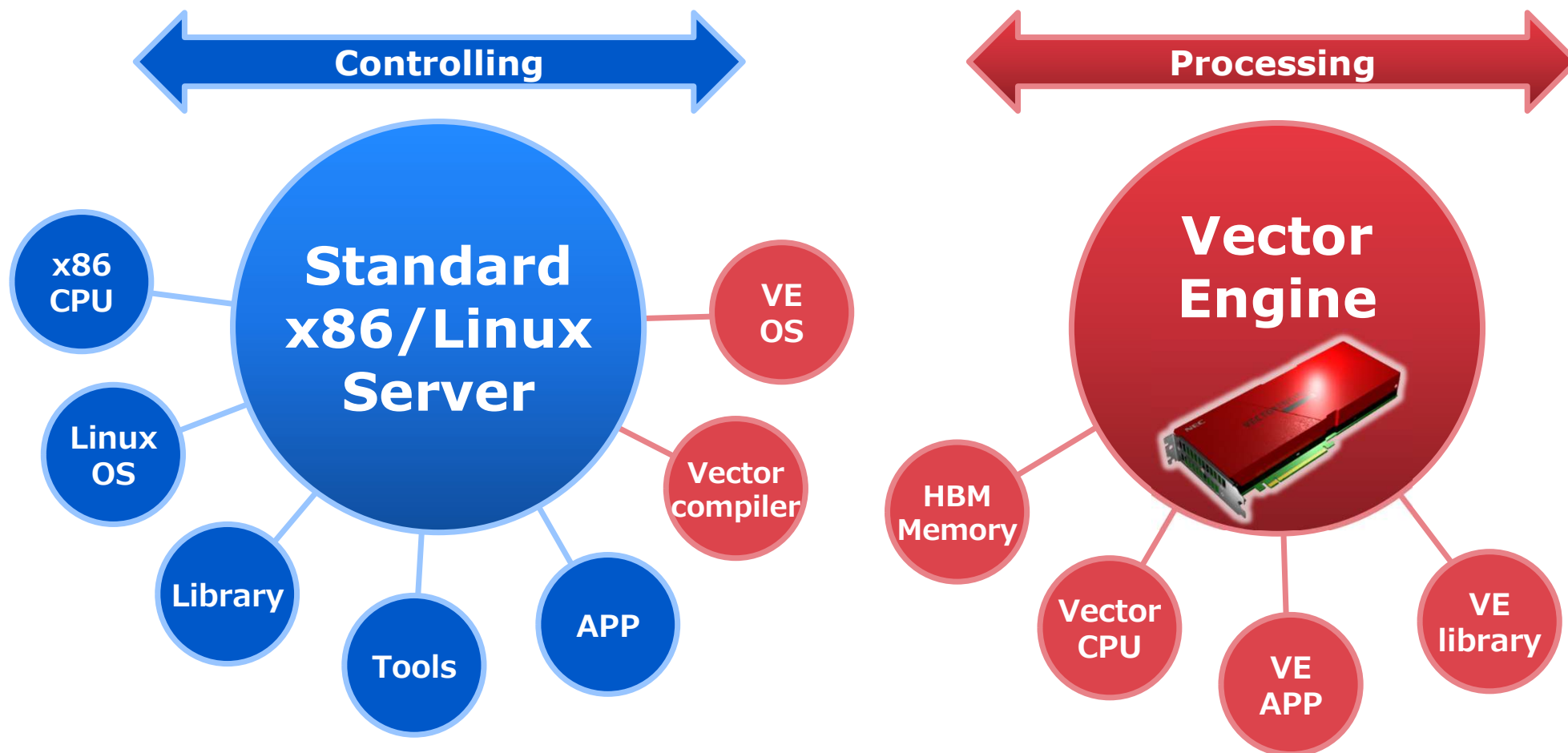
SX-Aurora TSUBASA Vector Engine

Overview

Native & accelerator modes

NEC SX-Aurora TSUBASA Vector Engine - Overview

Architecture



NEC SX-Aurora TSUBASA Vector Engine - Overview

VE20 (current generation)
Vector length 256 double floats (512 single)
64 vector registers

Processor Version	Type A	Type B
Cores/processor	10	8
Core performance	307GF (DP) 614GF (SP)	
Processor performance	3.07TF (DP) 6.14TF (SP)	2.45TF (DP) 4.91TF (SP)
Cache capacity	16MB	
Cache bandwidth	3TB/s	
Cache Function	Software Controllable	
Memory capacity	48GB	
Memory bandwidth	1.53TB/s	
Power	~300W (TDP) ~200W (Application)	

Development Philosophy

POINT
1

Strong cores

Highest memory bandwidth per core
Highest peak performance per core

POINT
2

Easy to Use

Fortran/C/C++ programming, OpenMP
Automatic vectorization/parallelization

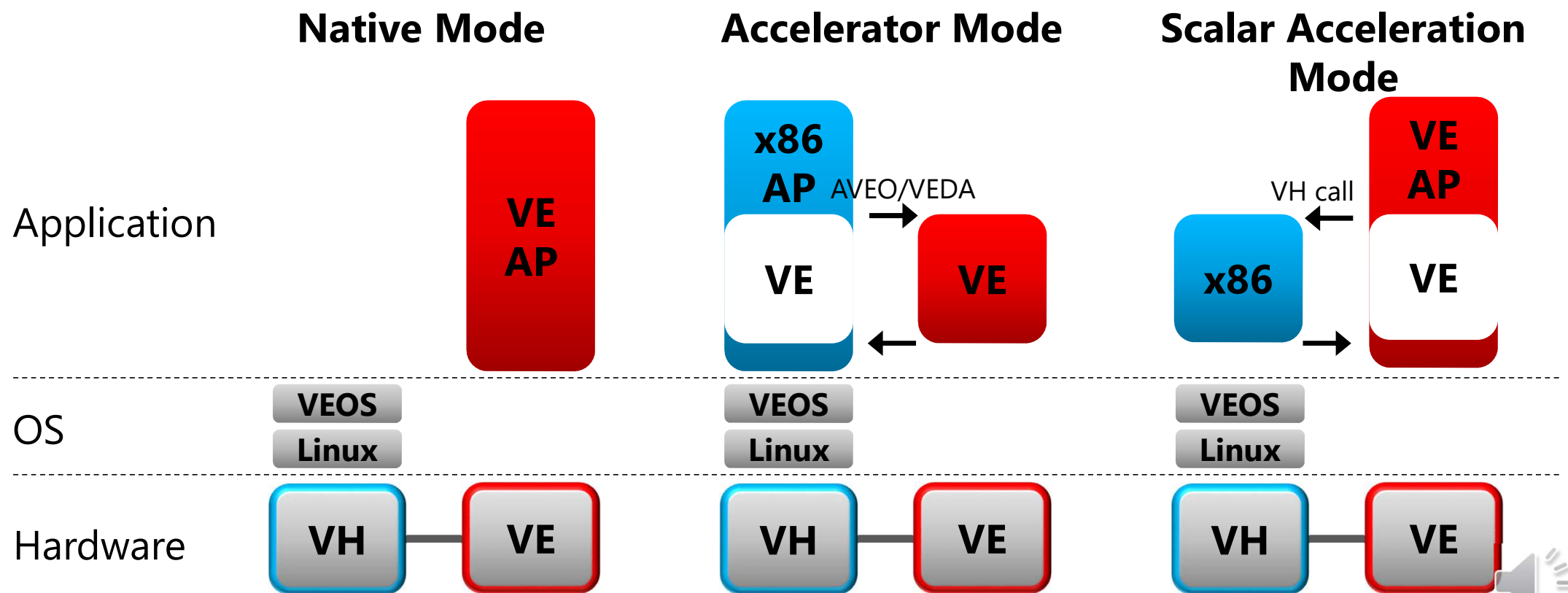
POINT
3

Energy Efficient

Power limitation for HPC systems becomes an issue
Higher sustained performance with lower power

NEC SX-Aurora TSUBASA VE – Native & accelerator modes

Vector Engine (VE) programming model: **a unique feature**



NEC SX-Aurora TSUBASA VE – Native & accelerator modes

Example with OpenFAST

- ◆ Open-source wind turbine simulation tool
 - Fortran
 - OpenMP
- ◆ Dependencies
 - BLAS
 - LAPACK
- ◆ Compilation with Cmake

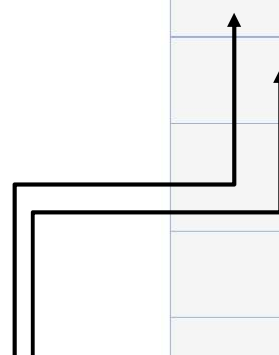
Porting on SX-Aurora with Native mode

- ◆ Straight forward compilation with nfort
 - Natural support for Fortran and OpenMP
- ◆ Linking with NLC (NEC Numeric Library Collection)



NEC NLC

Library		Functions
ASL	Native Interface	Scientific library with a wide variety of algorithms for numerical/statistical calculations
	Unified Interface	Fourier Transforms, Random Number Generators, Sortings
	FFTW3 Interface	Interface library to use Fourier Transform functions of ASL with FFTW (version 3.x) API
	BLAS	Basic Linear Algebra Subprograms
	LAPACK	Linear Algebra PACKage Simultaneous linear equations, Eigenvalue equations, and Singular value decomposition
	ScaLAPACK	Scalable Linear Algebra PACKage Simultaneous linear equations, Eigenvalue equations, and Singular value decomposition (for distributed memory parallel programs)
	BLACS	Basic Linear Algebra Communication Subprograms (uses MPI) Message passing library for performing basic operations on vectors and matrices (for distributed memory parallel programs)
	SBLAS	Sparse BLAS (from ACM Algorithm 692) Basic operations of sparse matrices
	HeteroSolver	Simultaneous linear equations (Direct sparse solver)
	Stencil Code Accelerator	Stencil Code Acceleration



NEC SX-Aurora TSUBASA VE – Native & accelerator modes

Example with OpenFAST

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Testing / benchmarking on SX-Aurora in progress

```
[vetienne@arrproto27 Release_nec_openmp]$ ./glue-codes/openfast/openfast -v

*****
OpenFAST

Copyright (C) 2022 National Renewable Energy Laboratory
Copyright (C) 2022 Envision Energy USA LTD

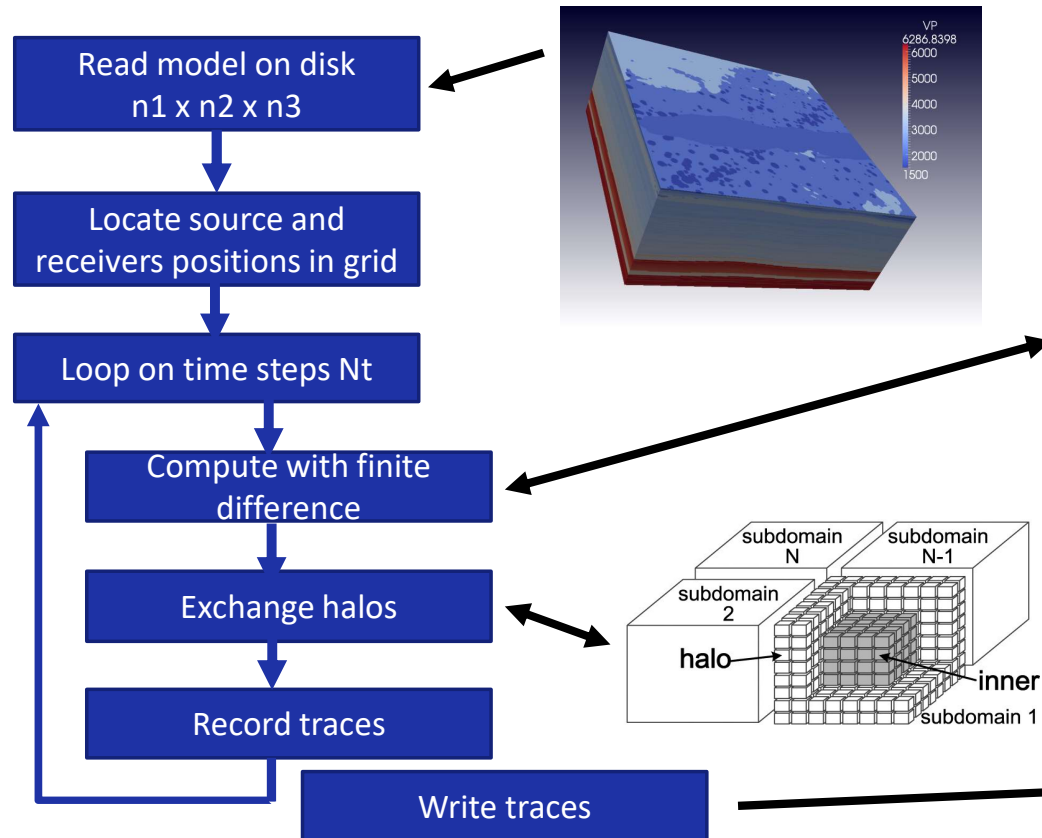
This program is licensed under Apache License Version 2.0 and comes with ABSOLUTELY NO WARRANTY.
See the "LICENSE" file distributed with this software for details.
*****

OpenFAST-v3.2.1-dirty
Compile Info:
- Compiler: nfort (NFORT) 3.5.1
- Architecture: 64 bit
- Precision: double
- OpenMP: Yes, number of threads: 8/8
- Date: Sep 13 2022
- Time: 10:13:33
Execution Info:
- Date: 09/13/2022
- Time: 11:11:49+0200

OpenFAST terminated normally.
```

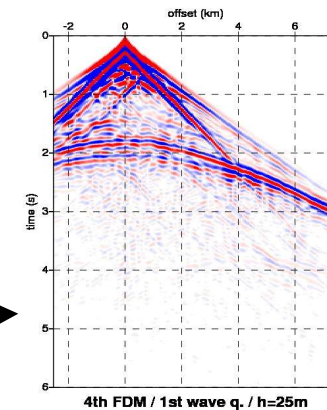
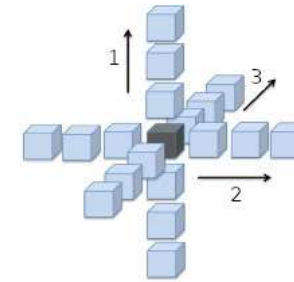

NEC SX-Aurora TSUBASA VE – Native & accelerator modes

Example of a seismic imaging application Porting with accelerator mode



Lots of disk IOs & data manipulations
→ Better do it on CPU (scalar operations)

Computation FD kernels localized
→ **Offloading kernels** on SX-Aurora appropriate



NEC SX-Aurora TSUBASA VE – Native & accelerator modes

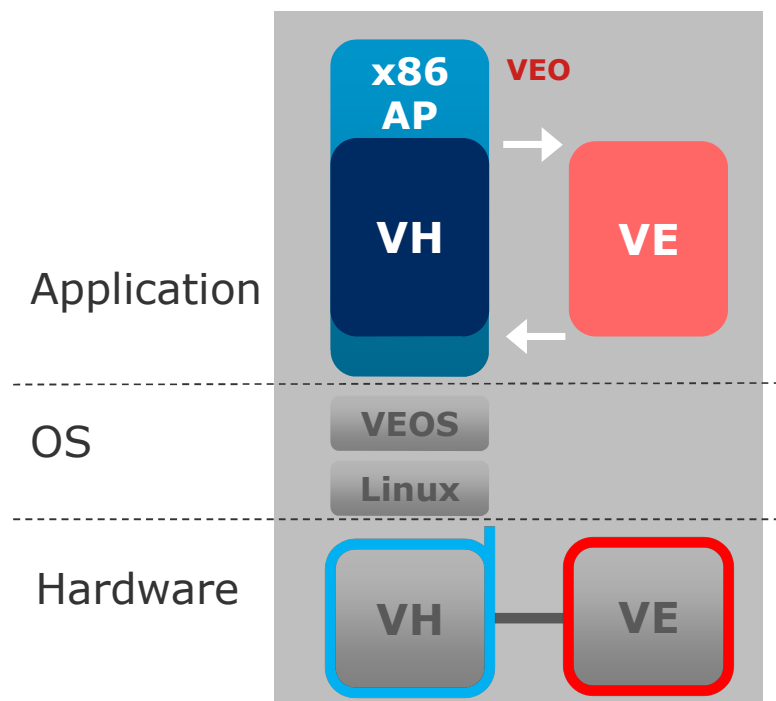
Accelerator mode with VEDA

- ◆ Offloading library developed by NEC LAB Europe
- ◆ Implements CUDA-like APIs
- ◆ Support C++

Benefits

- ◆ When existing code supports both CPU & GPU
 - Offloading sections already identified
 - Could **reuse CPU kernels** as is in VEDA kernels
- ◆ **Porting with VEDA is easy**
 - No need to learn new programming model → **Great benefit compared to constructor specific (ex: CUDA)**

Offloading with VEDA



NEC SX-Aurora TSUBASA VE – Native & accelerator modes

Original kernel CPU version

- $y = ax + b$
- C++
- simple OpenMP construct

```
template<typename ValueType>
void eq_axpb(ValueType* dst, ValueType a,
             const ValueType* x, ValueType b, size_t n) {
#pragma omp parallel for schedule(static) default(none) shared(dst,n,x,a,b)
    for (size_t i=0;i<n;i++) dst[i]=a*x[i]+b;
}
```

compiled with g++ or icpc

NEC SX-Aurora TSUBASA VE – Native & accelerator modes

Original kernel CPU version

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template<typename ValueType>
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    for (size_t i=0;i<n;i++) dst[i]=a*x[i]+b;
}
```

compiled with g++ or icpc

Equivalent VEDA version

- Launch VEDA kernel from CPU
- Similar to a CUDA call

```
template <typename ValueType>
void eq_axpb(ValueType *dst, ValueType a,
             const ValueType *x, ValueType b, size_t n, const geodrive::runtime::queue_t *qid)
{
    VEDAptr<ValueType> vdst = (VEDAdeviceptr)dst;
    VEDAptr<ValueType> vx = (VEDAdeviceptr)x;
    VEDAfunction func = Ctxt::vedaGetFunctionFromLibrary<ValueType>(Ctxt::vedaGetSystemLibrary(), "eq_axpb");
    VEDACHK(vedaLaunchKernel(func, 0, vdst.ptr(), a, vx.ptr(), b, n), "vedaLaunchKernel failed");
    Ctxt::sync();
}
```

compiled with g++ or icpc

NEC SX-Aurora TSUBASA VE – Native & accelerator modes

kernel VEDA version

- $y = ax + b$
- C++
- simple OpenMP construct

```
extern "C" void VEDA_FNAME(eq_axpb, ValueType) (ValueType *dst, ValueType a,
                                                const ValueType *x, ValueType b, size_t n)
{
    #pragma omp parallel for schedule(static) default(none) shared(dst, n, x, a, b)
    for (size_t i = 0; i < n; i++)
        dst[i] = a * x[i] + b;
}
```

Equivalent VEDA version

- Launch VEDA kernel from CPU
- Similar to a CUDA call

**compiled with nc++
C++ original CPU kernel as is**

```
template <typename ValueType>
void eq_axpb(ValueType *dst, ValueType a,
             const ValueType *x, ValueType b, size_t n, const geodrive::runtime::queue_t *qid)
{
    VEDAptr<ValueType> vdst = (VEDAdeviceptr)dst;
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    VEDACHK(vedaLaunchKernel(func, 0, vdst.ptr(), a, vx.ptr(), b, n), "vedaLaunchKernel failed");
    Ctxt::sync();
}
```

compiled with g++ or icpc

Traditional applications for O&G

Seismic imaging & full waveform inversion

Reservoir simulation

Geosciences & NEC

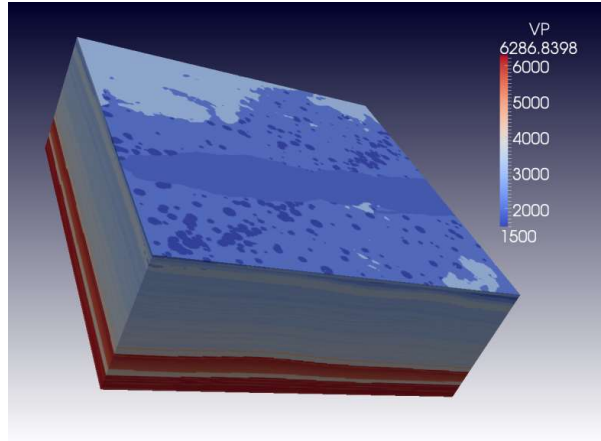
A bit of history with SPECFEM3D (Seismic/earthquake modelling)



Collaboration with D. Peter
Main developer of SPECFEM3D

Traditional applications for O&G

In O&G
3 main applications
Reservoir simulation
Seismic imaging &
inversion

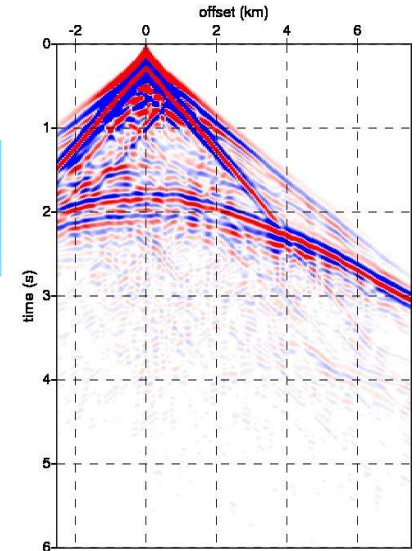


Model

Seismic modeling
Forward problem



Seismic inversion
Inverse problem



Data

Compute intensive
applications that require
HPC systems

Ex. Elastic modeling in
large scale survey
about 100 EFlop



HPC

Traditional applications – Memory bandwidth

Representative products available on the market (2021)

Spec.	CPU	GPU	VE
# cores	2 x 38	5 120	8
Memory type	DDR4	HBM2	HBM2
Capacity GB	256	32	48
Peak GB/s	410	900	1351

Measured memory bandwidth for simple operations on arrays

Case	CPU		GPU		VE	
	GB/s	% peak	GB/s	% peak	GB/s	% peak
Fill	200	49	725	81	1060	78
Copy	300	73	616	68	1085	80
Add	301	73	694	77	1083	80
Mul	300	73	695	77	1084	80
AddUpd	294	72	691	77	1083	80

Architectures reach 70-80 % of their peak bandwidth

Traditional applications – Seismic imaging

Optimal implementation for each architecture

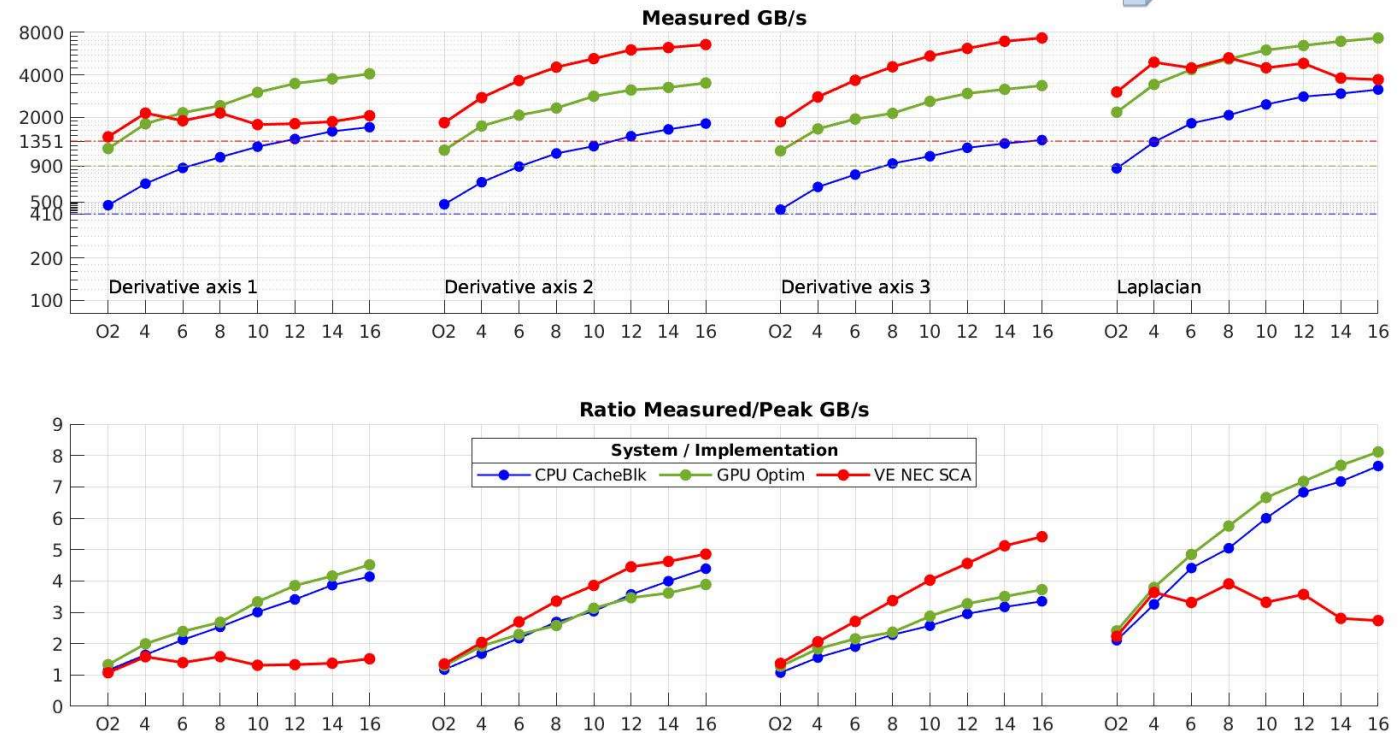
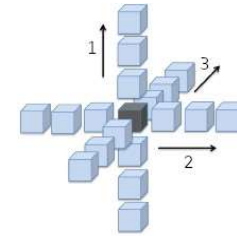
CPU: Cache blocking

GPU: Data prefetch in local (shared) memory

VE: NEC SCA library

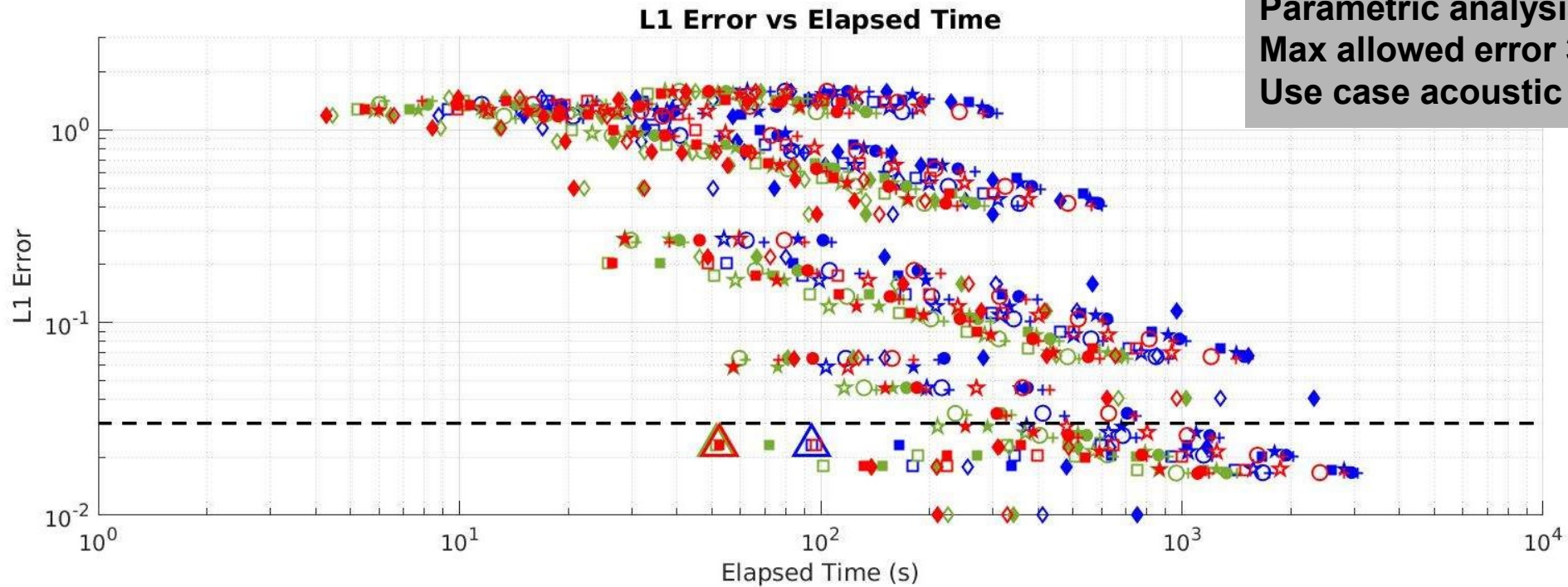
EAGE HPC 2021 workshop
Benchmark tool **hpcscan**
available on GitHub

Main compute kernel Finite difference operators



Each architecture has its specific performance 'signature'
How to select the optimum?

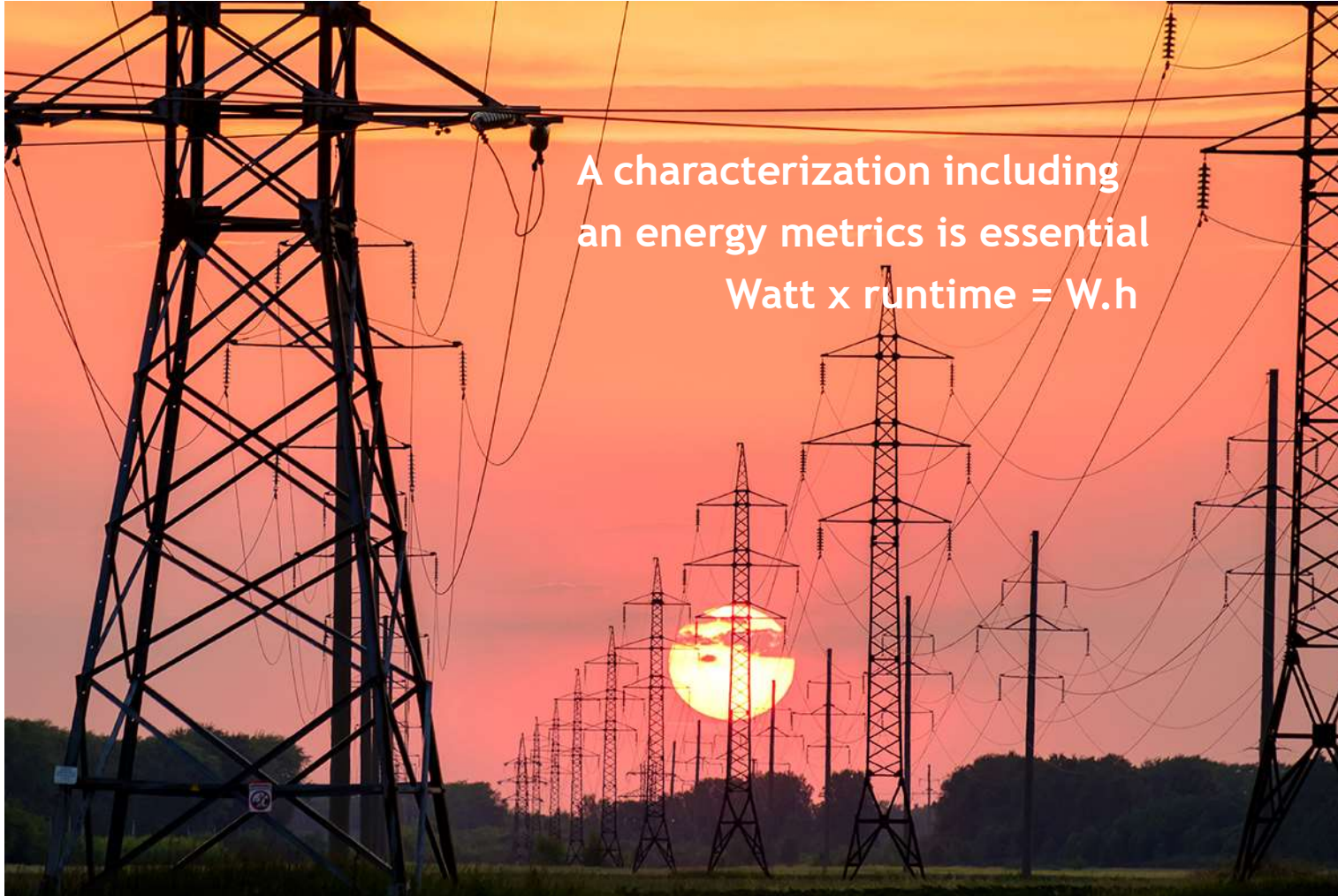
Traditional applications – Seismic imaging



	CPU	GPU	VE
Optimal algo.	Stand. O6	Stand. O6	Split O6
Optimal time (s)	94.4	51.4	52.5
Speed-up	1	1.8	1.8

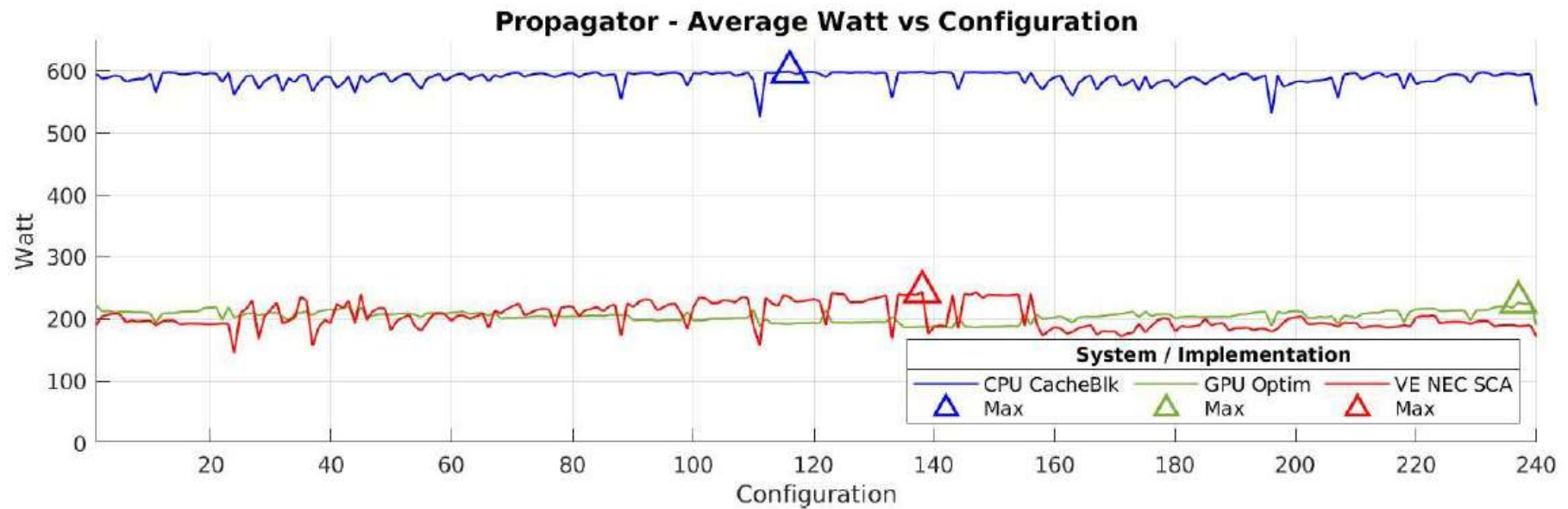
Optimal FD order 6
GPU & VE: 2x speedup vs CPU

Traditional applications – Seismic imaging

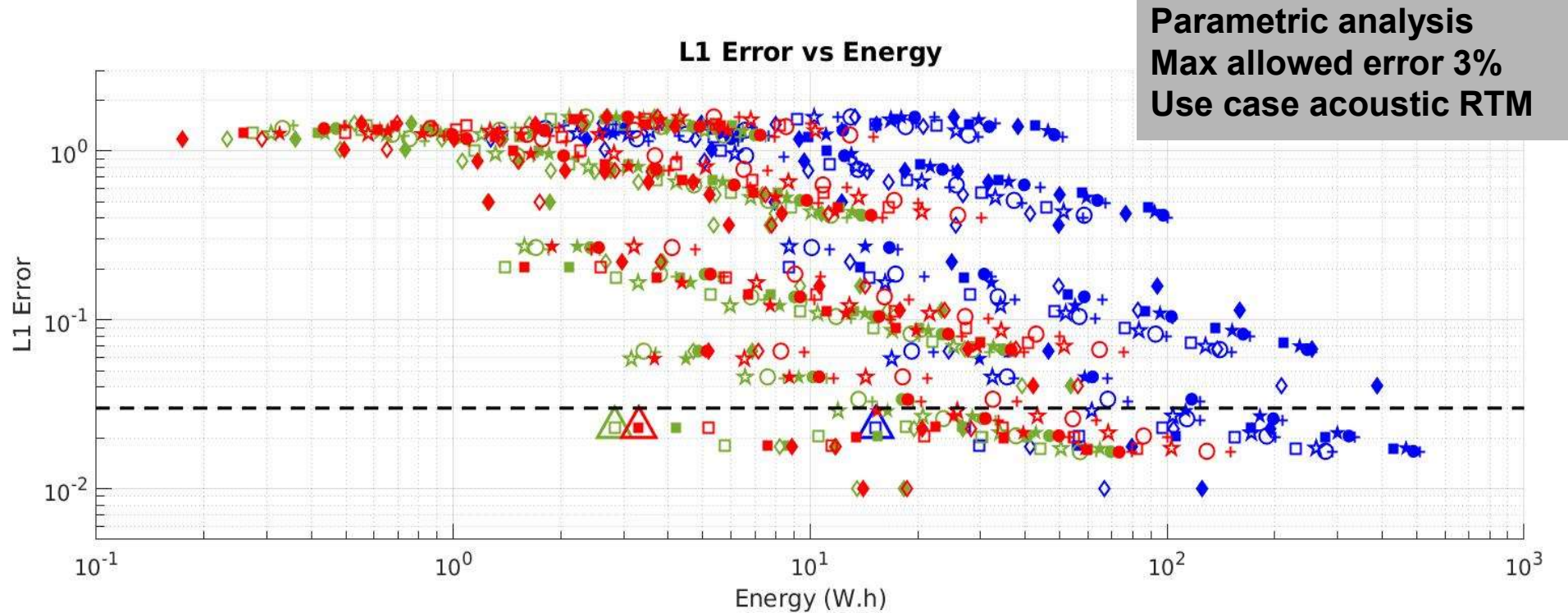


Traditional applications – Seismic imaging

Power consumption for seismic FD propagator
→ **VE & GPU about 3X less Watts than CPU**



Traditional applications – Seismic imaging



	CPU	GPU	VE
Optimal algo.	Stand. O6 N=500 0.1 CFL	Stand. O6 N=500 0.1 CFL	Split O6 N=500 0.1 CFL
Watt.Hour	15.24	2.83	3.31
Efficiency	1	5.4	4.6

**GPU & VE: Efficiency
Combines Runtime + Energy
About x5 vs CPU**

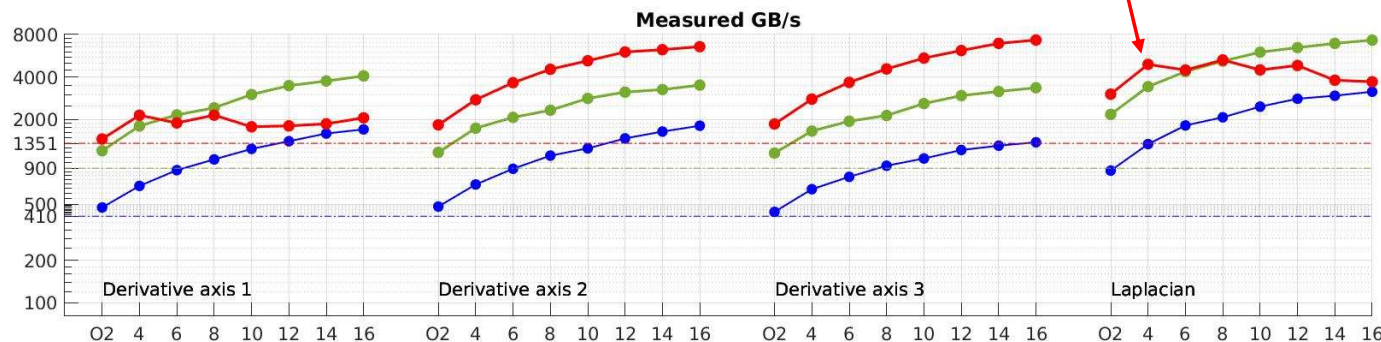
Traditional applications – Full waveform inversion

FWI (seismic inversion) shares common features with RTM (seismic imaging)

But discretizations are different

- RTM (high freq.) → **coarse grids** & high order stencils
- FWI (low freq.) → **dense grids** & low order stencils

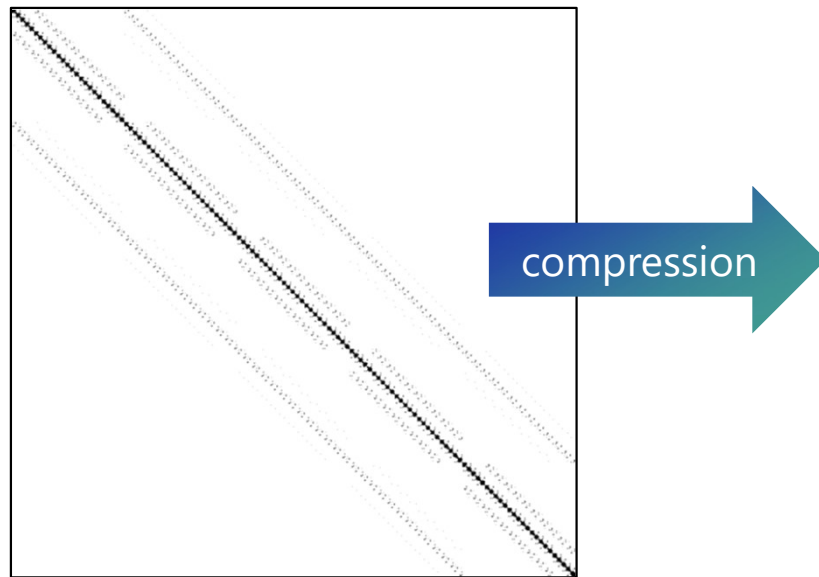
➡ Optimal FWI resolution 4 points / λ → suitable for FD O4 (VE performs best)



Traditional applications – Full waveform inversion

FWI can be formulated in the frequency domain

- Success of hierarchical approach from low to higher frequencies
- Mitigate local minimum attraction at vicinity of initial model
- In freq. domain, **direct (LU factorization) or iterative solvers can be used**



Example of acoustic impedance matrix

Collaborations with NEC

MUMPS Solver
mumps-consortium.org

Block Low-Rank
approximations to improve
multifrontal sparse solvers
with MUMPS consortium



Tile Low-Rank
approximations with
KAUST collaboration

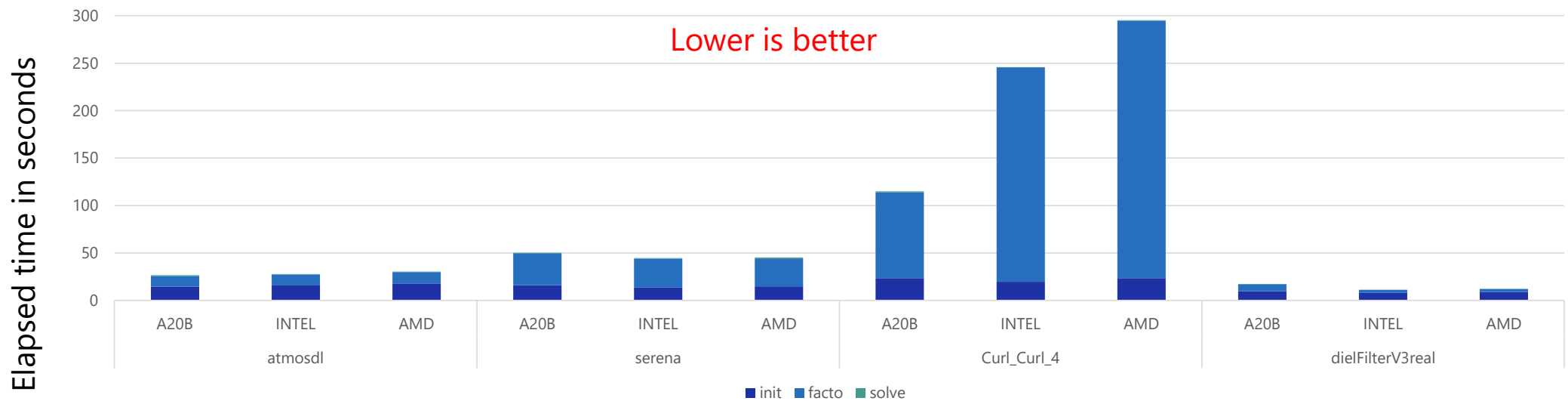
MUMPS Benchmark

◆ Direct solver for sparse linear systems

■ **Aurora VE version available in current version of MUMPS**

Solvers are also the main ingredients of reservoir simulator

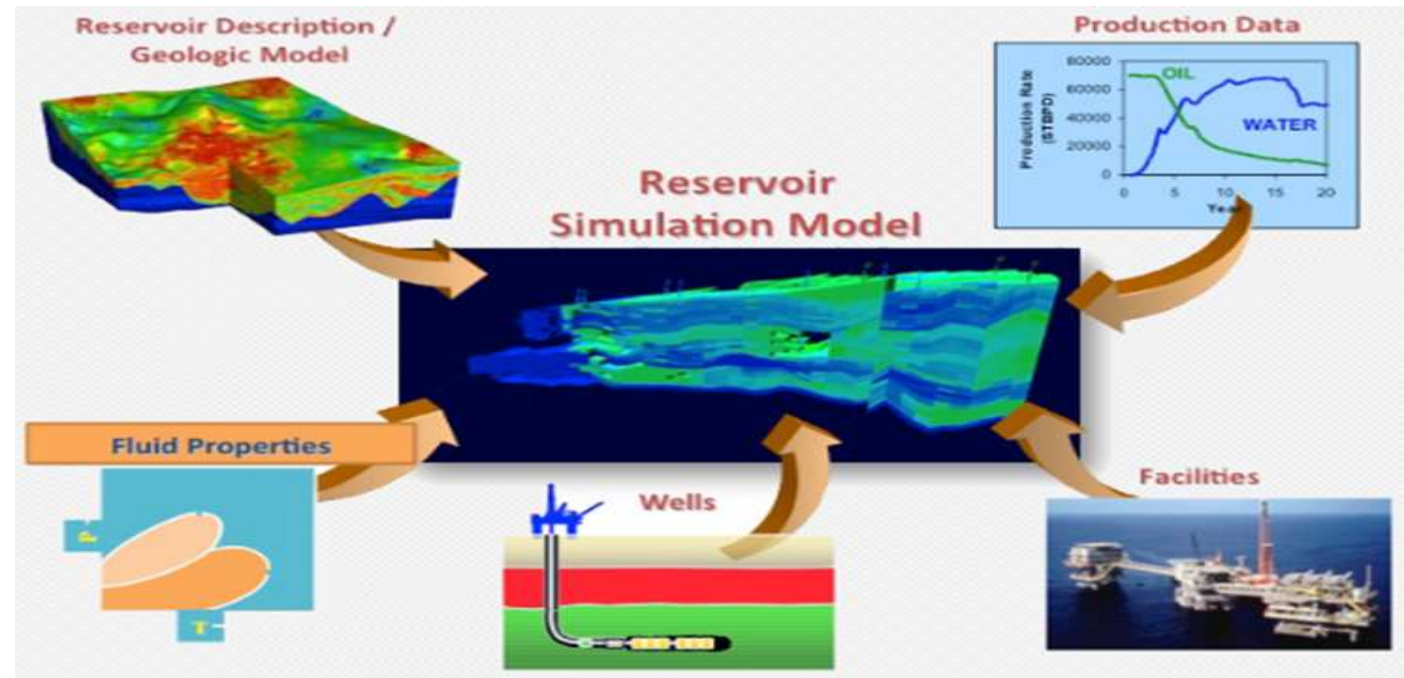
Comparison of the total execution time of the 3 phases (double precision + 7 RHS)



**1 VE AURORA 20B same performance as a server with 2 processors
And with much less power consumption!**

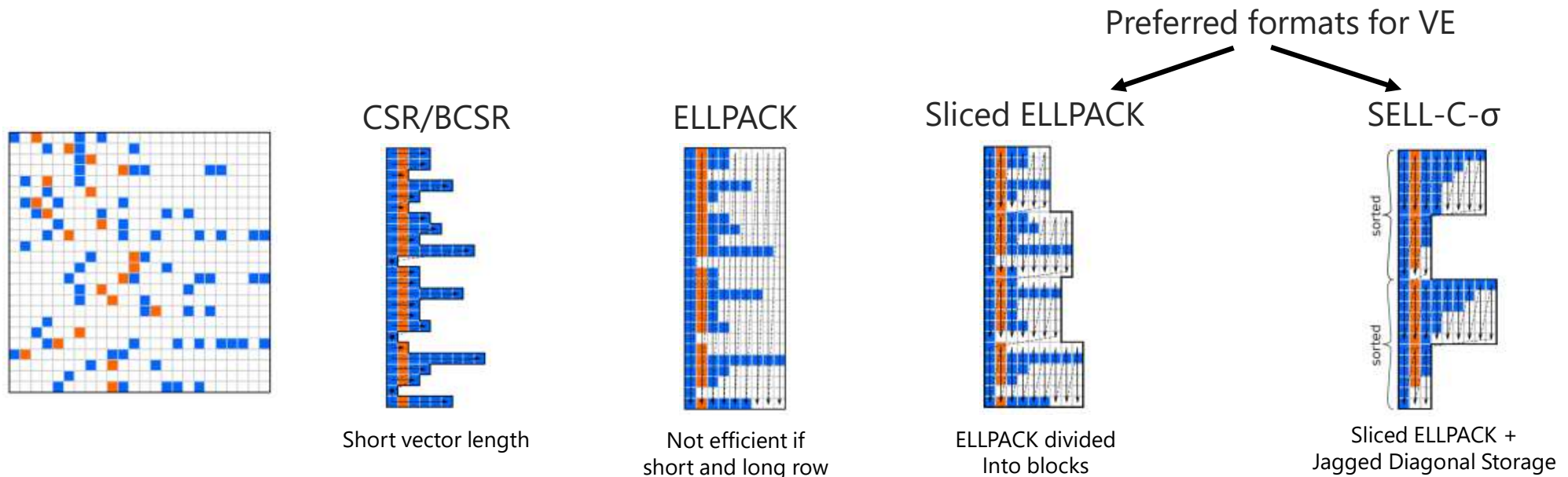
Traditional applications – Reservoir simulation

- ◆ Large models
 - Millions of cells
 - Large number of simulated years of production
- ◆ HPC KPIs
 - Memory intensive application
 - High pressure on the interconnect



Traditional applications – Reservoir simulation

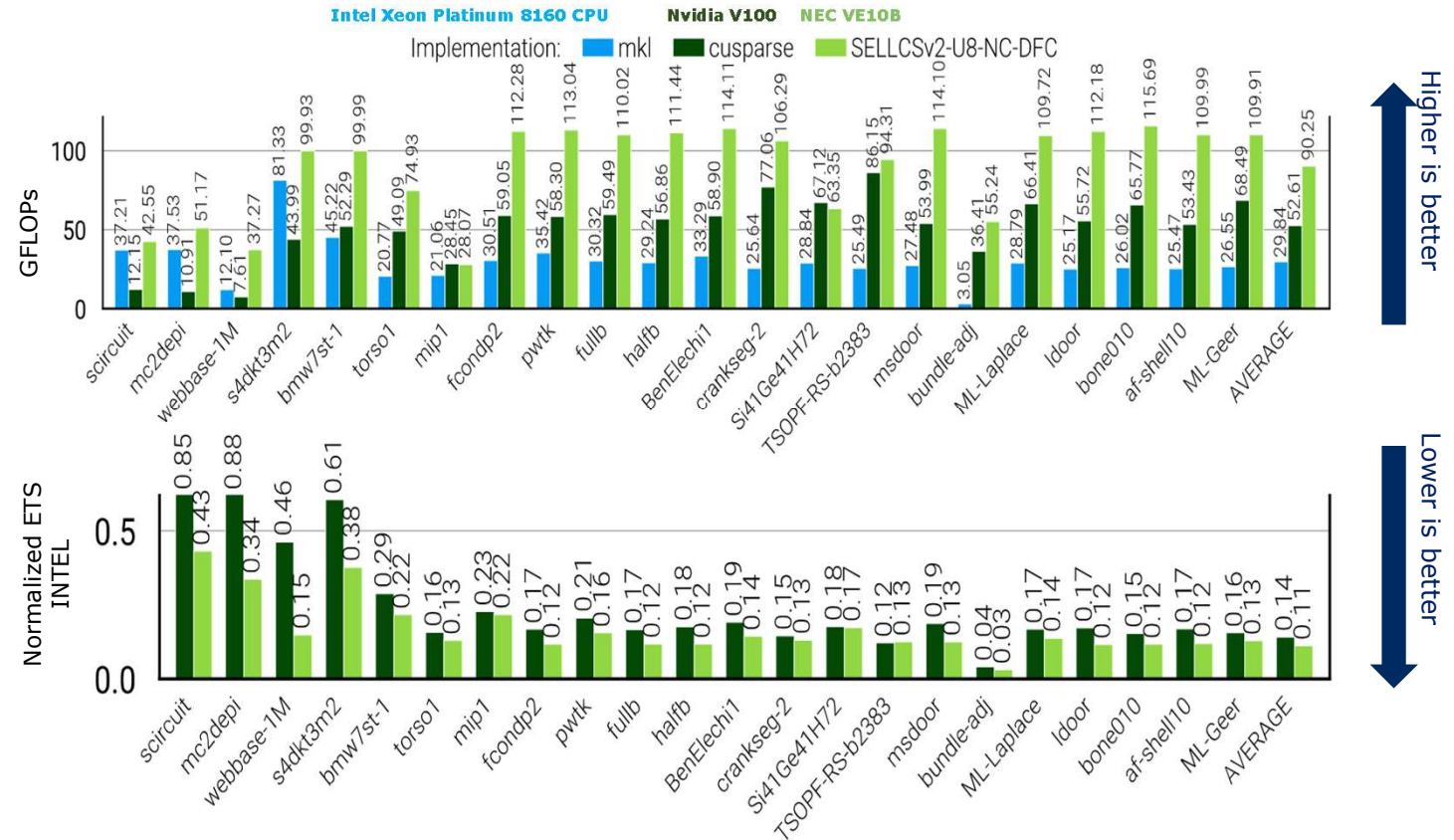
- ◆ Use sparse matrix as data structure
- ◆ **Main compute kernel is Sparse matrix-vector multiplication (SpMV) operation**
- ◆ SpMV performances depend on sparse matrix format and target architecture



Sparse Matrix Data Structures for High Performance Computing - <https://faculty.cc.gatech.edu/~echow/ipcc/hpc-course/sparsemat.pdf>

Traditional applications – Reservoir simulation

- ◆ Barcelona Supercomputer Center [BSC] technical report on SpMV performance comparison
- ◆ GFLOPS and Energy-to-Solution (ETS) shall be read simultaneously per each dataset
- ◆ **NEC always outperforms both Nvidia and Intel performance**
 - NEC VE's ETS is lower or in worst cases equals to Nvidia



Gómez, Constantino, M. Casas, F. Mantovani, and E. Focht. *Optimizing sparse matrix-vector multiplication in NEC SX-Aurora Vector Engine*. Technical Report, Barcelona Supercomputing Center, August 2020

New applications for O&G

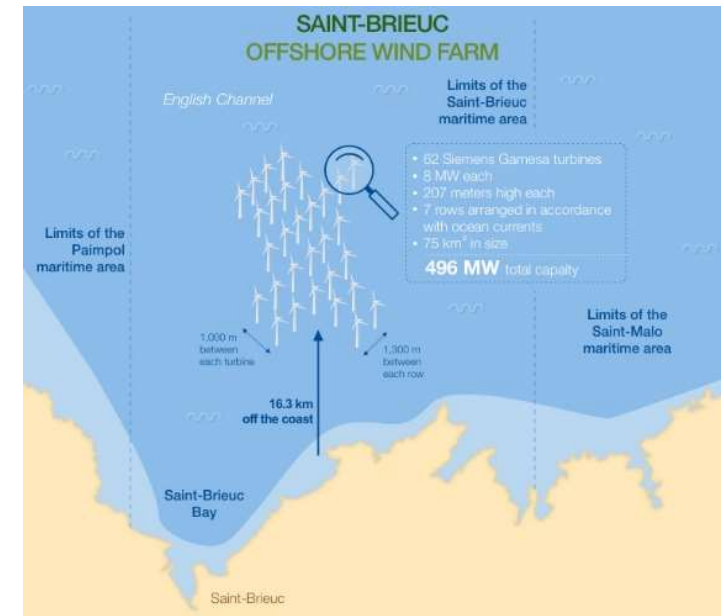
Wind turbine modeling

Carbone Capture and Storage

Optimization problems with quantum annealing

New applications – Wind turbine modeling

- ◆ According to International Energy Agency, **share of renewables** in global electricity production was **29% in 2020**
- ◆ **2/3 hydroelectricity**, and the rest **wind and solar energies**
- ◆ Considering **rapid growth of wind energy**, this resource is attracting significant interest from oil majors
- ◆ Oil companies have **experience with offshore oil rigs**. This is crucial to expedite development of offshore wind energy



New applications – Wind turbine modeling

- ◆ Has common roots with reservoir simulation as it involves **computational fluid dynamics (CFD)**
- ◆ Physical phenomena are distinct
 - In reservoirs, physics governed by flow of multiphase fluids in porous rocks
 - For wind turbine, **laws of aerodynamics** to compute flow of air (compressible fluid) around rotor blades of the turbine
- ◆ Complex problem when different interactions are taken into account, as for **floating offshore wind turbines**



New applications – Wind turbine modeling

- ◆ Several numerical schemes can be used to address this challenge

- 1st, **Blade Element Momentum Theory (BEMT)** used for long time in aerodynamics.

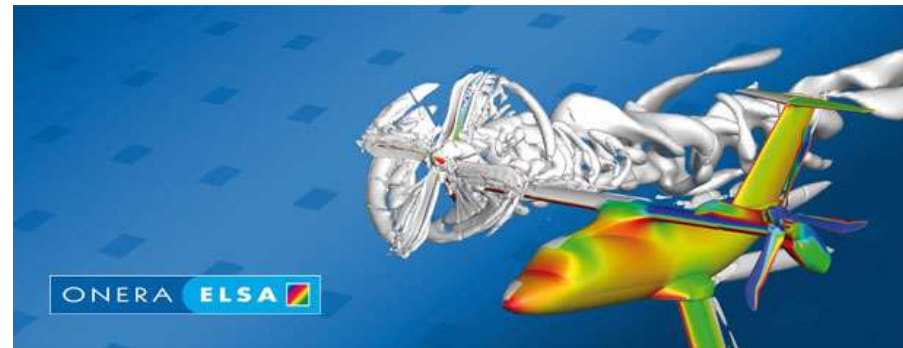
Reduces 3D problem into stationary Navier-Stokes 2D equations based on approximations

Ex: OpenFAST from US National Renewable Energy Laboratory (NREL)

- 2nd, **Reynolds Averaged Navier-Stokes (RANS)**

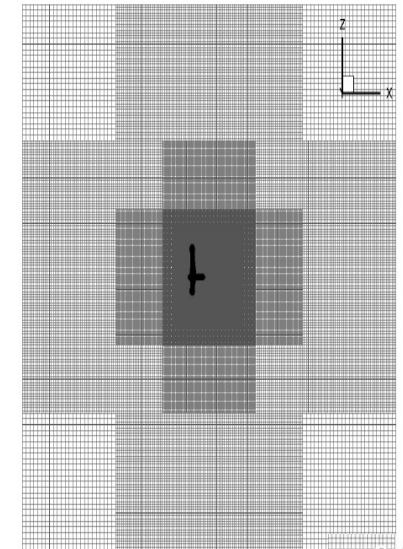
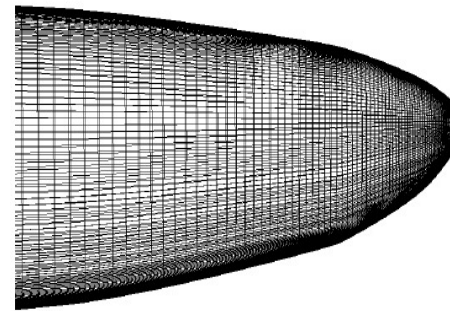
Solves 3D problem, hence more accurate but with higher computational cost

Ex: CFD package elsA from ONERA (French aerospace agency, former user of NEC SX systems)



New applications – Wind turbine modeling

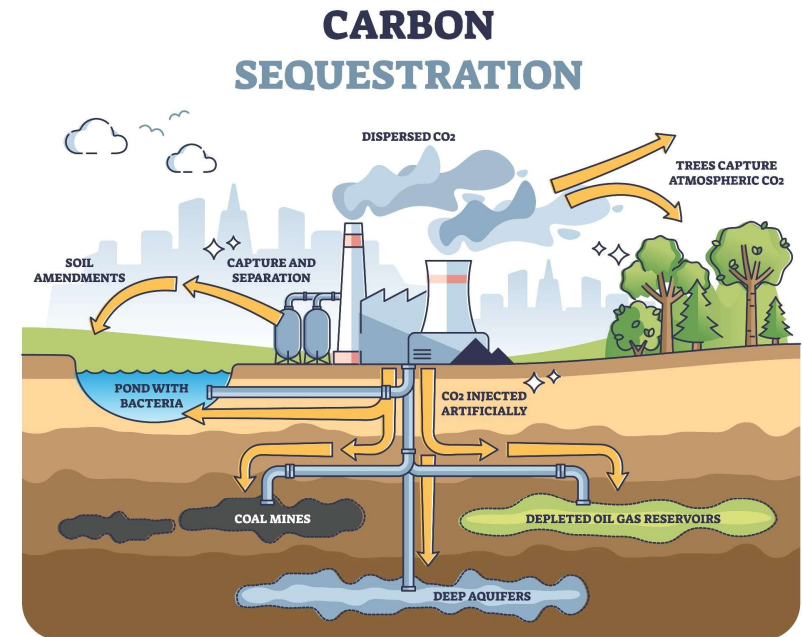
- ◆ elsA solves compressible RANS equations with **finite-volume cell-centered formulation** and **implicit time scheme**
- ◆ Mesh built with Chimera approach: curvilinear body-fitted grid built around rotor blades, embedded into set of Cartesian background grids
- ◆ From HPC point of view, elsA, and most CFD codes, **combine specificities of traditional O&G applications**
 - Finite-volume (i.e. modified 2nd-order FD) → **stencil-like operations** as in seismic imaging
 - Implicit time scheme → resolution of sparse linear system and **algebraic solvers** as in reservoir
 - **Expected to perform efficiently on VE** as it outperforms other architectures for low order stencils



elsA mesh (from Lienard et al., 2020)

New applications – Carbon Capture & Storage

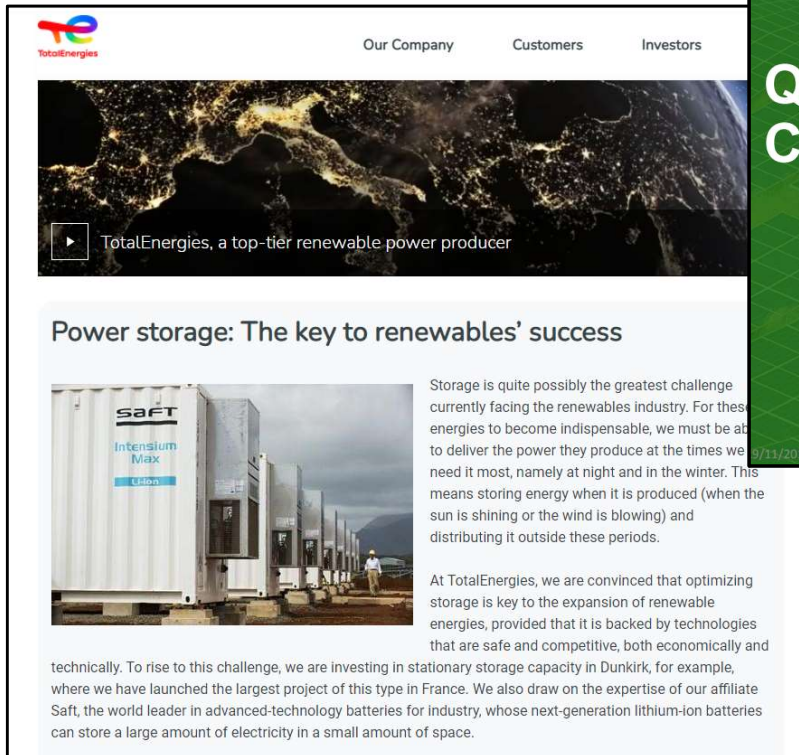
- ◆ Another **use of CFD** for **CCS modelling**
- ◆ Capturing CO₂ before it enters atmosphere, transporting & storing for very long time in deep geological formations
 - Compatible with large O&G production infrastructures in place
 - CO₂ injected for decades for various purposes, including enhanced oil recovery
 - But **long-term storage of CO₂ is new concept and requires further developments**



- ◆ Numerical methods for CCS modelling **similar to reservoir simulation** but with specificities
 - Complex fluid flow, thermal, and geomechanical effects, as implemented in GEOSX (Gross, 2021)
 - **We do not expect major different compute footprint** than traditional reservoir simulation

New applications – Optimization problems with quantum annealing

A wide range of applications for the energy distribution

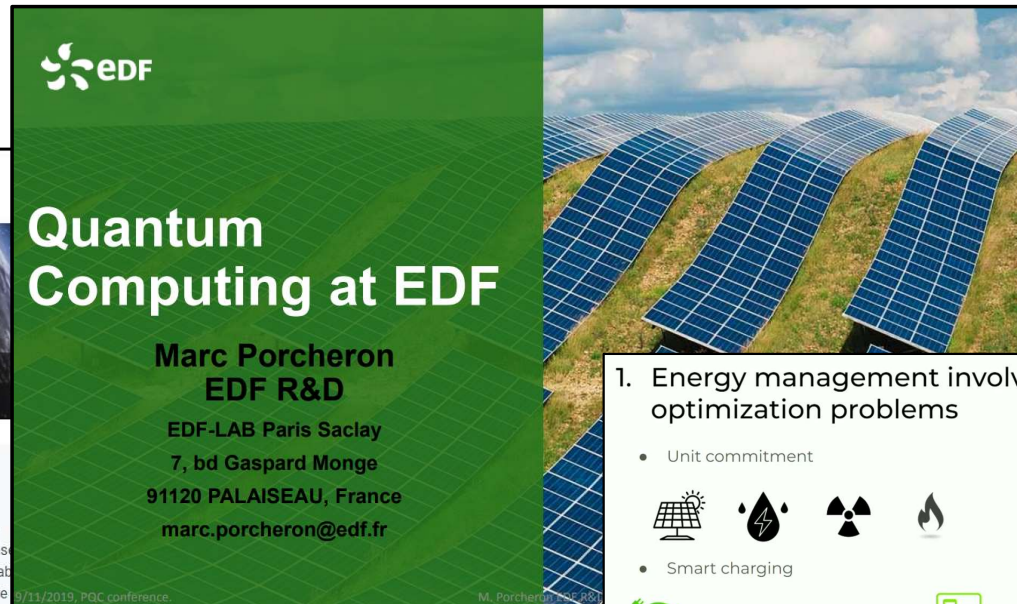


The screenshot shows the TotalEnergies website with a navigation bar (Our Company, Customers, Investors) and a main banner featuring a space image and the text "TotalEnergies, a top-tier renewable power producer". Below the banner is an article titled "Power storage: The key to renewables' success" which includes an image of Saft Intensium Max batteries and text discussing the challenges of energy storage and TotalEnergies' investment in stationary storage capacity in Dunkirk.

Power storage: The key to renewables' success

Storage is quite possibly the greatest challenge currently facing the renewables industry. For these energies to become indispensable, we must be able to deliver the power they produce at the times we need it most, namely at night and in the winter. This means storing energy when it is produced (when the sun is shining or the wind is blowing) and distributing it outside these periods.

At TotalEnergies, we are convinced that optimizing storage is key to the expansion of renewable energies, provided that it is backed by technologies that are safe and competitive, both economically and technically. To rise to this challenge, we are investing in stationary storage capacity in Dunkirk, for example, where we have launched the largest project of this type in France. We also draw on the expertise of our affiliate Saft, the world leader in advanced-technology batteries for industry, whose next-generation lithium-ion batteries can store a large amount of electricity in a small amount of space.



The slide features the EDF logo and a background image of solar panels. The title is "Quantum Computing at EDF". The speaker is Marc Porcheron, EDF R&D, located at EDF-LAB Paris Saclay, 7, bd Gaspard Monge, 91120 PALAISEAU, France, with email marc.porcheron@edf.fr.

Quantum Computing at EDF

Marc Porcheron
EDF R&D
EDF-LAB Paris Saclay
7, bd Gaspard Monge
91120 PALAISEAU, France
marc.porcheron@edf.fr

1. Energy management involves a lot of difficult optimization problems

- Unit commitment



- Smart charging

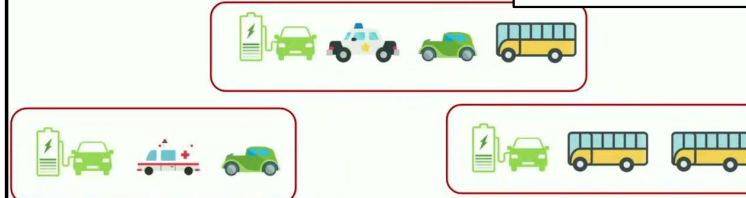
+ 160 000 VE (France, 2021)



50 000 charging station (France, 2021)

A big part of them are NP-hard!

Smart scheduling



- Minimize total completion time with priorities

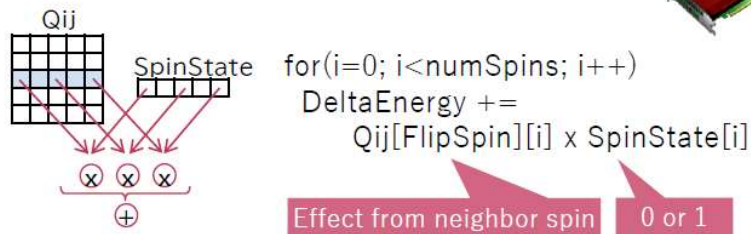
New applications – Optimization problems with quantum annealing

VA Performance is provided by:

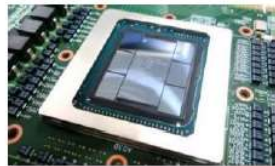
Matrix operation acceleration by VE, large and fast memory, and optimized algorithm for VE

Vector operation on VE

Energy calculation is matrix operation



Full connect 100k qbits/VE and high memory bandwidth

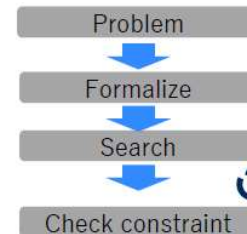
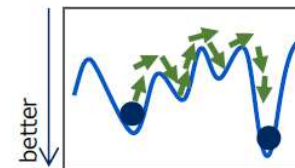


- 48GB memory capacity and 1.5TB/s memory bandwidth
- Multi card supports larger number of qbits (100k qbits x n)^{1/2}

Avoiding Redundant Search and Optimized algorithm for VE

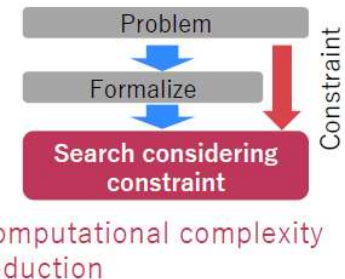
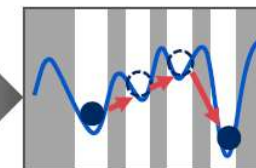
Existing search

Including constraint violations



VA search

skip constraint violations

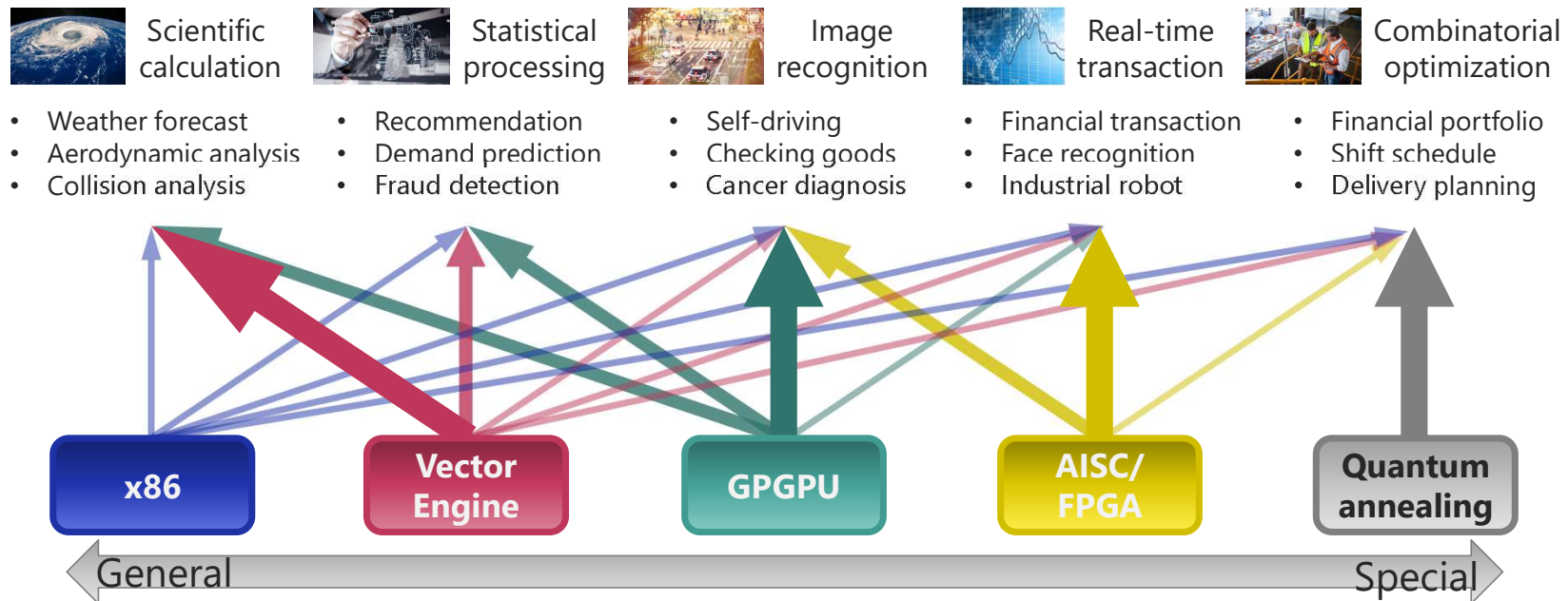


cf. talk from Deepak Pathania presented at the NEC Aurora Forum
Application of VA for O&G

Conclusions

Conclusions

- ◆ The energy sector has large needs for HPC systems
- ◆ This demand will increase for the energy transition
 - Variety of algorithms will increase, **heterogenous computing** is expected

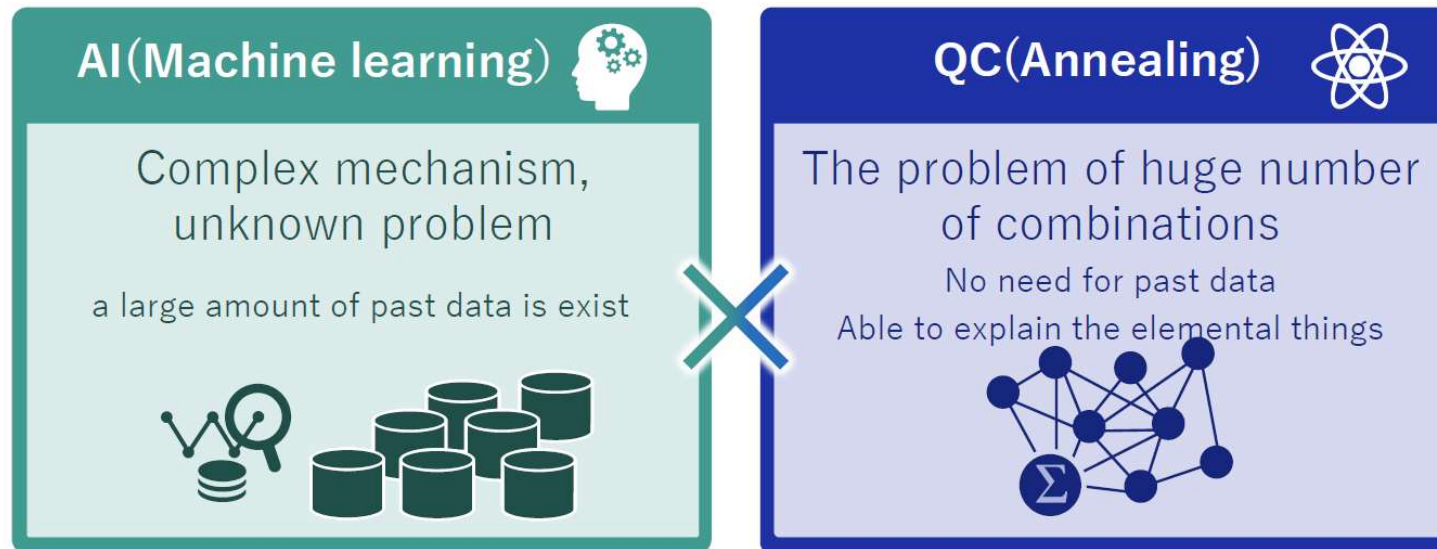


Not a single architecture optimal for all problems
Need for multi-architecture systems

Conclusions

- ◆ Applications with quantum annealing / computing just started
 - A lot of potentials
 - Hybrid algorithms QA / HPC or even QA / HPC / AI are to explore

Both AI and QA are enablers for solving problems that are difficult with conventional methods



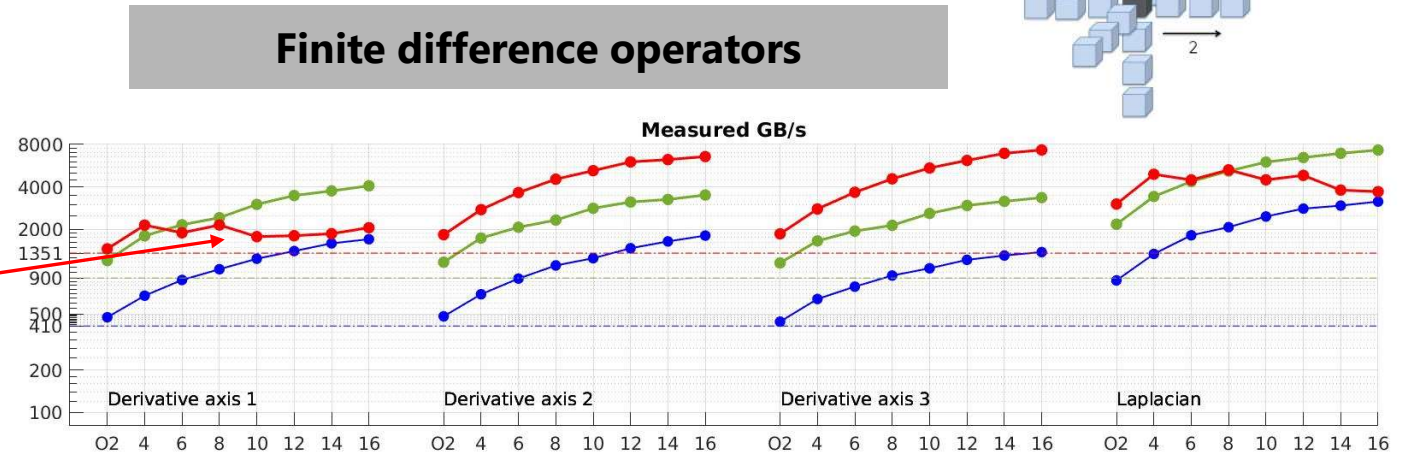
Conclusions

- ◆ NEC SX-Aurora Vector Engine is ready to address upcoming challenges
 - Easy and flexible porting (Native / Accelerator / Scalar modes)
 - Fruitful co-design projects with customers allow to identify critical enhancements

Example of co-design outcome
for seismic imaging

Optimization scheduled for
Aurora Generation 3

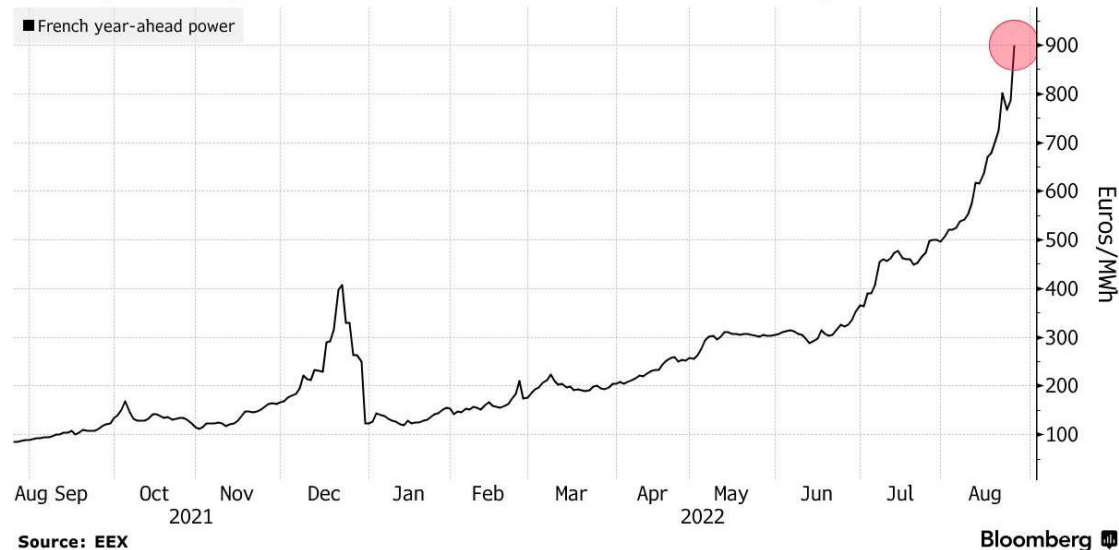
Introduction to new instruction
to increase performance on
axis 1



Conclusions

- ◆ NEC SX-Aurora Vector Engine is ready to address the most critical challenge for HPC: energy
 - **High ratio computing efficiency / energy consumption**

Power prices surged to a fresh record on further nuclear outages



Price of electricity in France (bloomberg.com)
10X in one year

June 2022 (top500)
For very first time, **EFlop/s is reached** (10^{18})
Frontier (DOE, ONL US)
1.1 EFlop/s with 21.1 MW

With current price in France

- 1 day → 456 000 € (455 000 USD)
- 1 year → 166 M€ (165.7 M USD)

Thank you very much for your attention

For questions or to get in touch contact me at **vincent.etienne@emea.nec.com**

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