Breaking the memory barrier (for finite difference modeling)

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Overview

- 1. Finite Difference Modeling
- 2. The Memory Barrier
- 3. Breaking the Memory Barrier
- 4. Results
- 5. Conclusions

- ▶ Model physical phenomena, e.g. wave propagation.
- Bigger models and more calculations allow higher accuracy.
- Overwhelming memory requirements for 3D.
- Circumvent the memory limitations.





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Model a given differential equation, e.g.:

$$\frac{\partial v_i}{\partial t} = \frac{1}{\rho} \left(\frac{\partial \tau_{ij}}{\partial j} \right)$$
(1)
$$\frac{\partial \tau_{ij}}{\partial t} = \delta_{ij} \lambda \frac{\partial v_k}{\partial k} + \mu \left(\frac{\partial v_i}{\partial j} + \frac{\partial v_j}{\partial i} \right)$$
(2)

- ► 3-dimensional array for each variable.
- Approximate derivatives by weighted sums.
- Update each value across a small Δt .

A FDM example

$$\frac{\partial}{\partial x}u_{i+\frac{1}{2},j,k}\approx$$

0.0038 $u_{i-3,j,k}$ -0.0211 $u_{i-2,j,k}$ +0.1049 $u_{i-1,j,k}$ -1.2327 $u_{i,j,k}$ +1.2327 $u_{i+1,j,k}$ -0.1049 $u_{i+2,j,k}$ +0.0211 $u_{i+3,j,k}$ -0.0038 $u_{i+4,j,k}$



$$\tau_{xy}^{n+\frac{1}{2}} = \tau_{xy}^{n-\frac{1}{2}} + \mu \Delta t \left(\frac{\partial}{\partial x} v_y^n + \frac{\partial}{\partial y} v_x^n \right)$$

Model the full frequency spectrum of a seismic shot, in a fully anisotropic medium, in an area corresponding to a single shot:

- Wavelengths of 10m.
- ► 4m×4m×4m cells.
- $1000 \times 1000 \times 500$ grid cells.
- ▶ 24GB of data. (61GB anisotropic.)

Split the model in smaller parts:



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The Memory Barrier – Data Transfer Slowdown

Moving the data to/from RAM becomes a bottleneck:

Work	Speed	Time
24 imes 2GB	50MB/s	1000s
$1.5\cdot 10^{11}$ flop	10 ⁹ flop/s	150s

- An 8-fold time increase is not acceptable.
- We want IO time \leq CPU time.

1 Correctly calculate all cell updates.

2 Increase the CPU vs IO ratio.

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- 2 Increase the CPU vs IO ratio. Calculate several time steps per sweep.

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- ► C = 200 and L = 4 gives 12 time steps per sweep, or 6 time steps per read/write.
- Almost the same as the IO / CPU ratio.
- Only $200 \times 200 \times 500 \times 12 \times 4 \approx 1$ GB.

- ▶ 1.25GB memory.
- Cycles of 7s CPU 8s IO.
- 150s per time step.
- ▶ 20 000 time steps: 5 weeks on the laptop.

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 - Time usage proportional to f^4 and v_{min}^{-4} .
 - Corresponds to 3h on a single GPU!

Conclusions

- Large scale FDM possible without large memory requirements.
- Computational speed is still an issue.
- Need to maintain the CPU vs IO ratio. Factors:
 - ► Faster CPU.
 - Faster IO.
 - More memory.
- Robust alternative to parallellisation when several modelings are needed.
- The only cost is code complexity!

Considerations

- Smaller steps halve the memory requirements.
- Eliminate the IO wait with asynchronous IO.
- Share static data within nodes.
- Use GPUs with data streamed from RAM.

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