

My background

- PhD from University of Oslo (2010)
- Associate professor (Førsteamanuensis) at OsloMet
- Researcher at SINTEF (currently on 80% leave)
- Previously lectured at NITH and UiO.
- Have taught computer graphics and geometry, multivariate calculus, algorithms and data structures, GPU computing, ...







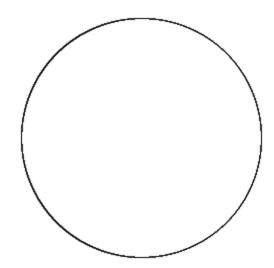
- Established 1950 by the Norwegian Institute of Technology.
- The largest independent research organisation in Scandinavia.
- A non-profit organisation.
- Motto: "Technology for a better society".
- Key Figures*
 - 2100 Employees from 70 different countries.
 - 73% of employees are researchers.
 - 3 billion NOK in turnover (about 360 million EUR / 490 million USD).
 - 9000 projects for 3000 customers.
 - Offices in Norway, USA, Brazil, Chile, and Denmark.



Todays "popular science" seminar

- Desktop parallel computing: What are GPUs, and why care about them?
- Things to consider when designing parallell code
- (Maybe) Videos
- Live coding session Python and GPUs for the wave equation









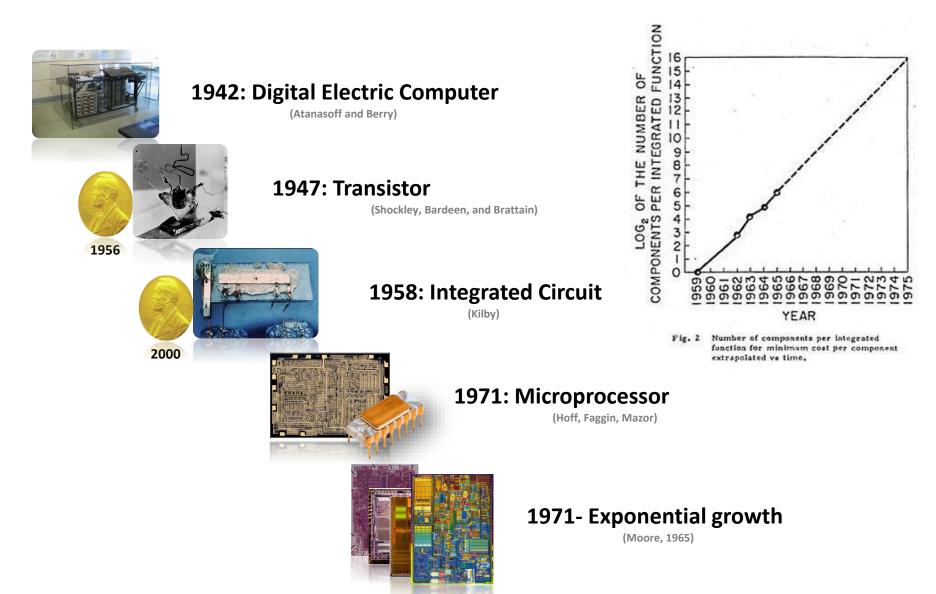
Heterogeneous Computing Group

http://hetcomp.com



Desktop parallell computing

History lesson: development of the microprocessor 1/2





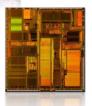
History lesson: development of the microprocessor 2/2



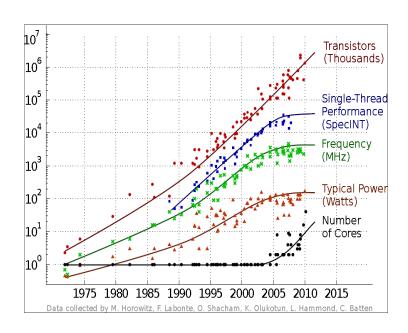
1971: 4004, 2300 trans, 740 KHz



1982: 80286, 134 thousand trans, 8 MHz

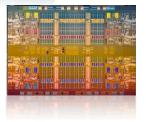


1993: Pentium P5, 1.18 mill. trans, 66 MHz





2000: Pentium 4, 42 mill. trans, 1.5 GHz

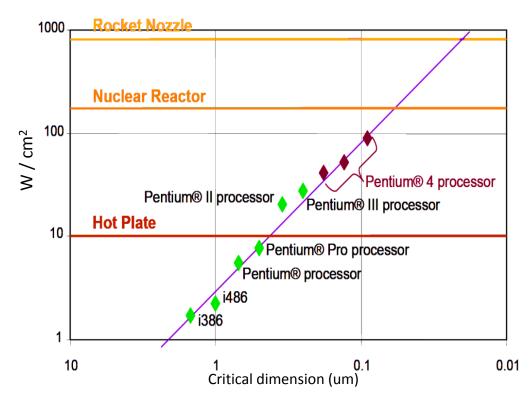


2010: Nehalem
2.3 bill. Trans, 8 cores, 2.66 GHz



Power density in CPUs

- Heat density approaching that of nuclear reactor core: Power wall
- Traditional cooling solutions (heat sink + fan) insufficient
- Industry solution: multi-core and parallelism!



Graph taken from G. Taylor, "Energy Efficient Circuit Design and the Future of Power Delivery" EPEPS'09



Why Parallelism?

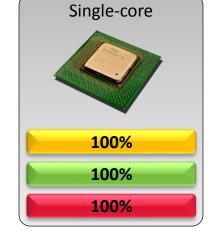
Frequency

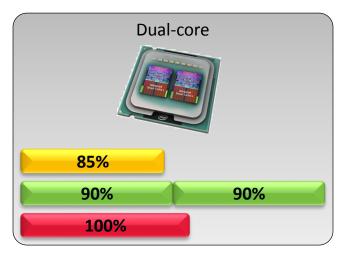
Power

Performance

The power density of microprocessors is proportional to the clock frequency cubed:¹

$$P_d \propto f^3$$

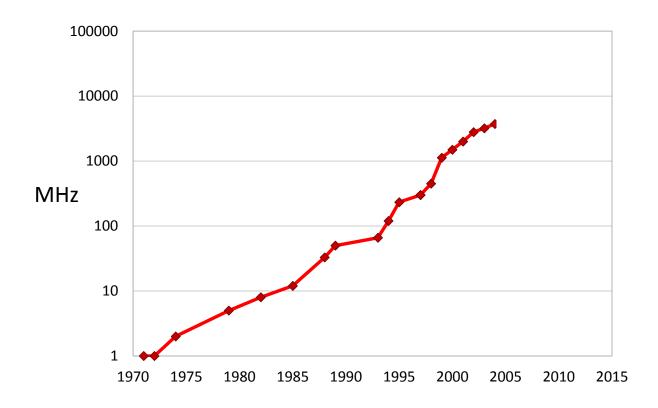




¹ Brodtkorb et al. State-of-the-art in heterogeneous computing, 2010

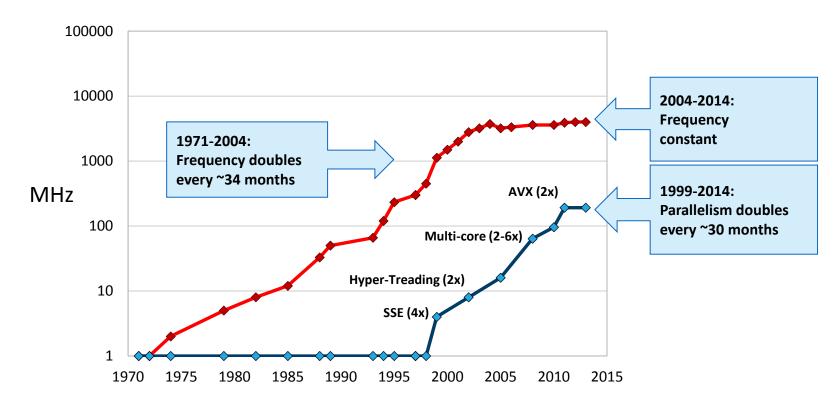


Frequency scaling





Frequency scaling



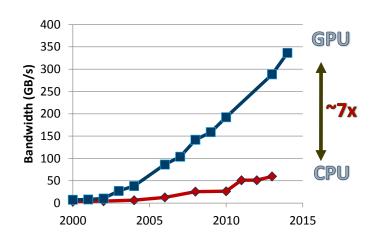
- 1970-2004: Frequency doubles every 34 months (Moore's law for performance)
- 1999-2014: Parallelism doubles every 30 months



Massive Parallelism: The Graphics Processing Unit

• Up-to <u>5760</u> floating point operations in parallel!

 5-10 times as power efficient as CPUs!



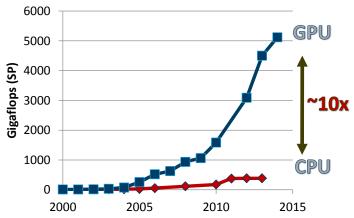








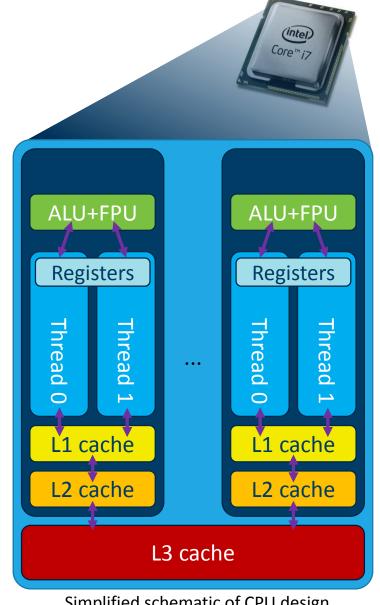






Multi-core CPU architecture

- A single core
 - L1 and L2 caches
 - 8-wide SIMD units (AVX, single precision)
 - 2-way Hyper-threading (<u>hardware</u> threads) When thread 0 is waiting for data, thread 1 is given access to SIMD units
 - Most transistors used for cache and logic
- Optimal number of FLOPS per clock cycle:
 - 8x: 8-way SIMD
 - 6x: 6 cores
 - 2x: Dual issue (fused mul-add / two ports)
 - Sum: 96!

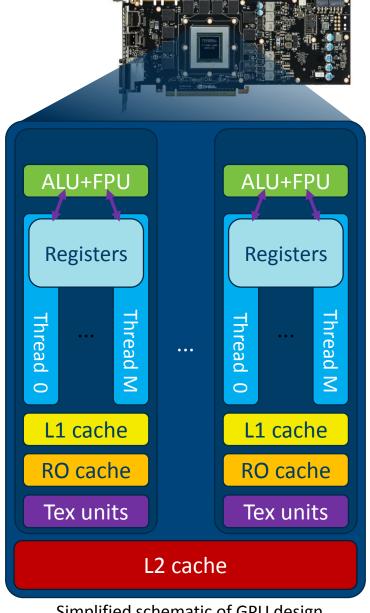


Simplified schematic of CPU design



Many-core GPU architecture

- A single core (Called streaming multiprocessor, SMX)
 - L1 cache, Read only cache, texture units
 - <u>Six</u> 32-wide SIMD units (192 total, single precision)
 - Up-to 64 warps simultaneously (<u>hardware</u> warps) Like hyper-threading, but a warp is 32-wide SIMD
 - Most transistors used for floating point operations
- Optimal number of FLOPS per clock cycle:
 - 32x: 32-way SIMD
 - 2x: Fused multiply add
 - 6x: Six SIMD units per core
 - 15x: 15 cores
 - Sum: 5760!

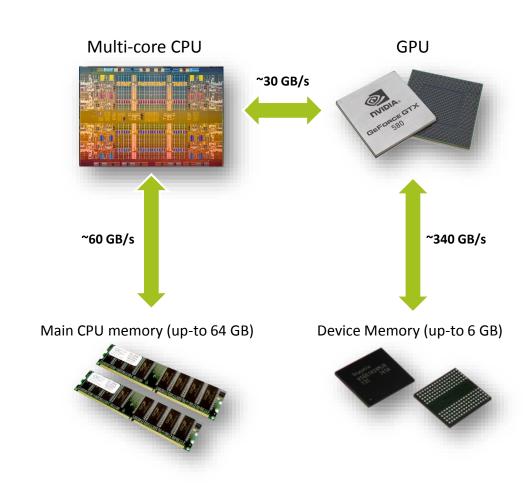


Simplified schematic of GPU design



Heterogeneous Architectures

- Discrete GPUs are connected to the CPU via the PCI-express bus
 - Slow: 15.75 GB/s each direction
 - On-chip GPUs use main memory as graphics memory
- Device memory is limited but fast
 - Typically up-to 6 GB
 - Up-to 340 GB/s!
 - Fixed size, and cannot be expanded with new dimm's (like CPUs)

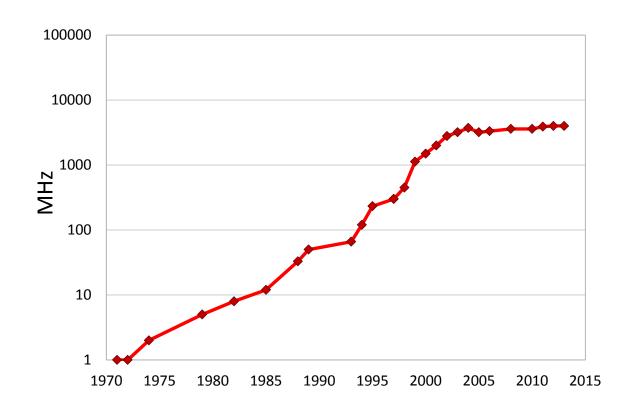






Parallell algorithm design

The beach law





Performance doubles every 18 months

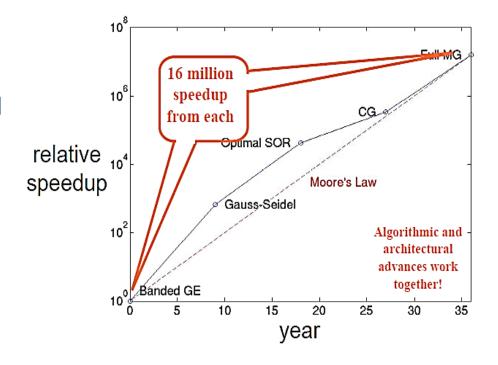
– go to the beach and relax



Algorithm design and computer architecture

• The key to increasing performance, is to consider the full algorithm and architecture interaction.

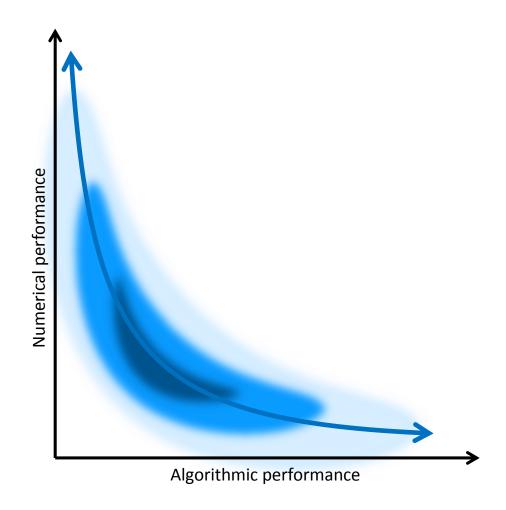
 A good knowledge of <u>both</u> the algorithm <u>and</u> the computer architecture is required.





Algorithmic and numerical performance

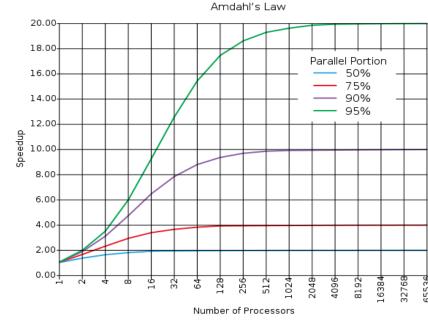
- Total performance is the product of algorithmic and numerical performance
- Your mileage may vary: algorithmic performance is highly problem dependent
- Many algorithms have low numerical performance
- Need to consider both the algorithm and the architecture for maximum performance





Limits on performance 1/4

- Most algorithms contains a mixture of work-loads:
 - Some serial parts
 - Some task and / or data parallel parts
- Amdahl's law:
 - There is a limit to speedup offered by parallelism
 - Serial parts become the bottleneck for a massively parallel architecture!
 - Example: 5% of code is serial: maximum speedup is 20 times!



$$S(N) = \frac{1}{(1-P) + \frac{P}{N}}$$

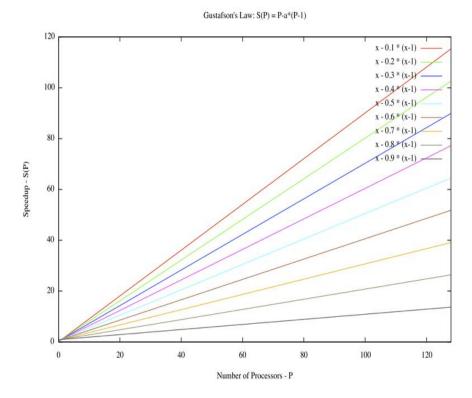
S: Speedup

P: Parallel portion of code

N: Number of processors

Limits on performance 2/4

- Gustafson's law:
 - If you cannot reduce serial parts of algorithm, make the parallel portion dominate the execution time
 - Essentially: solve a bigger problem!



$$S(P) = P - \alpha \cdot (P - 1).$$

S: Speedup

P: Number of processors

 α : Serial portion of code

Limits on performance 3/4

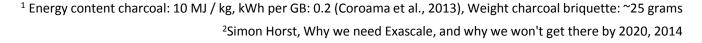
Moving data has become the major bottleneck in computing.

• Downloading 1GB from Japan to Switzerland consumes roughly the energy of 1 charcoal briquette¹.



A FLOP costs less than moving one byte².

• Key insight: <u>flops are free</u>, <u>moving data is expensive</u>

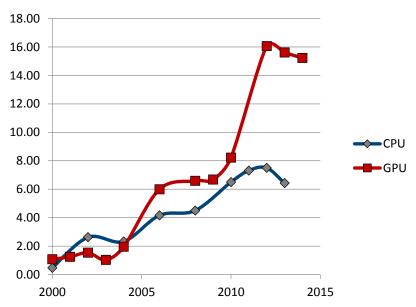




Limits on performance 4/4

- A single precision number is four bytes
 - You must perform <u>over 60 operations</u> for each float read on a GPU!
 - Over 25 operations on a CPU!
- This groups algorithms into two classes:
 - Memory bound
 Example: Matrix multiplication
 - Compute bound Example: Computing π
- The third limiting factor is latencies
 - Waiting for data
 - Waiting for floating point units
 - Waiting for ...

Optimal FLOPs per byte (SP)







Conservation laws

Conservation laws

- Conserve synonyms: keep up, maintain, preserve, save
- Examples: preserve amount of water, preserve total energy in a system, ...
- Can be formulated as partial differential equations





Example: The linear wave equation

$$\frac{\partial^2 u}{\partial t^2} = c^2 \, \nabla^2 u$$

- Can describes vibration of string (in 1D)
- u is the deflection of the string
- c is a material property (related to wave propagation speed)







Example: The 2D wave equation



$$\frac{\partial^2 u}{\partial t^2} = c \left[\frac{\partial^2 u}{\partial x^2} + \frac{\partial^2 u}{\partial y^2} \right]$$



$$\frac{1}{\Delta t^2} (u_{i,j}^{n+1} - 2u_{i,j}^n + u_{i,j}^{n-1})
= \frac{c}{\Delta x^2} (u_{i-1,j}^n - 2u_{i,j}^n + u_{i+1,j}^n) + \frac{c}{\Delta y^2} (u_{i,j-1}^n - 2u_{i,j}^n + u_{i,j+1}^n)$$

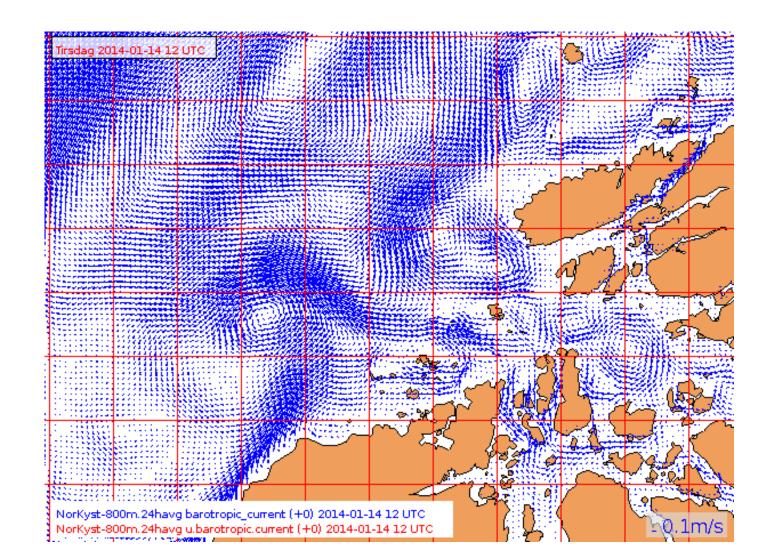






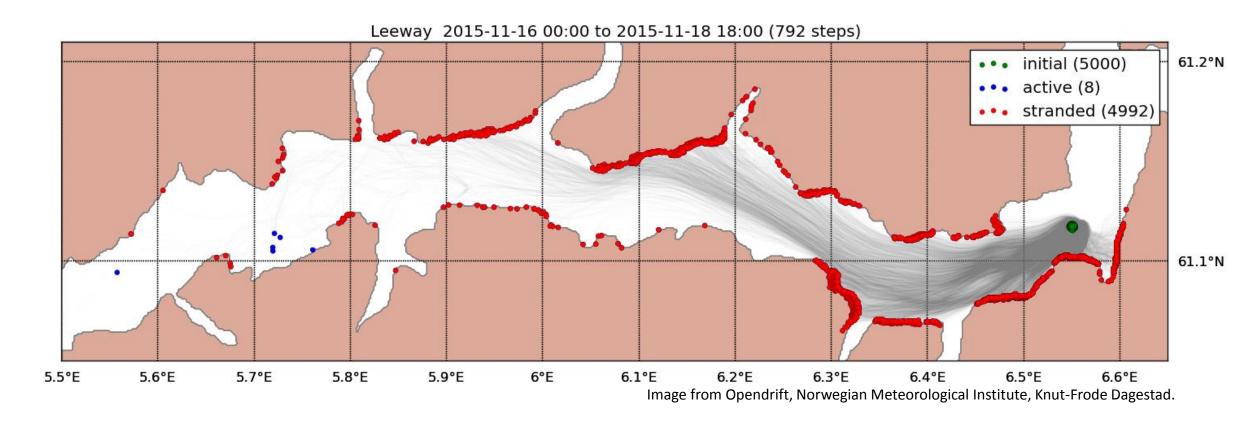
Using conservation laws in real life

Problem statement



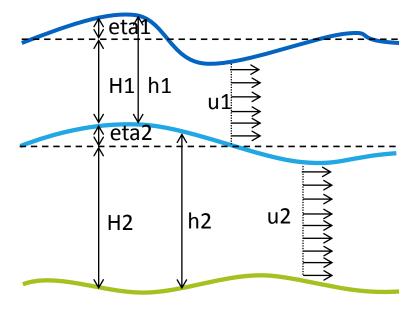


Solution strategy





2-layer non-linear scheme



- 1 layer model extendible to more layers
 - Ocean can be modeled as a stratisfied medium with multiple homogeneous layers
- Multiple layers enables baroclinic response from model

1 layer scheme, non-linear FD

$$\begin{split} &\eta_{jk}^{n+1} &= \eta_{jk}^{n-1} - \frac{2\Delta t}{\Delta x} \left(U_{jk}^{n} - U_{j-1k}^{n} \right) - \frac{2\Delta t}{\Delta y} \left(V_{jk}^{n} - V_{jk-1}^{n} \right), \\ &V_{jk}^{n+1} &= \frac{1}{C_{jk}^{v}} \left[V_{jk}^{n-1} + 2\Delta t \left(- f \overline{U}_{jk}^{n} + \frac{N_{jk}^{v}}{\Delta y} + \frac{P_{jk}^{v} + \hat{P}_{jk}^{v}}{\Delta y} + Y_{jk-1}^{n+1} + AE_{jk}^{v} \right) \right] \\ &C_{jk}^{v} &= 1 + \frac{2R\Delta t}{H_{jk}^{v}} + \frac{2A\Delta t (\Delta x^{2} + \Delta y^{2})}{\Delta x^{2} \Delta y^{2}}, \\ &N_{jk}^{v} &= \frac{1}{4} \left\{ \frac{\left(V_{jk+1}^{n} + V_{jk}^{n} \right)^{2}}{H_{jk+1}^{n} + \eta_{jk+1}^{n}} - \frac{\left(V_{jk}^{n} + V_{jk-1}^{n} \right)^{2}}{H_{jk}^{n} + \eta_{jk}^{n}} + \frac{\Delta y}{\Delta x} \left[\frac{\left(U_{jk+1}^{n} + U_{jk}^{n} \right) \left(V_{j+1k}^{n} + V_{jk}^{n} \right)}{H_{jk}^{n} + \eta_{jk}^{n}} - \frac{\left(U_{j-1k+1}^{n} + U_{j-1k}^{n} \right) \left(V_{jk+1}^{n} + V_{j-1k}^{n} \right)}{H_{jk-1}^{n} + \eta_{jk-1}^{n}} \right] \right\}, \\ &P_{jk}^{v} &= gH_{jk}^{v} \left(\eta_{jk+1}^{n} - \eta_{jk}^{n} \right), \quad \hat{P}_{jk}^{v} &= \frac{1}{2} \left[\left(\eta_{jk+1}^{n} \right)^{2} - \left(\eta_{jk}^{n} \right)^{2} \right], \\ &E_{jk}^{v} &= \frac{1}{\Delta x^{2}} \left(V_{j+1k}^{n} - V_{jk}^{n-1} + V_{j-1k}^{n} \right) + \frac{1}{\Delta y^{2}} \left(V_{jk+1}^{n} - V_{jk}^{n-1} + V_{jk-1}^{n} \right). \\ &U_{jk}^{n+1} &= \frac{1}{C_{jk}^{v}} \left[U_{jk}^{n-1} + 2\Delta t \left(f \overline{V}_{jk}^{n} + \frac{N_{jk}^{v}}{\Delta x} + \frac{P_{jk}^{v} + \hat{P}_{jk}^{v}}{\Delta x} + X_{jk}^{n+1} + AE_{jk}^{v} \right) \right], \\ &C_{jk}^{v} &= 1 + \frac{2R\Delta t}{H_{jk}^{v}} + \frac{2A\Delta t \left(\Delta x^{2} + \Delta y^{2} \right)}{\Delta x^{2} \Delta y^{2}}, \\ &N_{jk}^{v} &= \frac{1}{4} \left\{ \frac{\left(U_{j+1k}^{n} + U_{jk}^{n} \right)^{2}}{H_{j+1k} + \eta_{jk}^{n}} - \frac{\left(U_{jk}^{n} + U_{j-1k}^{n} \right)^{2}}{H_{jk} + \eta_{jk}^{n}}} + \frac{\Delta x}{\Delta y} \left[\frac{\left(U_{j+1k}^{n} + U_{jk}^{n} \right)^{2}}{H_{jk} + \eta_{jk}^{n}} - \frac{\left(U_{jk}^{n} + U_{j-1k}^{n} \right)^{2}}{H_{jk-1}^{n} + \eta_{jk-1}^{n}}} \right] \right\}, \quad (23) \\ &P_{jk}^{v} &= gH_{jk}^{v} \left(\eta_{j+1k}^{n} - \eta_{jk}^{n} \right), \quad \hat{P}_{jk}^{v} &= \frac{1}{2} \left[\left(\eta_{j+1k}^{n} - U_{jk}^{n+1} + U_{jk-1}^{n} \right) \right], \\ &E_{jk}^{v} &= \frac{1}{\Delta x^{2}} \left(U_{j+1k}^{n} - U_{jk}^{n-1} + U_{j-1k}^{n} \right) + \frac{1}{\Delta y^{2}} \left(U_{jk+1}^{n} - U_{jk}^{n-1} + U_{jk-1}^{n} \right), \\ \end{pmatrix}$$

