



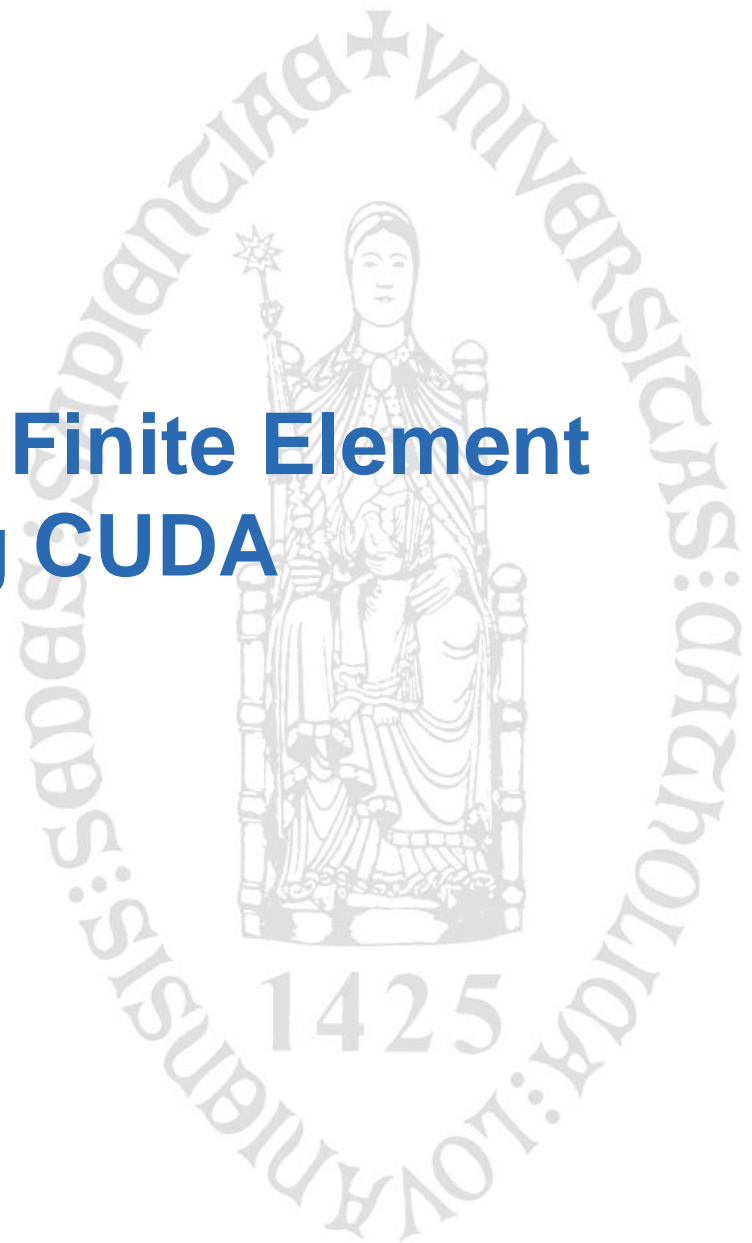
KATHOLIEKE UNIVERSITEIT  
**LEUVEN**

# Nonlinear Real-time Finite Element Analysis using CUDA

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Biomechanics section

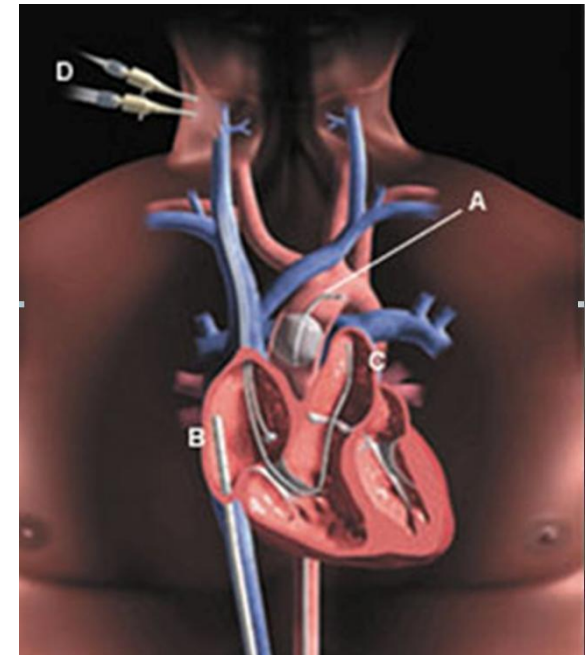
KU Leuven



- 
- Introduction
    - Biomechanics & surgery application
    - What are we trying to do?
    - Why is FE relevant to us?
  
  - FE and parallelization, TLED
    - Which formulation is better and why?
    - Basic Nonlinear Implicit
    - Total Lagrangian Explicit Dynamic
  
  - Results, use case
  
  
  - Conclusion

## Intro

- **Biomechanics**
  - Help DURING surgery, need fast computation of stresses/damage
    - Can't influence surgery workflow
    - tool: FE, generally very slow– we use CUDA
  
- **Finite Element Analysis**
  - Speed and accuracy is what we are going for.
  - Not “embarrassingly parallel”



Heartport Endoaortic clamp,  
Redwood, USA

Time integration:

Implicit  
 $Ku = R$

Explicit

$$M\ddot{u} + c_d M\dot{u} + F(u) = R$$

+ e.g. central difference method

$${}^{t+\Delta t}\mathbf{u}_e = a({}^t\mathbf{R}_e - {}^t\mathbf{F}_e) + b{}^t\mathbf{u}_e - c{}^{t-\Delta t}\mathbf{u}_e$$

$$a = -\frac{2(\Delta t)^2}{2t\Delta t M_e} \quad b = \frac{1+(2-q\Delta t)}{2+q\Delta t} \quad c = \frac{2-q\Delta t}{2+q\Delta t}$$

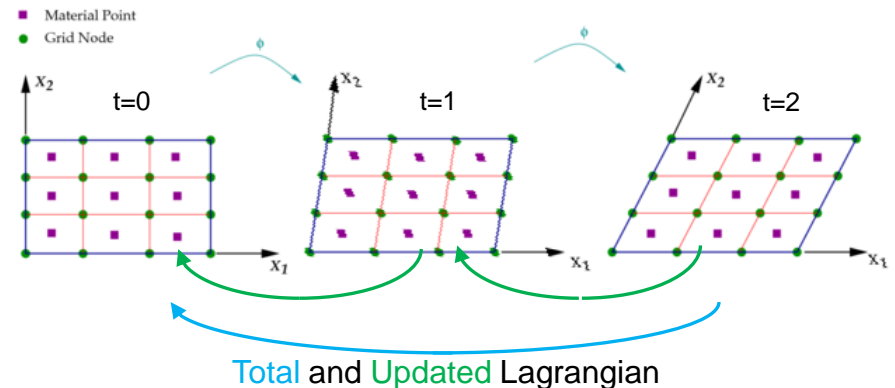
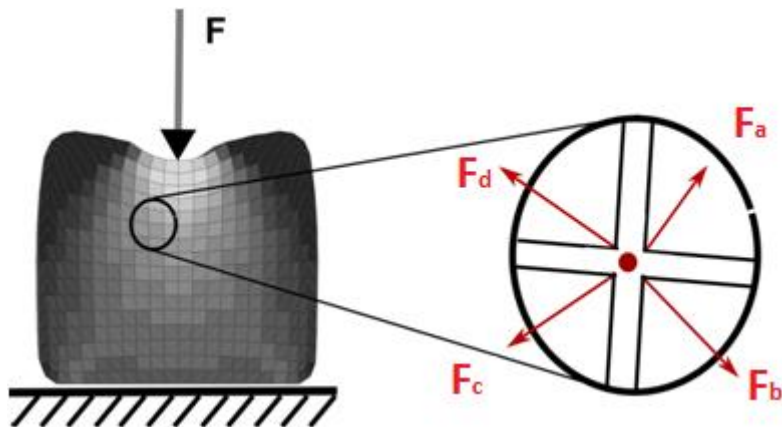
To solve:

Iterative schemes  
(Newton-Raphson)

Convergence:

$$(R - F) < tolerance$$

Energy has stabilized



# Basic nonlinear implicit FE, example on elasticity

Implicit, nonlinear:

```

set  $\nabla \mathbf{N}_{iso}$   $\longrightarrow$  precomputed
for  $t = 1 \dots timesteps$ 
  for  $i = 1..maxIterations$ 
    for  $e = 1 \dots elements\{$ 
      construct  ${}^e \mathbf{D}$ 
       $\mathbf{J} = \nabla \mathbf{N}_{iso}({}^e \mathbf{d} + {}^e \mathbf{u})$  cheap
       $\nabla \mathbf{N}_{glo} = \mathbf{J}^{-1} * \nabla \mathbf{N}_{iso}$  cheap
      construct  ${}^e \mathbf{B}$ 
       $\mathbf{I} = {}^e \mathbf{B}^T {}^e \mathbf{D} {}^e \mathbf{B} * |J|$  cheap
       ${}^e \mathbf{K} = (\text{Gauss}) * \mathbf{I}$ 
    }
     $\mathbf{K} = \sum {}^e \mathbf{K}$  cheap
    Impose B.C.s cheap
     $\Delta \mathbf{u}_i = \mathbf{K}^{-1}(\mathbf{R} - \mathbf{F}_{i-1})$  expensive
     $\mathbf{u}_i = \mathbf{u}_{i-1} + \Delta \mathbf{u}_i$  cheap
     $\mathbf{F}_i = \mathbf{K}_{i-1} * \mathbf{u}_i$  cheap
    Check  $(\mathbf{R} - \mathbf{F}_i)$  and continue/exit
  }
}
    
```

} Newton-Raphson iteration

- Pros:
  - less timesteps
  - unconditionally stable
  - insensitive to incompressibility

- Cons:
  - Large matrix inversion
  - More memory intensive
  - More computationally intensive
  - Little is precomputed

\*D=stress/strain matrix

\*B=strain/displacement matrix

# Total Lagrangian Explicit Dynamic

```

set  $\nabla \mathbf{N}_{iso}$ 
Loop over elements {
     ${}^e J = \nabla \mathbf{N}_{iso} {}^e \mathbf{u}$ 
     $\nabla \mathbf{N}_{glo} = J^{-1} \nabla \mathbf{N}_{iso}$ 
    construct  ${}^e \mathbf{B}$ 
}
construct (diagonalized) mass matrix  $M$ 
Loop over timesteps  $t$  {
    Loop over elements  $e$  {
         ${}^t \mathbf{X}_e = \mathbf{I} + \nabla \mathbf{N}_e {}^t \mathbf{u}_e$ 
         ${}^t \mathbf{C}_e = {}^t \mathbf{X}_e^T {}^t \mathbf{X}_e$ 
         $J = \det(\mathbf{X}_e)$ 
         ${}^t \mathbf{S}_e = (1 - d) \left[ \mu J^{-1} \left( \mathbf{I} - \frac{\text{tr}({}^t \mathbf{C}_e)}{2} {}^t \mathbf{C}_e^{-1} \right) + \kappa J (J - 1) {}^t \mathbf{C}_e^{-1} \right]$ 
         ${}^t \mathbf{F}_e = \nabla \mathbf{N}_e * {}^t \mathbf{S}_e * {}^t \mathbf{X}_e * \text{weight}(e)$ 
    end element loop
     ${}^t \mathbf{F} = \sum {}^t \mathbf{F}_e$ 
     ${}^{t+\Delta t} \mathbf{u}_e = a({}^t \mathbf{R}_e - {}^t \mathbf{F}_e) + b {}^t \mathbf{u}_e - c {}^{t-\Delta t} \mathbf{u}_e$ 
     ${}^{t+\Delta t} \mathbf{u}_e = \mathbf{d}(t + \Delta t)$   ${}^{t+\Delta t} \mathbf{F}_e = \mathbf{F}(t + \Delta t)$ 
}
    
```

precompute();

$$a = -\frac{2(\Delta t^2)}{2t\Delta t M_e}$$

$$b = \frac{1+(2-q\Delta t)}{2+q\Delta t}$$

$$c = \frac{2-q\Delta t}{2+q\Delta t}$$

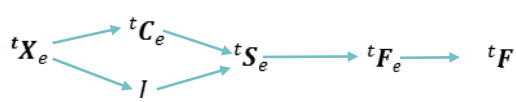
computeForces();

updateDisplacements();

doLoading();

- Pros:
  - lots of precomputation
  - no K matrix assembly
  - no K inversion
  - elements “independent”
  - much less memory required
- Cons:
  - many more timesteps
  - addit. time to converge
  - conditionally stable
  - sensitive to incompressibility

\*neo-hookean material  
 \*d is damage



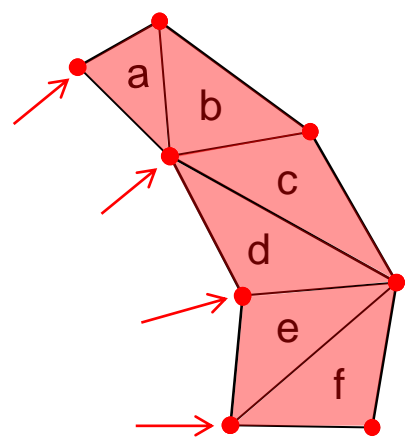
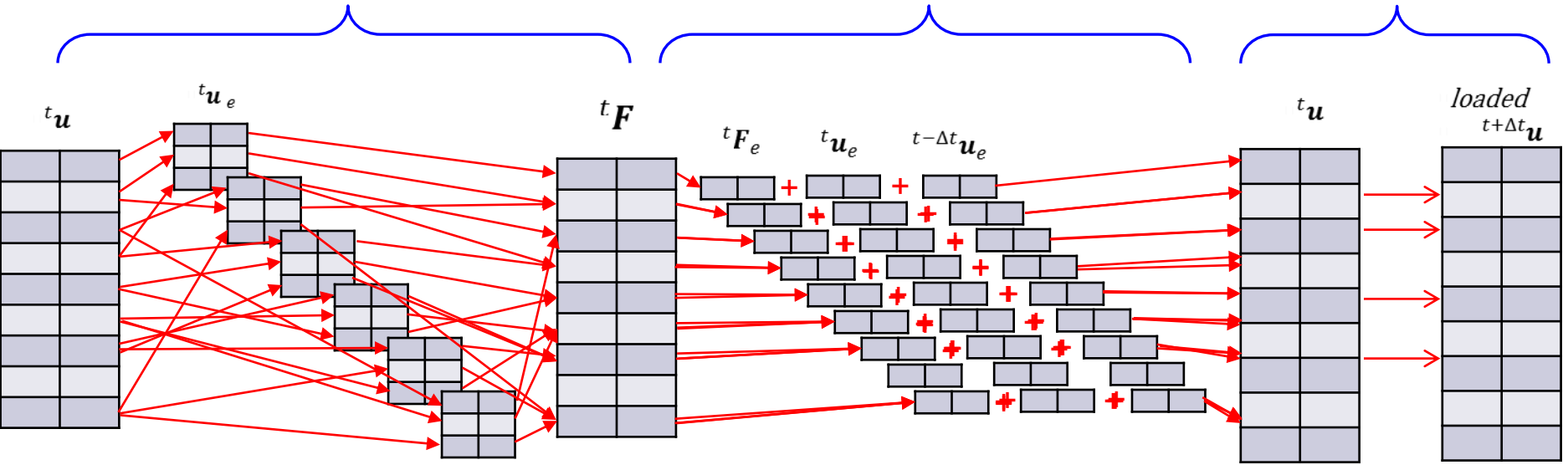
Get forces()

$$t+\Delta t \mathbf{u}_e = a(t\mathbf{R}_e - t\mathbf{F}_e) + b t\mathbf{u}_e - c t-\Delta t \mathbf{u}_e$$


Update Displacements()

$$t+\Delta t \mathbf{u}_e = d(t + \Delta t)$$

doDisplacementLoading()



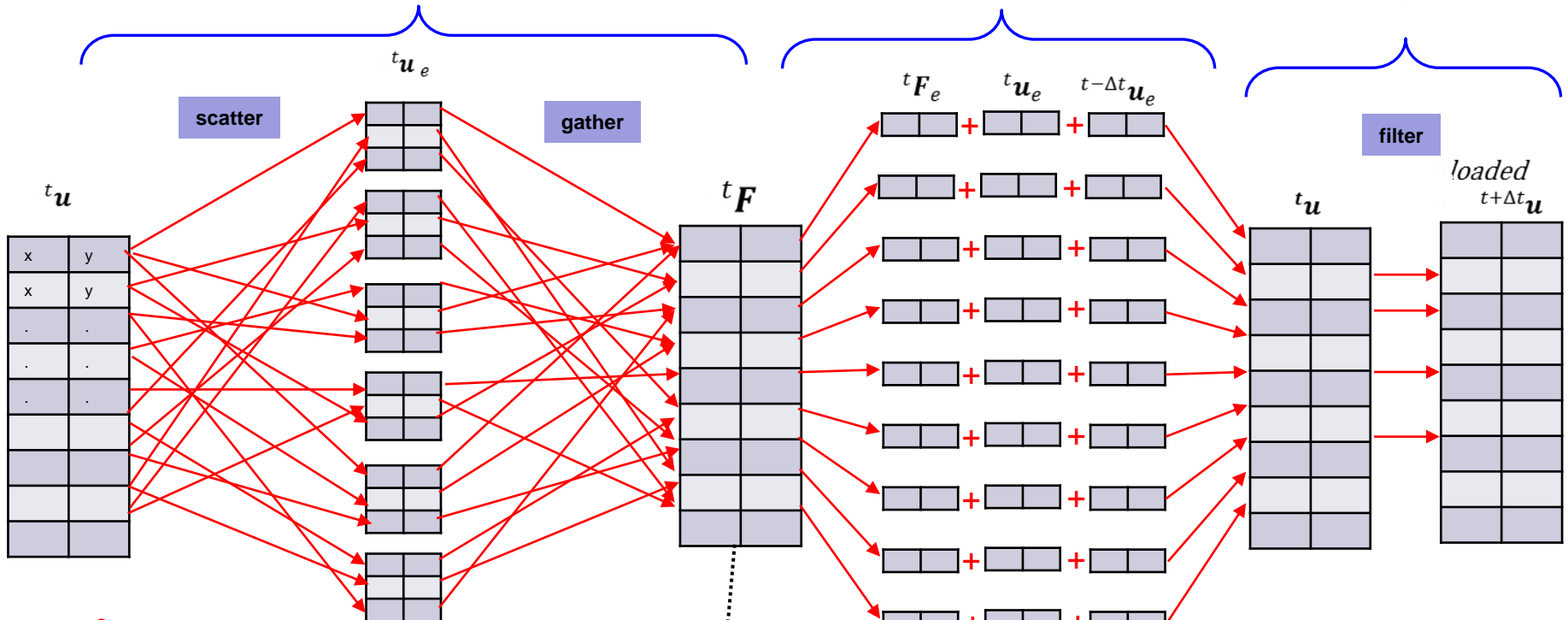
- ①  $t\mathbf{X}_e = \mathbf{I} + \nabla \mathbf{N}_e \mathbf{u}_e$
- ②  $t\mathbf{C}_e = t\mathbf{X}_e^T t\mathbf{X}_e$
- ③  $J = \det(\mathbf{X}_e)$
- ④  $d = \gamma(1 - e^{-\frac{\beta}{m}})$
- ⑤  $t\mathbf{S}_e = (1 - d) \left[ \mu J^{-1} \left( \mathbf{I} - \frac{\text{tr}(t\mathbf{C}_e)}{2} t\mathbf{C}_e^{-1} \right) + \kappa J (J - 1) t\mathbf{C}_e^{-1} \right]$
- ⑥  $t\mathbf{F}_e = \nabla \mathbf{N}_e * t\mathbf{S}_e * t\mathbf{X}_e * \text{weight}(e)$
- ⑦  $t\mathbf{F} = \sum t\mathbf{F}_e$

 = precomputed

Get forces(...)

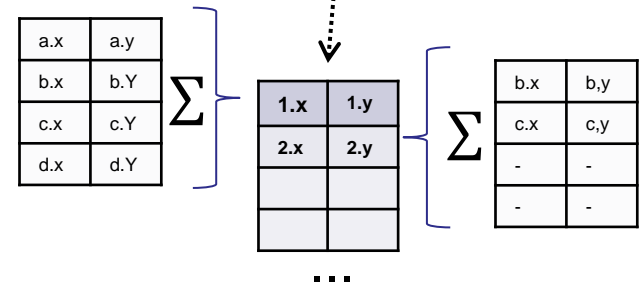
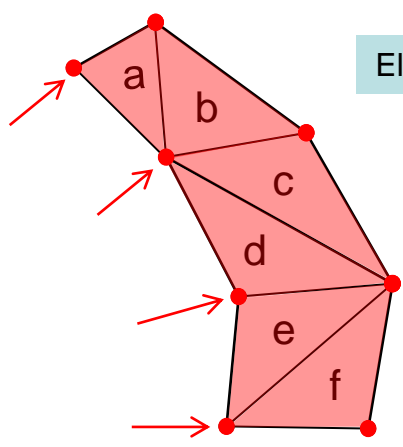
Update Displacements(...)

doLoading(...)



Element granularity

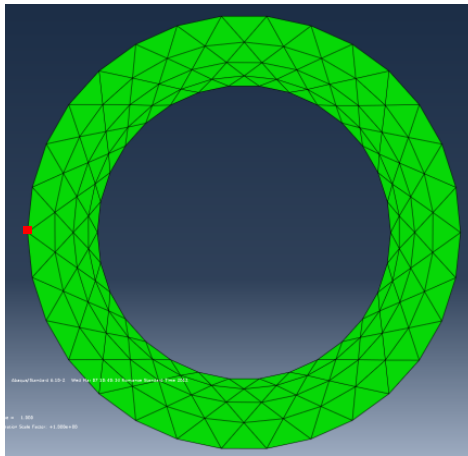
Nodal granularity



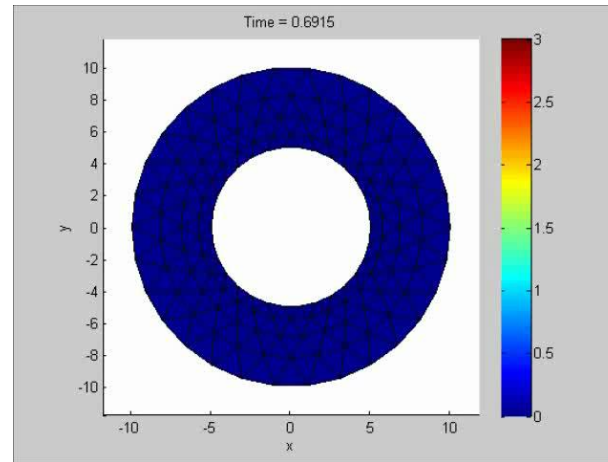
- New: scatter, gather & filter.
- One more summation step.



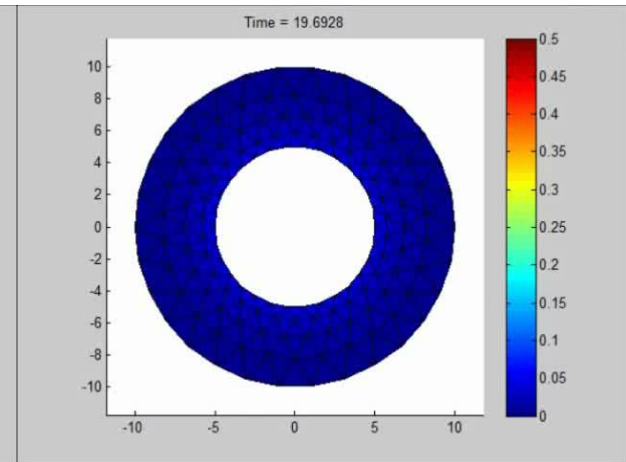




Abaqus/Standard solution



Cuda/TLED: displacement (mm)



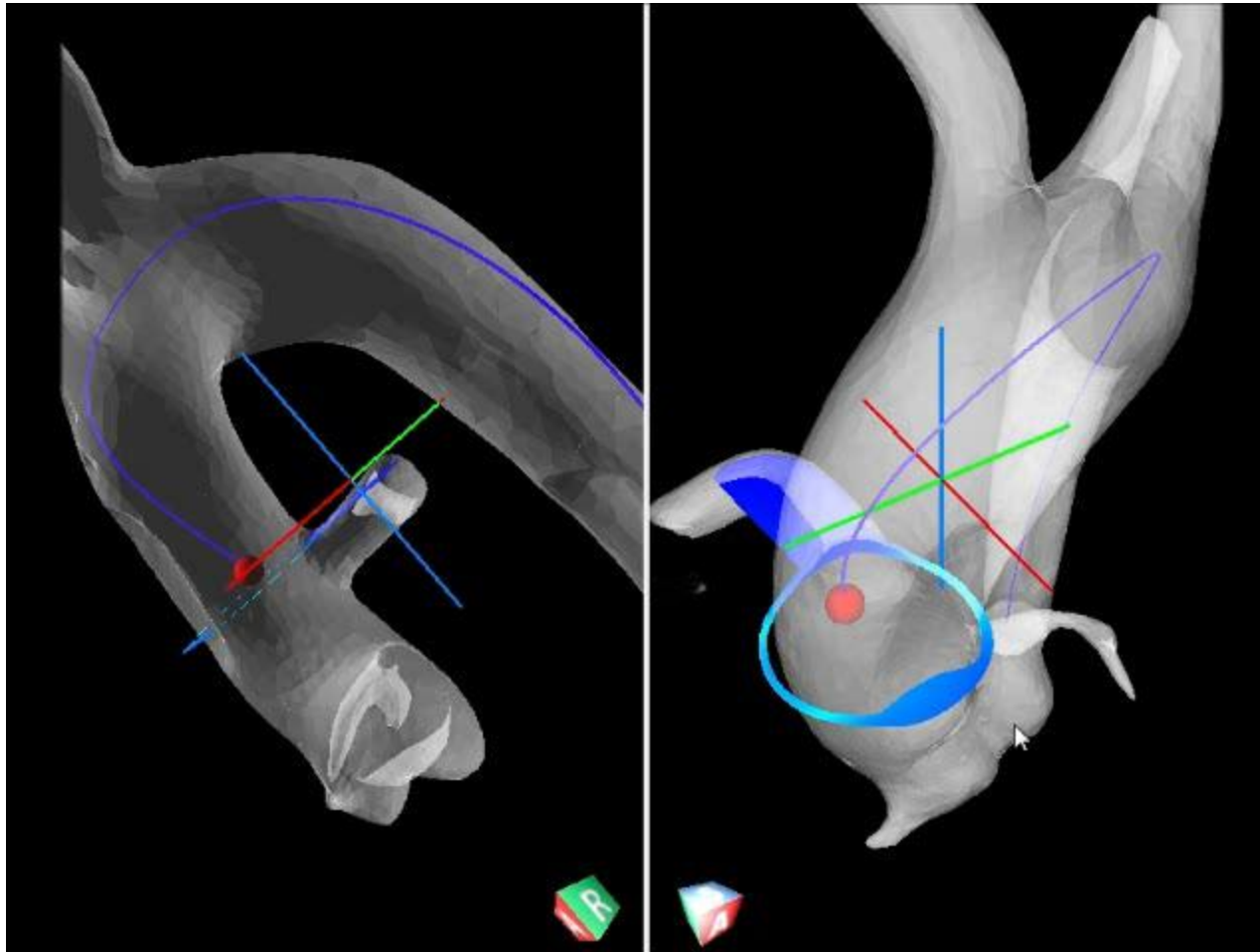
damage(-)

Parameter	Value
$\mu$	1006.7 MPa
$\kappa$	50e3 MPa
$\gamma$	0.9 [-]
$m$	1000 MPa

Method	Material Model	Displacement of node x[mm]	% deviation	Calculation time
Abaqus explicit	Neo Hookean Damage	1.8135 1.8119		4min 4.6min
Abaqus implicit	Neo Hookean Damage	1.8125 1.8109		400ms 500ms
CUDA TLED	Neo Hookean Damage	1.8163 1.8147	0.21% 0.21%	81ms 87ms

\*xeon e6520 vs Tesla C2075(in single precision)

## Code in action: human aorta model + slices



- Mimicking endoclamp balloon expansion.
- Loading is 0-500mmHg sine wave.
- Simulation at 14 fps  
~70ms per solution, 1000 timesteps each.
- Material is Neo-Hookean with  $E=3000\text{Pa}$ ,  $\nu=0,49$ .
- Small enough to fit all element/node data into registers.

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**Thanks!**

## TLED – Precomputation phase

- Initialize and compute
  - Mass matrix
  - Shape function derivatives
  - **CDM coefficients**
  
- Explicit integration scheme
  - Central Difference Method
  - No need for optimization/iterative methods
  - Covers nonlinear behavior in an intuitive way – given small enough timestep is used.

$${}^t\dot{\mathbf{u}} = \frac{1}{2\Delta t} ({}^{t+\Delta t}\mathbf{u} - {}^{t-\Delta t}\mathbf{u})$$

$${}^t\ddot{\mathbf{u}} = \frac{1}{2\Delta t^2} ({}^{t-\Delta t}\mathbf{u} - 2{}^t\mathbf{u} + {}^{t+\Delta t}\mathbf{u})$$

$$\mathbf{M}{}^t\ddot{\mathbf{u}} + \mathbf{K}({}^t\mathbf{u}){}^t\mathbf{u} = {}^t\mathbf{R}$$

$${}^t\mathbf{F}$$

$${}^{t+\Delta t}\mathbf{u} = \frac{\Delta t^2}{\mathbf{M}} ({}^t\mathbf{R} - {}^t\mathbf{F}) + 2{}^t\mathbf{u} - {}^{t-\Delta t}\mathbf{u}$$

diagonalized

$$a = -\frac{2(\Delta t^2)}{2t\Delta t M_e}$$

$${}^{t+\Delta t}\mathbf{u}_e = \frac{\Delta t^2}{m_e} ({}^t\mathbf{R}_e - {}^t\mathbf{F}_e) + 2{}^t\mathbf{u}_e - {}^{t-\Delta t}\mathbf{u}_e$$

$$b = \frac{1 + (2 - q\Delta t)}{2 + q\Delta t}$$

$$c = -\frac{2 - q\Delta t}{2 + q\Delta t}$$

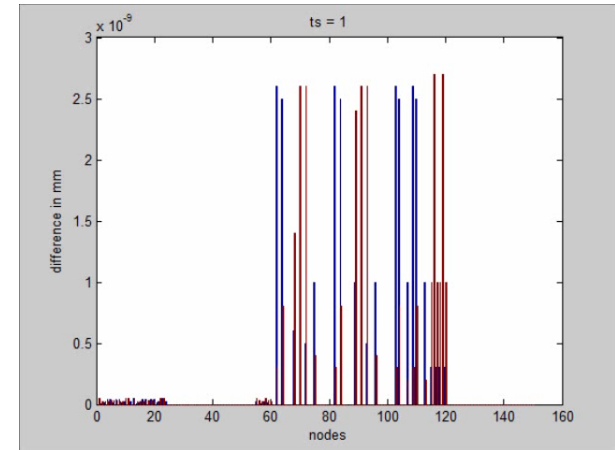
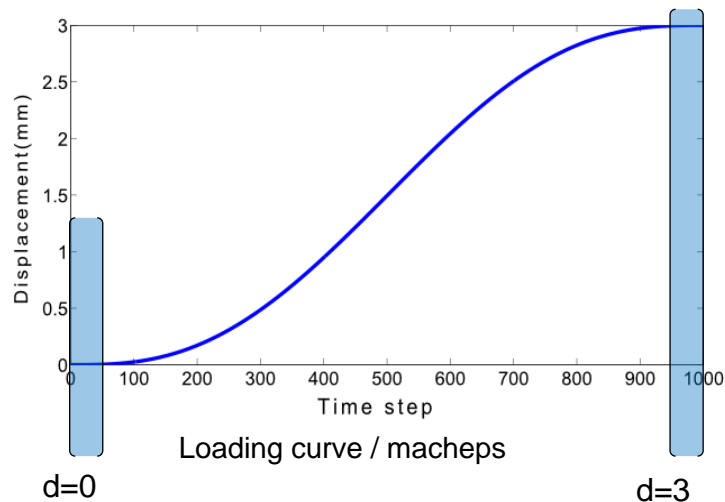
$$q = \frac{2(1 - C_r^2)}{\Delta t(1 + C_r)}$$

- Add mass proportional damping according to The Dynamic Relaxation algorithm

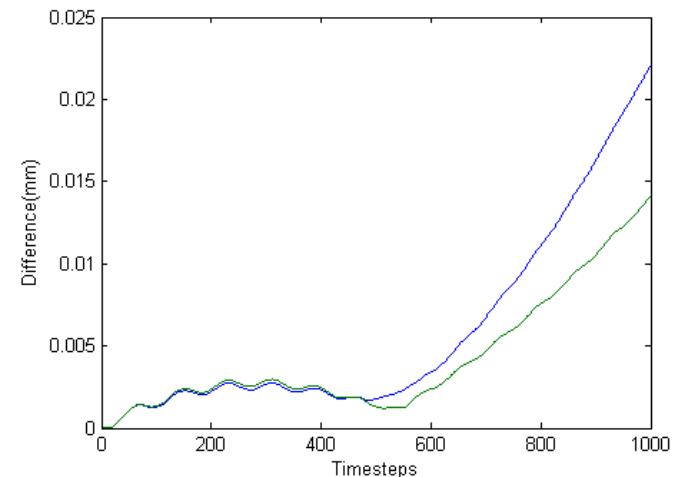
\*CDM = central difference method

## Accuracy considerations

- Single precision
- Some increments  $<$  macheps
- **Total error after 1000ts  $<$  1 mm = single precision warranted**

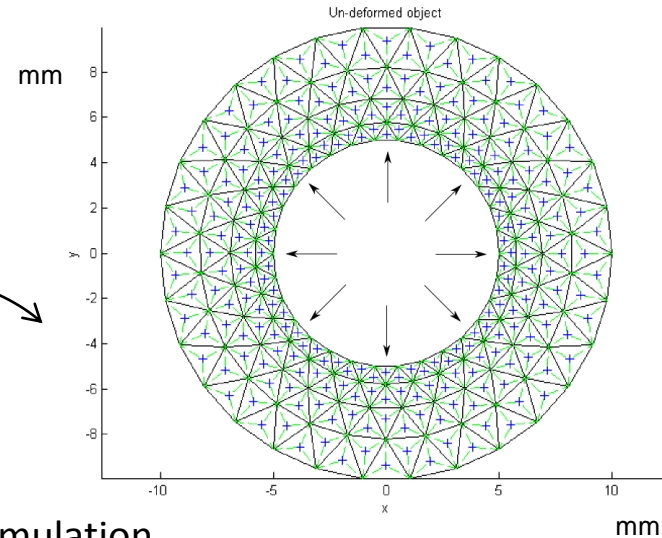
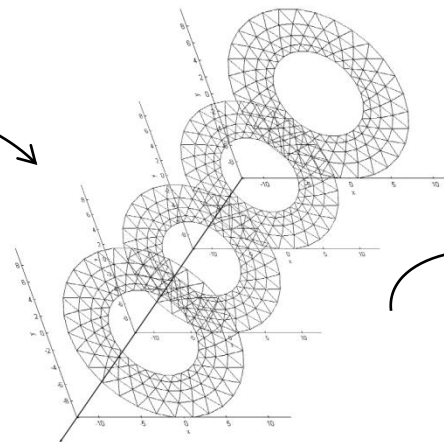


Accumulation of error (CUDA vs Matlab)



Evolution of error (single vs double)

# Endoclamp case study



- Hyperelastic constitutive model
  - Neo-hookean matrix material of artery
  - + damage
- Plane strain formulation
- 3-noded linear triangular element
- single point Gaussian integration

$$\mathbf{S} = 2 \frac{\partial \Psi}{\partial \mathbf{C}}$$

$$\Psi = \frac{\mu}{2} (\mathbf{I}_1 - 3) + \frac{\kappa}{2} (J - 1)^2$$

$$\mathbf{S} = \mu J^{-1} \left( \mathbf{I} - \frac{\text{tr}(\mathbf{C})}{2} \mathbf{C}^{-1} \right) + \kappa J (J - 1) \mathbf{C}^{-1} \longrightarrow$$

$$d = \gamma (1 - e^{-\frac{\beta}{m}})$$

$$\Psi^{dev} = (1 - d) \Psi^{dev}$$

$$\mathbf{S}^{dam} = (1 - d) \mathbf{S}$$

$$\beta = \max_{0 < t < T} (\Psi(t) - \Psi_0)$$

$$\gamma \in ]0, 1]$$