

# High-Order Adaptive Time Stepping of Vesicle Suspensions

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# Outline

- Vesicle suspensions
- Numerical methods and algorithmic challenges
- High-order adaptive time stepping
- Low resolution simulations
- Concluding remarks

# What is a vesicle?

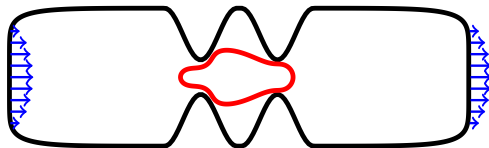
- Deformable (2D) capsule filled and submerged in viscous fluid
- Boundary is locally inextensible
- Dynamics governed by hydrodynamic interactions with other vesicles/walls, resistance to bending and inextensibility
- Basic model for biomembranes such as red blood cells and platelets

# Governing equations

## Spectral representation

$$\mathbf{x}(\theta, t) = \sum_n \hat{\mathbf{x}}_n(t) e^{in\theta}$$

$$\sigma(\theta, t) = \sum_n \hat{\sigma}_n(t) e^{in\theta}$$



## Fourier differentiation

$$\sigma_s = \frac{\sum_n in \hat{\sigma}_n(t) e^{in\theta}}{\left| \sum_n in \hat{\mathbf{x}}_n(t) e^{in\theta} \right|}$$

- Similar non-linear expression for  $\mathbf{x}_{SSSS}$

## Vesicle equations

$$\frac{d\mathbf{x}}{dt} = \mathbf{v}_\infty + \mathcal{S}[\boldsymbol{\xi}] \quad \mathbf{x} \in \gamma$$

$$0 = \text{div}_\gamma(\mathbf{v}_\infty + \mathcal{S}[\boldsymbol{\xi}]) \quad \mathbf{x} \in \gamma$$

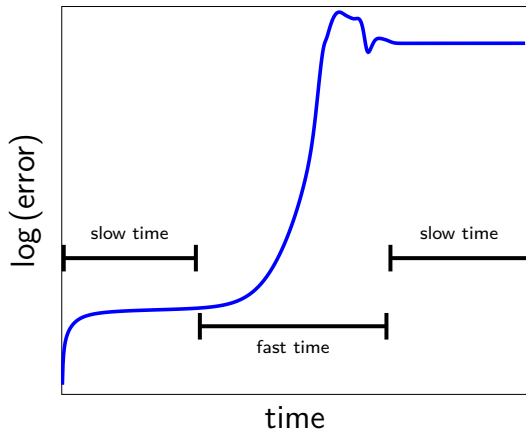
$$\boldsymbol{\xi} = \boldsymbol{\xi}(\mathbf{x}, \sigma) = -\mathbf{x}_{SSSS} + (\sigma \mathbf{x}_s)_s$$

## Stokes single-layer potential

$$\mathcal{S}[\boldsymbol{\xi}] = \frac{1}{4\pi} \int_\gamma \left( -\log \rho + \frac{\mathbf{r} \otimes \mathbf{r}}{\rho^2} \right) \boldsymbol{\xi}(\mathbf{y}) ds_{\mathbf{y}}$$

# Algorithmic challenges

- Moving boundaries
- High-order derivatives
- Stiffness
- Nearly touching interfaces
- Fast summations
- **High-order time adaptivity**
- Stability at low resolutions



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- Moving boundaries
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# Semi-implicit time stepping

Quaife and Biros, J. Comput. Phys. 2014

Semi-implicit discretization is

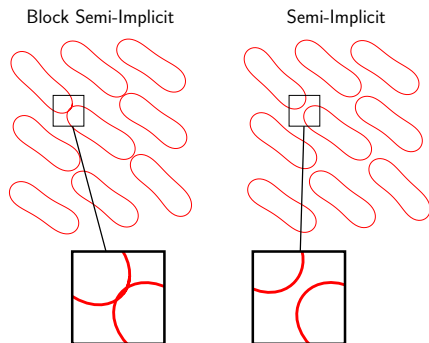
$$\frac{\mathbf{x}_j^{N+1} - \mathbf{x}_j^N}{\Delta t} = \mathbf{v}(\mathbf{x}_j^N; \mathbf{x}_j^{N+1}) + \sum_{k \neq j} \mathbf{v}(\mathbf{x}_j^N; \mathbf{x}_k^{N+1})$$

- Resolves self-interaction stiffness

$$\mathbf{x}_{SSSS} \quad \Delta t \sim N^{-3}$$

$$(\sigma \mathbf{x}_s)_s \quad \Delta t \sim N^{-1}$$

- Resolves stiffness from nearly touching vesicles



# Couette

- $N = 96$
- $N_{\text{wall}} = 256$
- $e_A = 7.3\text{E}-3$
- $e_L = 1.6\text{E}-2$
- 40.5 % volume fraction



# Spectral deferred correction for vesicle suspensions

- Vesicle suspension governed by  $\dot{\mathbf{x}}(t) = \mathcal{M}[\mathbf{x}](\mathbf{x})$
- Time stepping matrix is  $\mathcal{M} = (\alpha I - \mathcal{D})^{-1} \mathcal{S} \mathcal{B}$
- Form provisional solution  $\tilde{\mathbf{x}}$  of

$$\mathbf{x}(t) = \mathbf{x}_0 + \int_0^t \mathcal{M}[\mathbf{x}](\mathbf{x}) d\tau$$

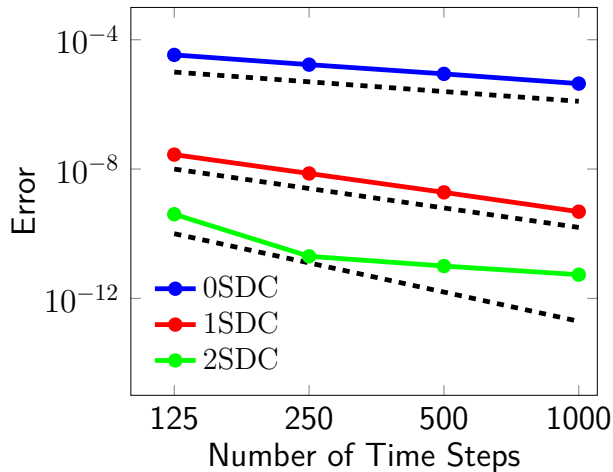
- Compute the residual

$$\mathbf{r}(t) = \mathbf{x}_0 - \tilde{\mathbf{x}}(t) + \int_0^t \mathcal{M}[\tilde{\mathbf{x}}](\tilde{\mathbf{x}}) d\tau$$

- Define the new provisional solution  $\tilde{\mathbf{x}} + \tilde{\boldsymbol{\delta}}$  where  $\tilde{\boldsymbol{\delta}}$  approximates

$$\boldsymbol{\delta}(t) = \mathbf{r}(t) + \int_0^t (\mathcal{M}[\tilde{\mathbf{x}} + \boldsymbol{\delta}] - \mathcal{M}[\tilde{\mathbf{x}}])(\tilde{\mathbf{x}}) d\tau + \int_0^t \mathcal{M}[\tilde{\mathbf{x}} + \boldsymbol{\delta}](\boldsymbol{\delta}) d\tau$$

# SDC convergence



- Gauss-Lobatto substeps used to compute residual
- Spatial errors, ill-conditioning, and GMRES tolerance restrict high-order convergence
- Order reduction

# Adaptive time stepping

Quaife and Biros, J. Comput. Phys. 2016

- Automatically adjust  $\Delta t$
- Crucial for problems with multiple time scales and long time horizons
- Error in area and length estimate the local truncation error

$m$	$e_P$	$e_A$	$e_L$
125	1.1E-4	3.4E-6	3.4E-5
250	5.3E-5	1.7E-6	1.7E-5
500	2.7E-5	8.8E-7	8.8E-6
1000	1.3E-5	4.4E-7	4.4E-6
125	1.4E-7	2.8E-8	1.3E-8
250	3.4E-8	7.3E-9	2.6E-9
500	8.3E-9	1.9E-9	4.5E-10
1000	2.0E-9	4.8E-10	6.9E-11

# Time step size selection

- Solution accepted if  $|A(t + \Delta t) - A(t)| \leq \epsilon A(t) \Delta t$
- Optimal time step size

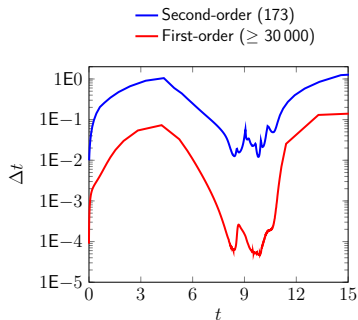
$$\Delta t_{\text{opt}} = \left( \frac{\epsilon A(t) \Delta t}{|A(t + \Delta t) - A(t)|} \right)^{1/k} \Delta t$$

- New time step size

$$\Delta t_{\text{new}} = \alpha^{1/k} \min(\beta_{\text{up}} \Delta t, \max(\Delta t_{\text{opt}}, \beta_{\text{down}} \Delta t))$$

- $\alpha < 1$ ,  $\beta_{\text{up}} > 1$ ,  $\beta_{\text{down}} < 1$  are safety parameters

# First- vs. second-order adaptive time stepping



- Much larger allowable time steps by using one SDC iteration
- 23X speedup for three digits of accuracy

# Controlling the error

## First-order: 0 SDC

Tolerance	Error	Accepts	Rejects	$\mathcal{S}$ evals	CPU
$1\text{E}-2$	$7.9\text{E}-3$	3397	21	$1.1\text{E}5$	5.7
$1\text{E}-3$	$9.4\text{E}-4$	33554	12	$1.1\text{E}6$	59

## Second-order: 1 SDC

Tolerance	Error	Accepts	Rejects	$\mathcal{S}$ evals	CPU
$1\text{E}-2$	$2.6\text{E}-3$	68	30	$2.1\text{E}4$	1.2
$1\text{E}-3$	$5.6\text{E}-4$	172	68	$4.8\text{E}4$	2.6

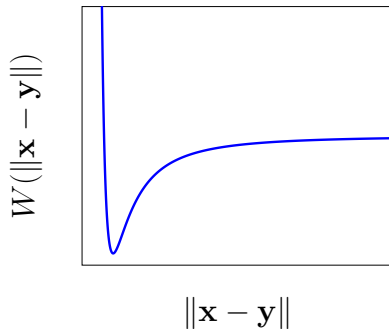
# Tank treading and tumbling

Quaife and Biros, Procedia IUTAM 2014

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- 
- Tank treading for  $\nu < \nu_c$  and tumbling for  $\nu > \nu_c$ ,  $\nu_c \approx 4.05$
  - 1 SDC correction
  - Tolerance of  $10^{-2}$

# Adding adhesion

$$\mathcal{W}(\mathbf{x}) := \sum_{k \neq j} \int_{\gamma_k} W(\|\mathbf{x} - \mathbf{y}\|) ds$$





# Adaptive time stepping at low resolutions

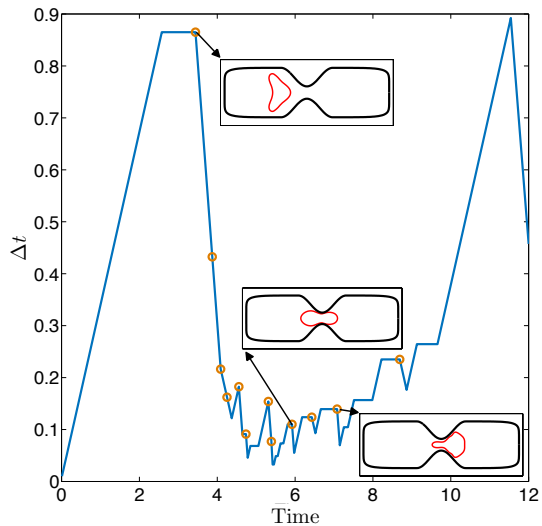
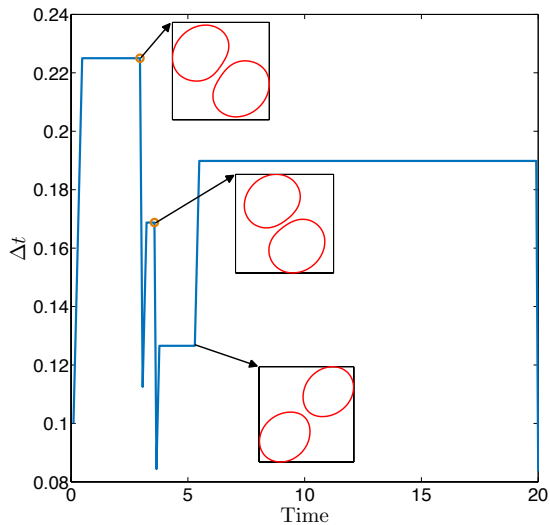
- Asymptotic estimates are invalid at low resolutions and large time step sizes
- If asymptotic estimates are invalid, change time step size by a constant factor

## Time step selection

```
An+1 ← getArea(xn+1) compute area of new shape  
qA ← computeTimeDerivatives(xn+1) compute  $q_A = \left| \frac{dA}{dt} / A \right|$  analytically  
εA ← computeError(An+1, An) compute error in area  
if  $\rho_{\min} \leq \epsilon_A \leq \rho_{AL}$  then error is marginally below tolerance  
    accept = true  
     $\Delta t_{\text{new}} = \Delta t_n$  keep same time step size  
else if  $\epsilon_A < \rho_{\min}$  then error is substantially below tolerance  
    accept = true  
    if  $|\epsilon_A - q_A \Delta t| \leq \rho_{\text{up}} \epsilon_A$  then asymptotic assumption is valid  
         $\Delta t_{\text{new}} = \rho_{AL} / q_A$   
    else asymptotic assumption is invalid  
         $\Delta t_{\text{new}} = \beta_{\text{up}} \Delta t_n$   
    end if  
else if  $\epsilon_A > \rho_{AL}$  then error is greater than tolerance  
    accept = false  
    if  $|\epsilon_A - q_A \Delta t_n| \leq \rho_{\text{down}} \epsilon_A$  then asymptotic assumption is valid  
         $\Delta t_{\text{new}} = \rho_{AL} / q_A$   
    else asymptotic assumption is not valid  
         $\Delta t_{\text{new}} = \beta_{\text{down}} \Delta t_n$   
    end if  
end if  
return  $\Delta t_{\text{new}}$ , accept
```

# Adaptive time stepping at low resolutions

Kabacaoglu, Quaife, and Biros, *J. Comput. Phys.* 2018



# Summary

## Conclusions

- High-order adaptivity achieved with IMEX Euler and SDC
- Used error in area and length to estimate local truncation error
- No trial and error procedure to find correct time step size

## Future Work

- Better understanding of order reduction for SDC
- Multirate time integrator
- Parallel implementation