Computation of matrix functions with fully automatic Schur-Parlett and Rational Krylov methods.

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#### About me

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Master's degree thesis under supervision of professor Federico Poloni.

### MatFun

A Julia package for computing dense and sparse matrix functions automatically (no user input other than f and A).

https://github.com/robzan8/MatFun.jl

#### Matrix Functions

sin(A), exp(A), f(A) ... where A is a square matrix.

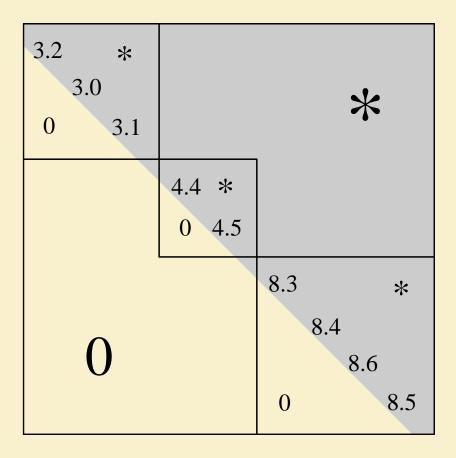
Key factor: values of f and its derivatives on A's spectrum.

Some specialized methods (*expm*, *logm*, *sqrtm*), we are looking for generic ones.

### Schur-Parlett – dense matrix functions

$$A = QTQ^* \Rightarrow f(A) = Qf(T)Q^*$$
  
Need to compute  $F = f(T)$ :

- Group eigenvalues in blocks by proximity;
- > Compute  $f(T_{ii})$  of diagonal blocks with Taylor;
- > Use the Parlett recurrence:  $T_{ii}F_{ij} - F_{ij}T_{jj} = F_{ii}T_{ij} - T_{ij}F_{jj} + \sum_{k=i+1}^{j-1} (F_{ik}T_{kj} - T_{ik}F_{kj})$



### Schur-Parlett - automatic differentiation

Derivatives of *f* are required for Taylor.

Dual numbers:  $x + y\varepsilon$  with  $\varepsilon \neq 0, \varepsilon^2 = 0$ 

$$f(x + y\varepsilon) = f(x) + f'(x)y\varepsilon \Rightarrow$$

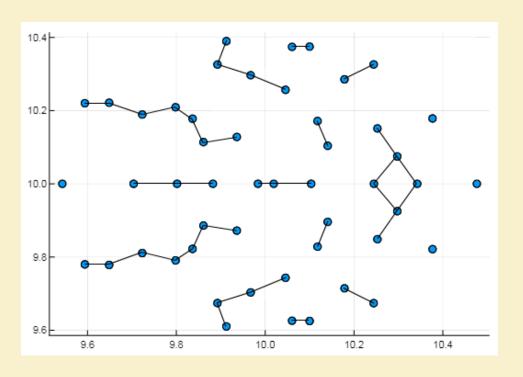
$$f'(x) = Eps(f(x + \varepsilon))$$

In Julia, a function accepting any real number also accepts dual numbers and is therefore automatically differentiable.

We use TaylorSeries.jl for higher-order differentiation.

## Schur-Parlett – numerical accuracy

Boils down to how well the eigenvalues can be clustered:



Relative error versus MatLab's Symbolic Toolbox (on 50x50 random matrix):

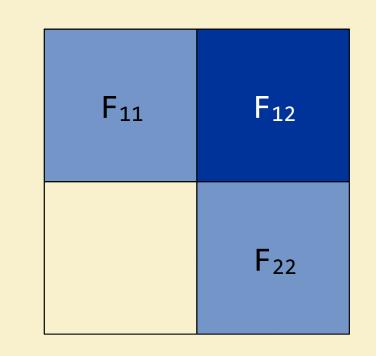
	Specialized method	Schur-Parlett	
exp	1.0e-14	1.5e-13	
log	4.0e-14	4.1e-14	
sqrt	4.3e-14	4.5e-14	

Can behave badly (7e-4) with "snake" eigenvalues (as shown in original paper, experiment 4)

# Schur-Parlett – performance improvements

Parlett recurrence made recursive and cache-oblivious.

F <sub>ii</sub>	$F_{\mathtt{ik}}$	F <sub>ij</sub>
		F <sub>kj</sub>
		F <sub>jj</sub>



$$T_{ii}F_{ij} - F_{ij}T_{jj} = F_{ii}T_{ij} - T_{ij}F_{jj}$$

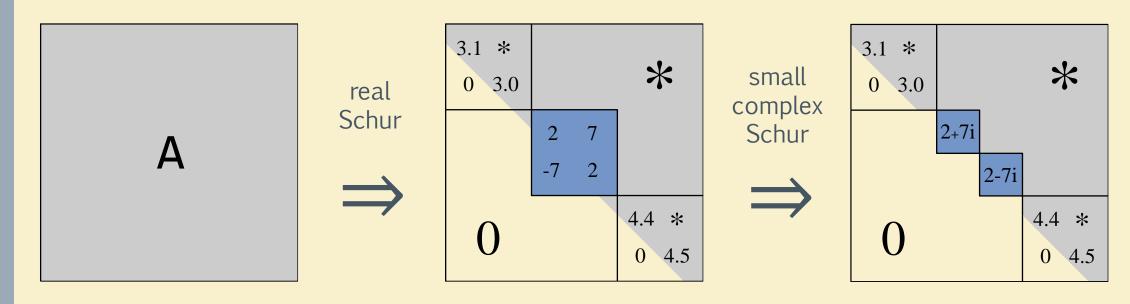
$$+ \sum_{k=i+1}^{j-1} (F_{ik}T_{kj} - T_{ik}F_{kj})$$

$$T_{11}F_{12} - F_{12}T_{22} = F_{11}T_{12} - T_{12}F_{22}$$
  
~3x speedup for n = 2500

Problem: conjugated eigenvalues with big imaginary part must go in different blocks, even with real *A*.

Original solution: do everything in complex arithmetic.

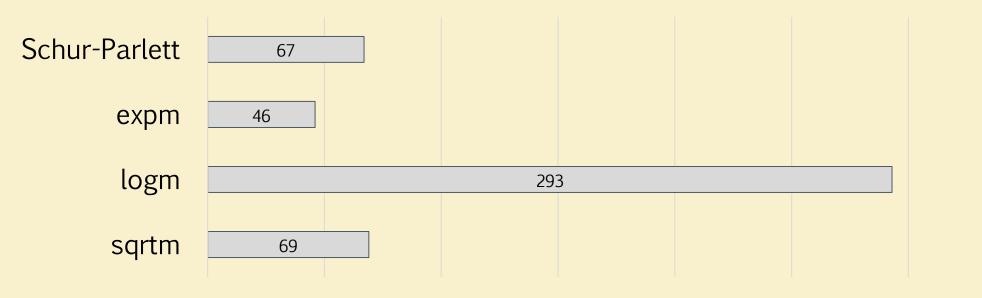
Our solution: complex Schur factorization can be "delayed" and done on small blocks:



Also allows for Parlett recurrence with real Sylvester equations, ~2x speedup for n = 2500 on whole Schur-Parlett.

# Schur-Parlett – performance results

The whole procedure typically spends  $\sim 2/3$  of the time doing A's Schur decomposition (varies depending on the eigenvalues' distribution).



Execution time (s) on randn(2500, 2500)