# A domain decomposition approach to exponential methods for PDEs 

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## Outline of the talk

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- Some preliminary numerical results
- Conclusions and perspectives for atmospheric modelling


## Basic idea of exponential methods

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- Various extensions to nonlinear problems are available


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- Stiff one step, one stage second order solver with one evaluation of RHS: think of the physics...


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- Krylov space dimension (and cost of time step) depend on the Courant number
- Alternative techniques imply similar costs for large scale problems


## Some numerical results

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|  | $h$ error |  |  |  |
| :---: | :---: | :---: | :---: | :---: |
|  | LP | CN | EX2 | EX3 |
| Test 5 | $1.2 \mathrm{e}-2$ | $9.1 \mathrm{e}-3$ | $1.2 \mathrm{e}-3$ | $1.1 \mathrm{e}-3$ |
| Test 6 | $5.9 \mathrm{e}-2$ | $1.7 \mathrm{e}-2$ | $3.8 \mathrm{e}-4$ | $4.0 \mathrm{e}-4$ |

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- Local problems discretized by FD, FV, FE methods yield sparse matrices
- Exponential of a sparse matrix is almost sparse (Iserles 2001)
- For $s$-banded $\mathbf{A}=\left(a_{i, j}\right)$ with $\left|a_{i, j}\right| \leq \rho$, let $\exp (\mathbf{A})=\left(e_{i, j}\right)$.

$$
\begin{aligned}
\left|e_{i, j}\right| & \leq\left(\frac{\rho s}{|i-j|}\right)^{\frac{|i-j|}{s}}\left[e^{\frac{|i-j|}{s}}-\sum_{k=0}^{|i-j|-1} \frac{(|i-j / s|)^{k}}{k!}\right] \\
& \approx\left(\frac{\rho s}{|i-j|}\right)^{\frac{|i-j|}{s}} \frac{(|i-j| / s)^{|i-j|}}{|i-j|!}
\end{aligned}
$$

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- There is no real need to compute a global exponential matrix: Local Exponential Methods (LEM)


## LEM: a domain decomposition approach

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- Decompose mesh in overlapping regions

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\mathcal{M}=\bigcup_{i=1}^{N} \mathcal{M}_{i} \quad \mathcal{M}_{i}=\mathcal{D}_{i} \cup \mathcal{B}_{i}
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- Overwrite degrees of freedom belonging to $\mathcal{B}_{i}$


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- For small enough $\mathcal{D}_{i}$ local matrices can be stored: computational gain if Jacobian is frozen every few time steps and in the limit of large number of advected species


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- Next on the to do list: use Local Exponential Methods in a high order FE framework and with complex forcing terms (multiple ARD with chemistry)

