



Stability and Receptivity of the Swept Wing Attachment-Line Boundary Layer: a Multigrid Numerical Approach

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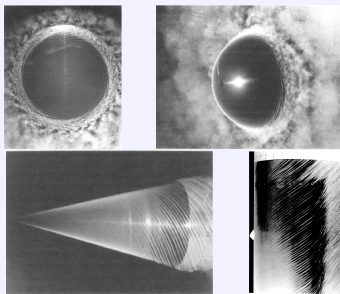
Motivation

- ▶ Gain a better understanding of the coherent flow structures and their dynamics in the leading-edge region of swept wings
- ▶ Identify the regions of maximum receptivity and sensitivity

Industrial research



Academic research

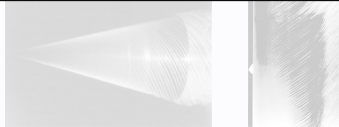


Motivation

- ▶ Gain a better understanding of the coherent flow structures

“The goal of any scientific study of a fluid-dynamical process is not in the reproduction of its physical features by direct numerical simulations but in the extraction of the governing underlying mechanisms from the data the DNS produces. In other words, we are interested in the intrinsic flow behavior captured by the dynamics of coherent structures.”

Mack & Schmid, 2010



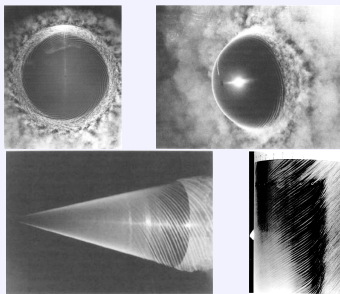
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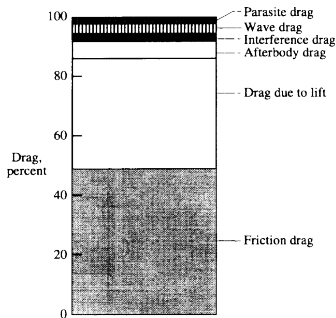
Industrial research



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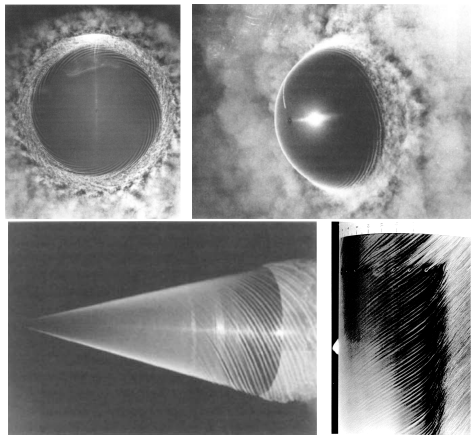
Industrial research



Drag components for a subsonic aircraft; Thibert, Reneaux & Schmitt, 1990

- ▶ Friction drag $\approx 50\%$ of total drag for a subsonic aircraft
- ▶ Laminar friction drag $\approx 1/10$ of turbulent friction drag
- ▶ Large savings are possible by extending the laminar flow region
- ▶ An understanding of the instabilities leading to turbulence is required

Academic research



Reed and Saric, 1989; Poll, 1978

Non-alignment between the velocity vector and the pressure gradient is at the origin of three dimensional boundary layers

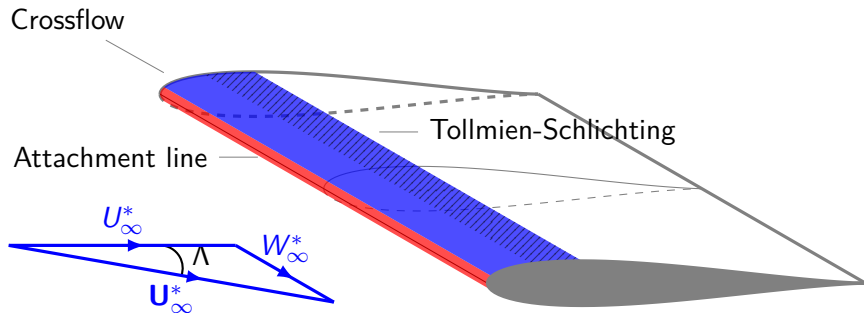
Outline

- ▶ Swept-wing instabilities
- ▶ Procedure and tools
- ▶ Multigrid
- ▶ Base flow
- ▶ Perturbations: a modal description
- ▶ Receptivity and sensitivity
- ▶ Conclusions and future work

Outline

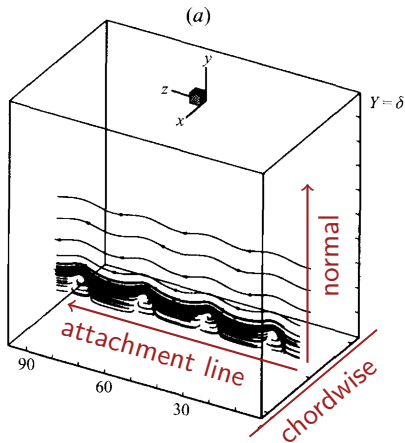
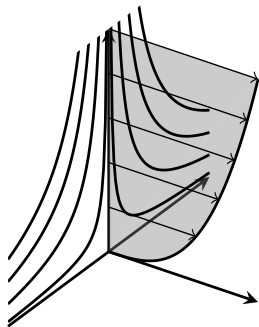
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Swept-wing instabilities



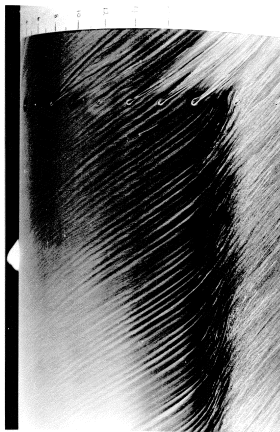
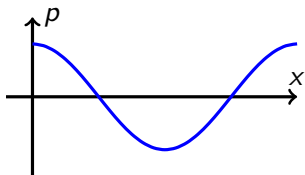
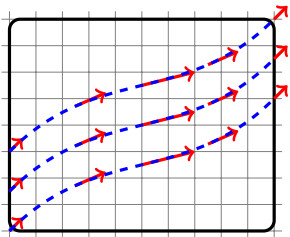
Local analysis based on simplified flow models provides us with different instability mechanisms for different regions

Swept-wing instabilities — Attachment line

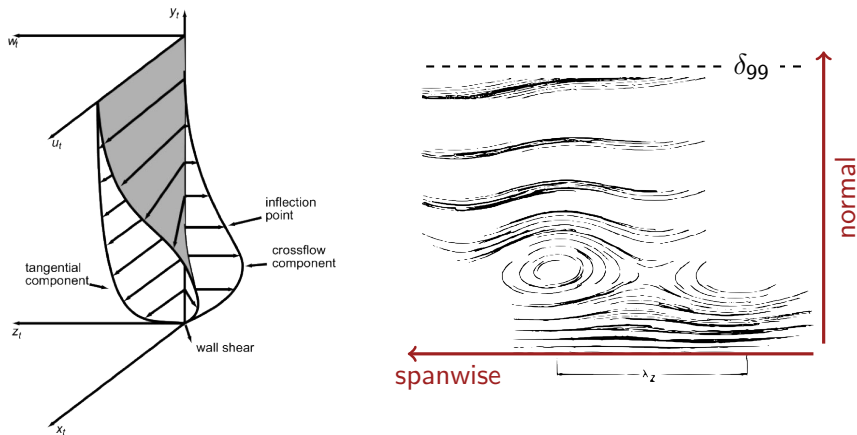


Attachment-line instability: streamlines under the influence of the most unstable modes; Lin & Malik 1996.

Swept-wing instabilities — Crossflow vortices

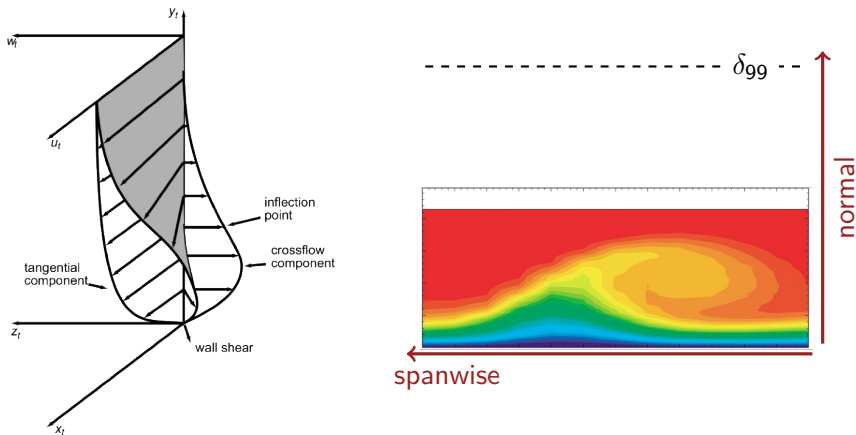


Swept-wing instabilities — Crossflow vortices



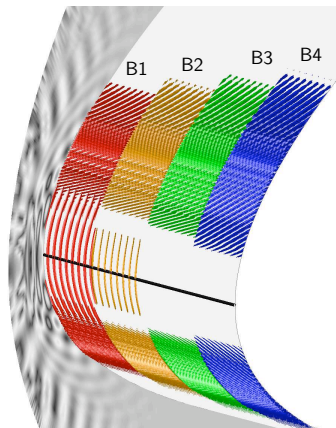
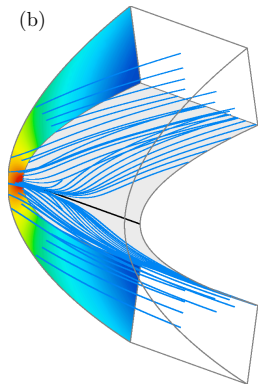
**Crossflow instability: streamlines of crossflow vortices
obtained from experimental results; Reed, 1988**

Swept-wing instabilities — Crossflow vortices



Streamwise velocity contours under the influence of crossflow vortices, from experimental results; Haynes & Reed, 2000

Swept-wing instabilities — Global analysis

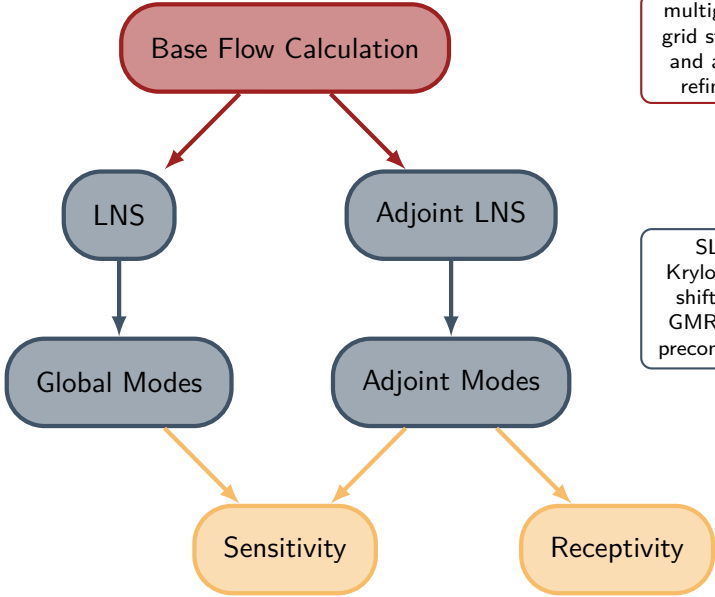


Isosurfaces of the normal velocity component of the disturbance
Hypersonic flow with bow-shock inflow; Mack, 2009

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Procedure and tools



multigrid with
grid stretching
and adaptive
refinement

SLEPc:
Krylov-Schur,
shift-invert,
GMRES, ILU
preconditioning

Procedures and tools — Governing equations

Base Flow

$$\mathcal{R}(\mathbf{q}) \equiv \begin{cases} \nabla \mathbf{u} \mathbf{u} - \nu \Delta \mathbf{u} + \nabla p & = 0 \\ \nabla \cdot \mathbf{u} & = 0 \end{cases}$$

Linearized equations for the perturbations \mathbf{q}'

$$\mathcal{L} \mathbf{q}' = \left(B \partial_t + \frac{\partial \mathcal{R}}{\partial \mathbf{q}} \Big|_{\mathbf{Q}} \right) \mathbf{q}' \quad \Longrightarrow \quad \mathcal{L}(\mathbf{Q}) \mathbf{q}' = \mathbf{f}'$$

Modal decomposition $\hat{\mathbf{q}}'$

$$\hat{\mathbf{q}}' = \int_0^{\infty} \mathbf{q}' e^{-\sigma t} dt \quad \Longrightarrow \quad \hat{\mathcal{L}} \hat{\mathbf{q}}' = (\sigma B + A) \hat{\mathbf{q}}' = \hat{\mathbf{f}}' + \mathbf{q}_0$$

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Procedures and tools — Receptivity and sensitivity

Lagrangian functional

$$\mathcal{I}(\text{obj}, \hat{\mathbf{q}}, \hat{\mathbf{q}}^+, \hat{\mathbf{f}}, A) = \text{obj} - \langle \hat{\mathbf{q}}^+, (\sigma B + A) \hat{\mathbf{q}} - \hat{\mathbf{f}} \rangle$$

$$\langle \mathbf{a}, \mathbf{b} \rangle = \int_{\Omega} \mathbf{a}^H \mathbf{b} d\Omega$$

Receptivity

What is the variation of the objective functional given a variation in the forcing?

$$\frac{\partial \mathcal{I}}{\partial \hat{\mathbf{f}}} \delta \hat{\mathbf{f}} = 0$$

$$\delta(\text{obj}) = -\langle \hat{\mathbf{q}}^+, \delta \hat{\mathbf{f}} \rangle$$

Sensitivity

What is the variation of the objective functional given a variation in the operator A ?

$$\frac{\partial \mathcal{I}}{\partial A} \delta A = 0$$

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Procedures and tools — Receptivity and sensitivity

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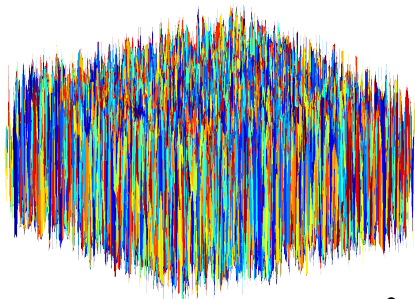
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$m = 0$

Poisson problem: error $e^m = \bar{q} - q^m$ of the approximate solution at iteration m

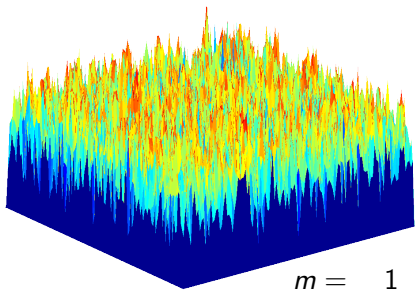
▶ $m = 0$

▶ $m = 10$

▶ $m = 100$

The Gauss-Seidel algorithm

- ▶ Simple and memory efficient algorithm
- ▶ Fast convergence for high wavenumbers
- ▶ Most of the computational cost is related to the reduction of the low wavenumber error components



$m = 1$

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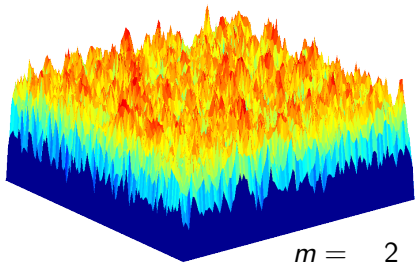
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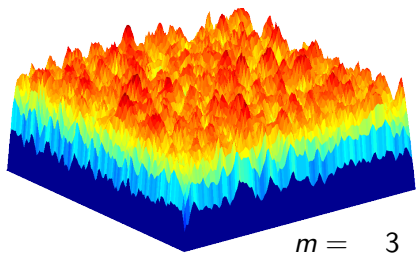
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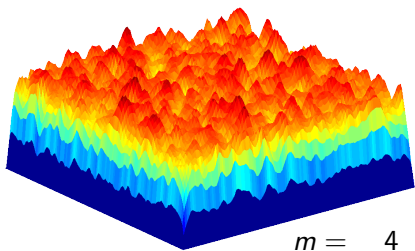
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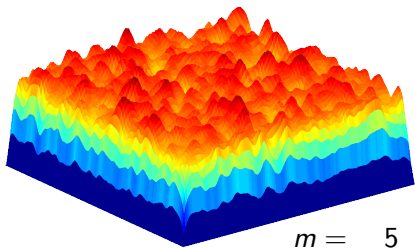
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$m = 5$

Poisson problem: error $e^m = \bar{q} - q^m$ of the approximate solution at iteration m

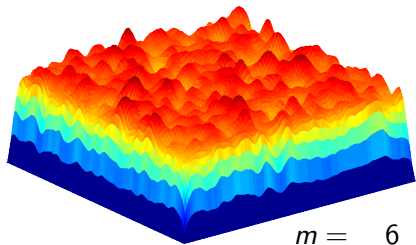
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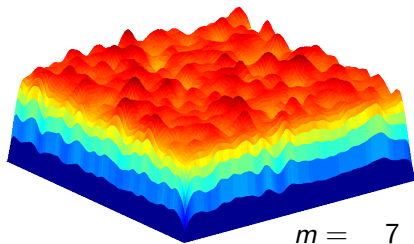
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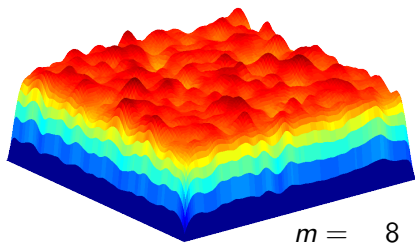
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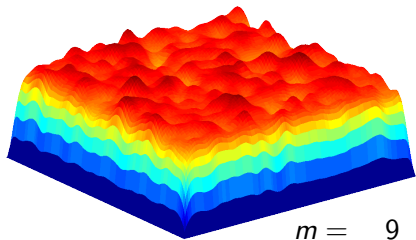
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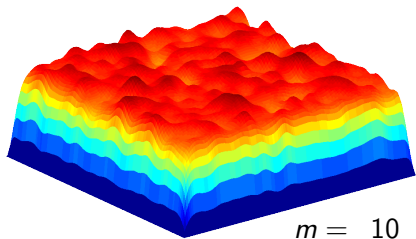
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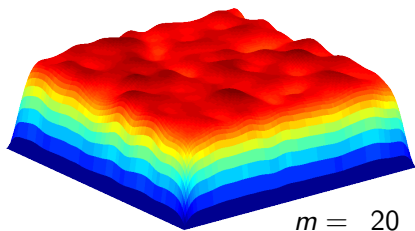
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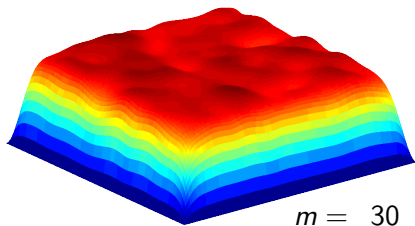
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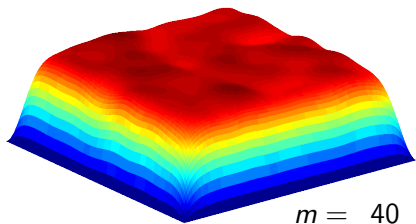
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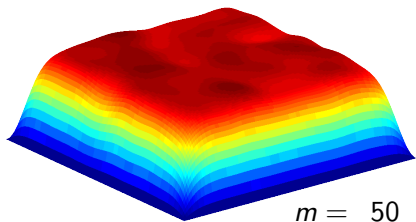
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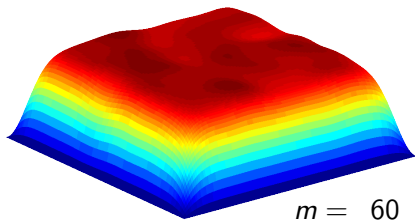
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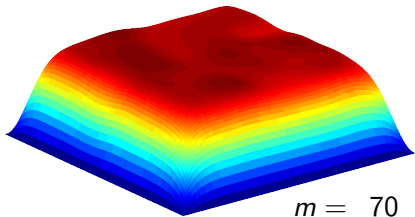
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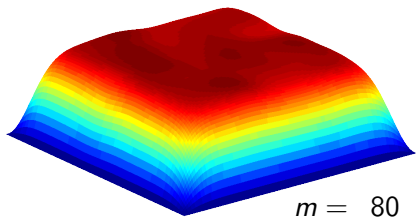
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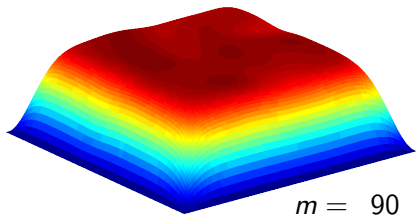
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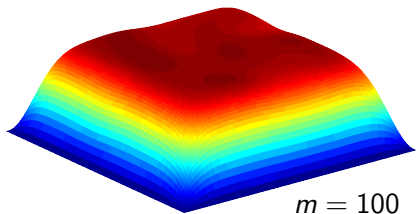
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Error definition

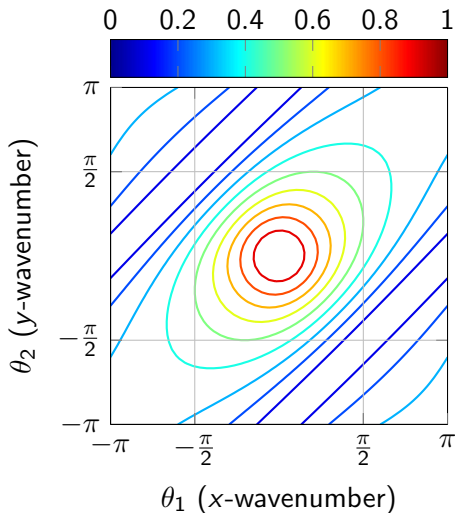
$$e^m = \bar{q} - q^m = \sum_i \varepsilon_i^m \psi_i(x) \quad \underbrace{\psi_i(x) = \exp(i\theta^T \mathbf{x}/h)}_{\text{eigenvectors decomposition}}$$

Error evolution

$$e^{m+1} = Me^m \quad \implies \quad |\varepsilon_{\theta}^{m+1}| = |\lambda_{\theta}| |\varepsilon_{\theta}^m|$$

$$|\lambda_{\theta}| > 1 \quad \rightarrow \quad \text{divergence}$$

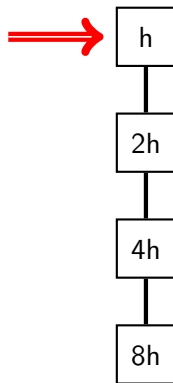
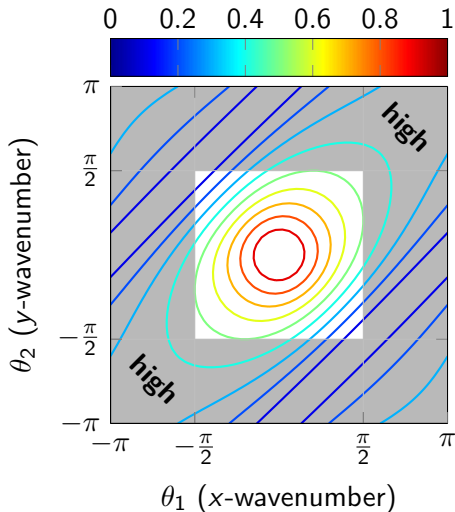
$$|\lambda_{\theta}| < 1 \quad \rightarrow \quad \text{convergence}$$



amplification factor $\rho = |\lambda_\theta|$ for the error amplitude: $|\varepsilon_\theta^{m+1}| = |\lambda_\theta| |\varepsilon_\theta^m|$

Multigrid — Gauss-Seidel

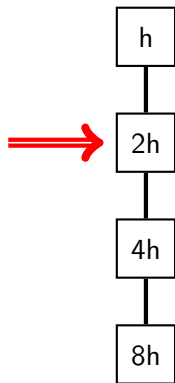
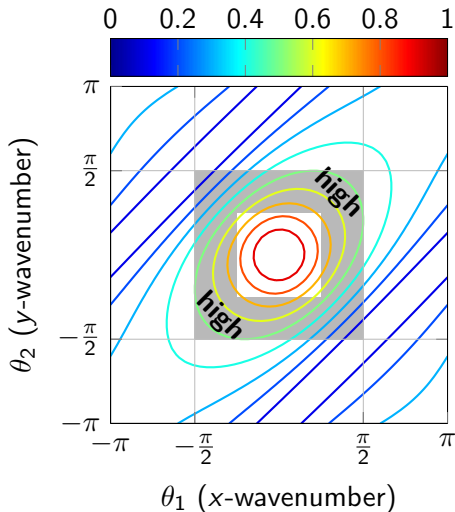
$$\Delta q = f$$



amplification factor $\rho = |\lambda_\theta|$ for the error amplitude: $|\varepsilon_\theta^{m+1}| = |\lambda_\theta| |\varepsilon_\theta^m|$

Multigrid — Gauss-Seidel

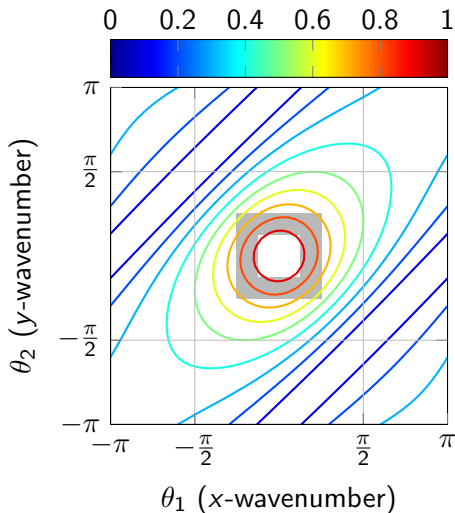
$$\Delta q = f$$



amplification factor $\rho = |\lambda_\theta|$ for the error amplitude: $|\varepsilon_\theta^{m+1}| = |\lambda_\theta| |\varepsilon_\theta^m|$

Multigrid — Gauss-Seidel

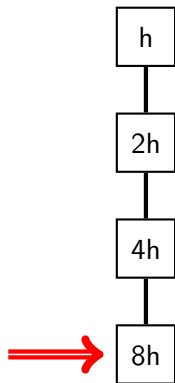
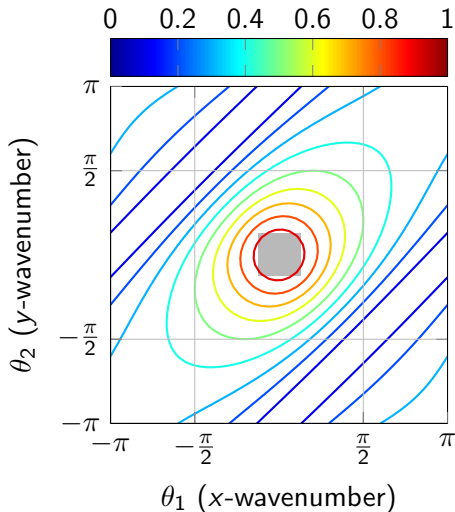
$$\Delta q = f$$



amplification factor $\rho = |\lambda_\theta|$ for the error amplitude: $|\varepsilon_\theta^{m+1}| = |\lambda_\theta| |\varepsilon_\theta^m|$

Multigrid — Gauss-Seidel

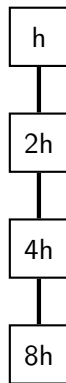
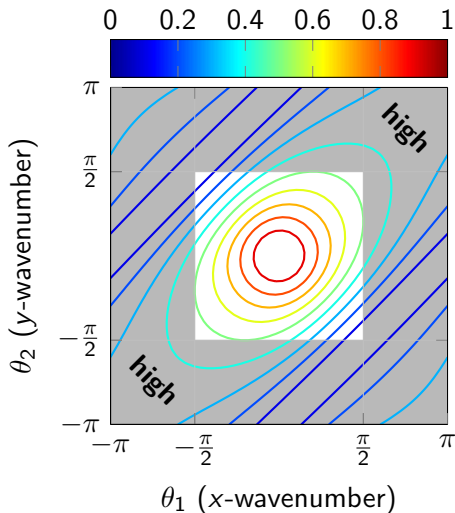
$$\Delta q = f$$



amplification factor $\rho = |\lambda_\theta|$ for the error amplitude: $|\varepsilon_\theta^{m+1}| = |\lambda_\theta| |\varepsilon_\theta^m|$

Multigrid — Gauss-Seidel

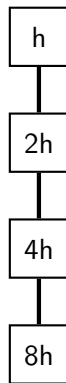
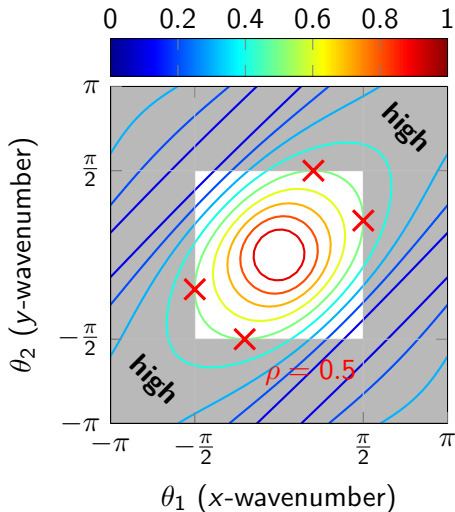
$$\Delta q = f$$



amplification factor $\rho = |\lambda_\theta|$ for the error amplitude: $|\varepsilon_\theta^{m+1}| = |\lambda_\theta| |\varepsilon_\theta^m|$

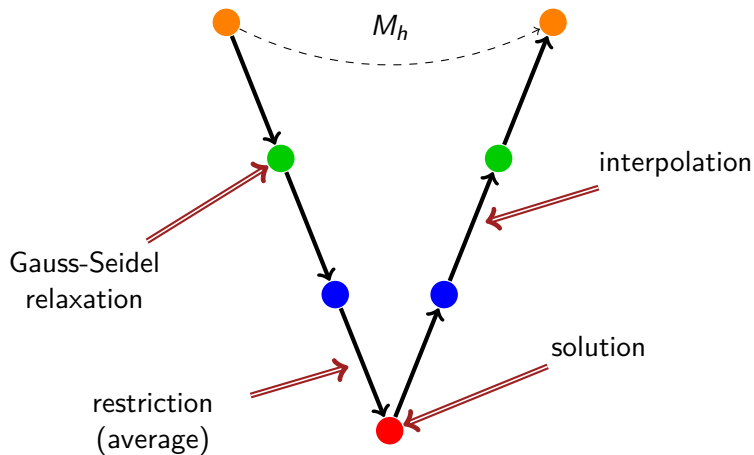
Multigrid — Gauss-Seidel

$$\Delta q = f$$

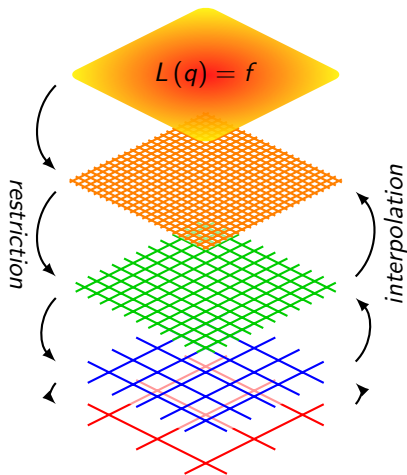


amplification factor $\rho = |\lambda_\theta|$ for the error amplitude: $|\varepsilon_\theta^{m+1}| = |\lambda_\theta| |\varepsilon_\theta^m|$

Multigrid — V-cycle



Multigrid — Full approximation scheme



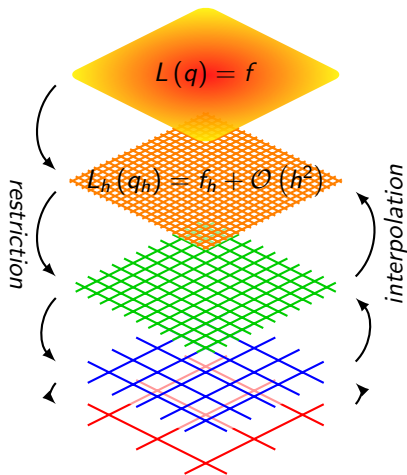
Ideas

- ▶ Represent the full solution on all grids
- ▶ τ_h^H : forcing of coarser grids' equations guarantees the same solution on all grids

Advantages

- ▶ Treatment of non-linear equations
- ▶ Natural approach to adaptive grid refinement

Multigrid — Full approximation scheme



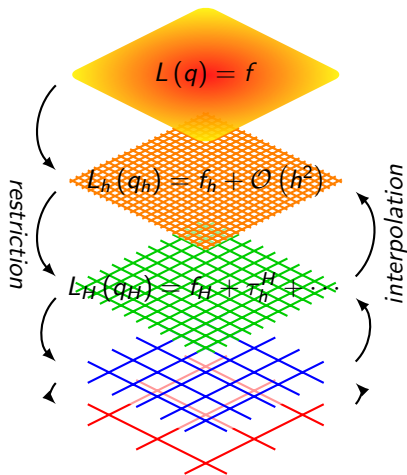
Ideas

- ▶ Represent the full solution on all grids
- ▶ τ_h^H : forcing of coarser grids' equations guarantees the same solution on all grids

Advantages

- ▶ Treatment of non-linear equations
- ▶ Natural approach to adaptive grid refinement

Multigrid — Full approximation scheme



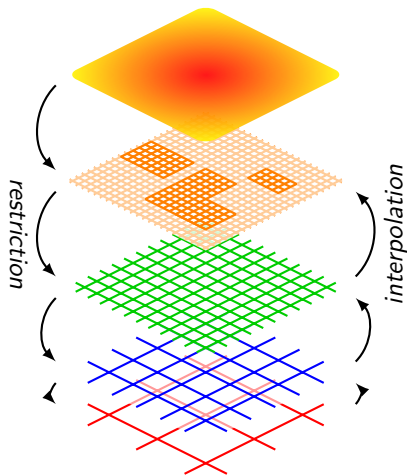
Ideas

- ▶ Represent the full solution on all grids
- ▶ τ_h^H : forcing of coarser grids' equations guarantees the same solution on all grids

Advantages

- ▶ Treatment of non-linear equations
- ▶ Natural approach to adaptive grid refinement

Multigrid — Full approximation scheme



Ideas

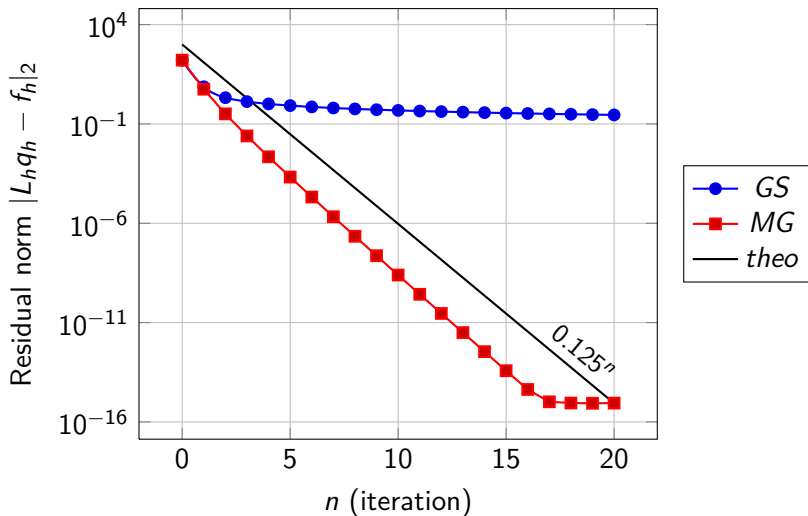
- ▶ Represent the full solution on all grids
- ▶ τ_h^H : forcing of coarser grids' equations guarantees the same solution on all grids

Advantages

- ▶ Treatment of non-linear equations
- ▶ Natural approach to adaptive grid refinement

Multigrid — Test case

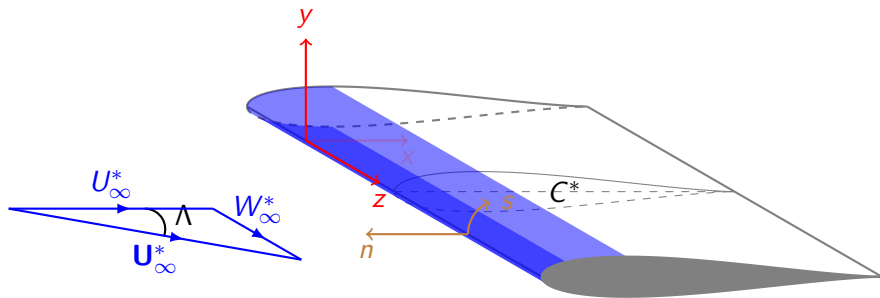
$$\Delta q = f$$



Outline

- ▶ Swept-wing instabilities
- ▶ Procedure and tools
- ▶ Multigrid
- ▶ **Base flow**
- ▶ Perturbations: a modal description
- ▶ Receptivity and sensitivity
- ▶ Conclusions and future work

Base Flow — Configuration

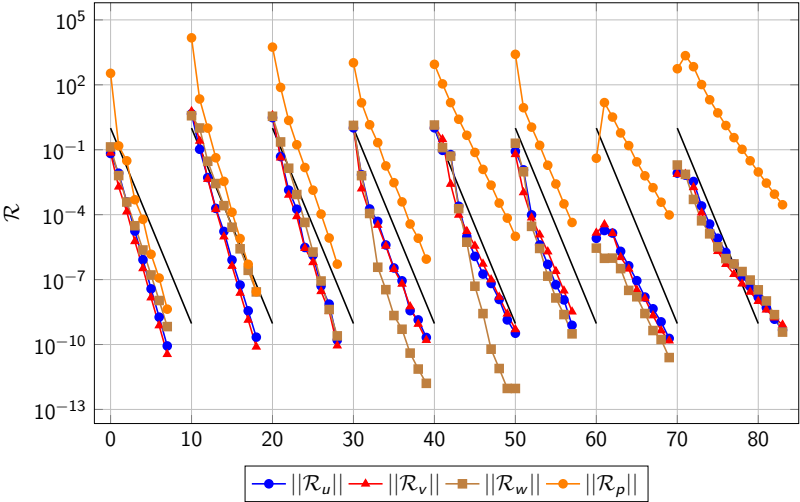


$$Re_C = 10^6 \quad \Lambda = 45^\circ$$

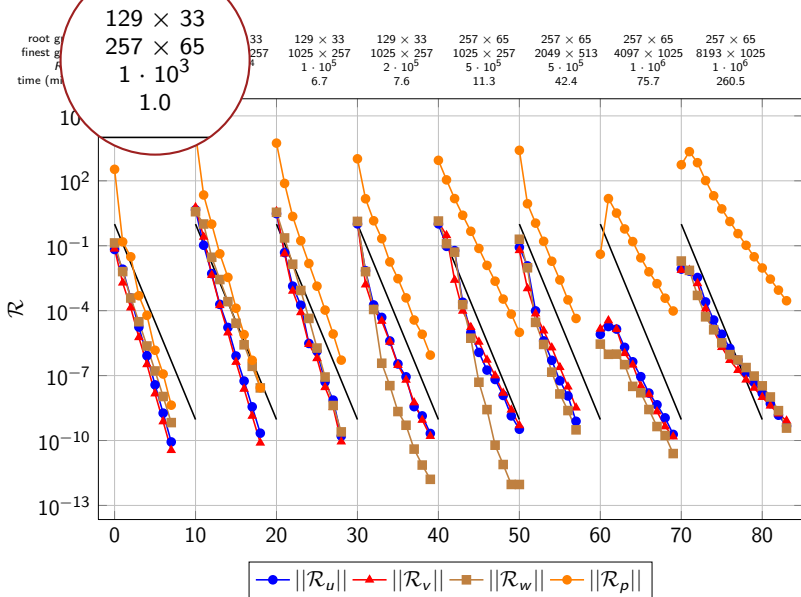
$$Re_r = 16000 \quad Re_s = 126$$

Base Flow — Computation

root grid =	129 × 33	129 × 33	129 × 33	129 × 33	257 × 65	257 × 65	257 × 65	257 × 65
finest grid =	257 × 65	1025 × 257	1025 × 257	1025 × 257	1025 × 257	2049 × 513	4097 × 1025	8193 × 1025
$Re_c =$	$1 \cdot 10^3$	$1 \cdot 10^4$	$1 \cdot 10^5$	$2 \cdot 10^5$	$5 \cdot 10^5$	$5 \cdot 10^5$	$1 \cdot 10^6$	$1 \cdot 10^6$
time (min) =	1.0	6.8	6.7	7.6	11.3	42.4	75.7	260.5



Base Flow — Computation

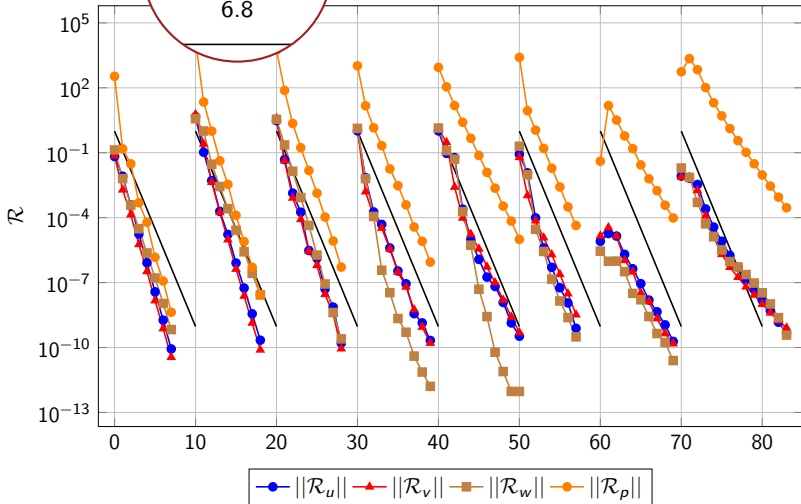


Base Flow — Computation

root grid = 129
 finest grid = 257
 $Re_c = 1$
 time (min) =

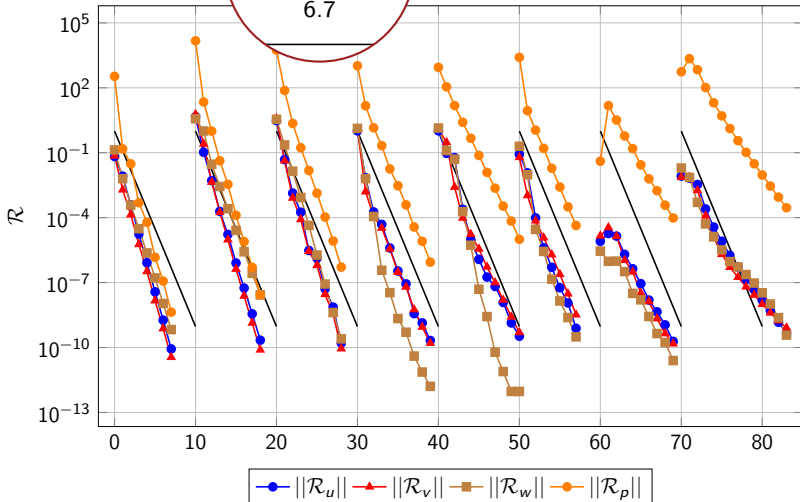
129×33
 1025×257
 $1 \cdot 10^4$
 6.8

33	129×33	257×65	257×65	257×65	257×65
257	1025×257	1025×257	2049×513	4097×1025	8193×1025
5	$2 \cdot 10^5$	$5 \cdot 10^5$	$5 \cdot 10^5$	$1 \cdot 10^6$	$1 \cdot 10^6$
	7.6	11.3	42.4	75.7	260.5



Base Flow — Computation

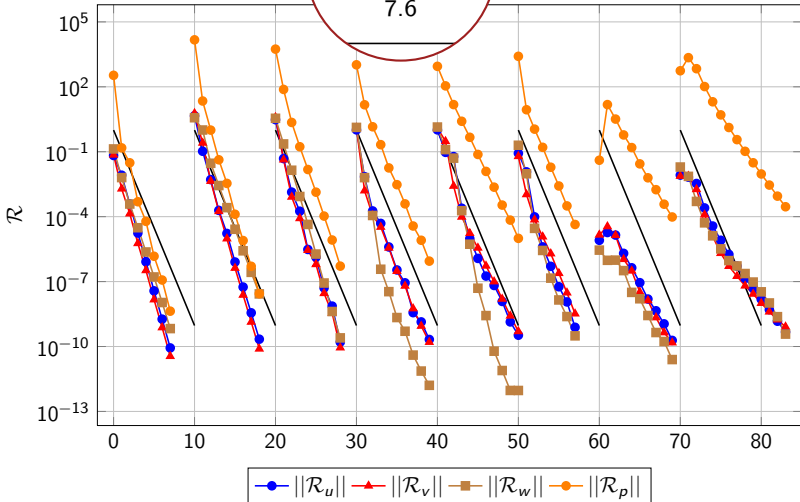
root grid =	129×33	129×33	129×33	129×33	129×33	129×33	129×33	129×33
finest grid =	257×65	1025×257	1025×257	1025×257	257×65	257×65	257×65	257×65
$Re_c =$	$1 \cdot 10^3$	$1 \cdot 10^5$	$1 \cdot 10^5$	$1 \cdot 10^5$	$5 \cdot 10^3$	$5 \cdot 10^3$	$5 \cdot 10^3$	$5 \cdot 10^3$
time (min) =	1.0	6.7	6.7	6.7	11.3	42.4	75.7	260.5



Base Flow — Computation

root grid =	129×33	129×33	129×33	129×33	129×33	129×33	129×33
finest grid =	257×65	1025×257	1025×257	1025×257	2049×513	4097×1025	8193×1025
$Re_c =$	$1 \cdot 10^3$	$1 \cdot 10^4$	$1 \cdot 10^4$	$1 \cdot 10^4$	$5 \cdot 10^3$	$1 \cdot 10^6$	$1 \cdot 10^6$
time (min) =	1.0	6.8	102	102	42.4	75.7	260.5

129×33
 1025×257
 $2 \cdot 10^5$
 7.6

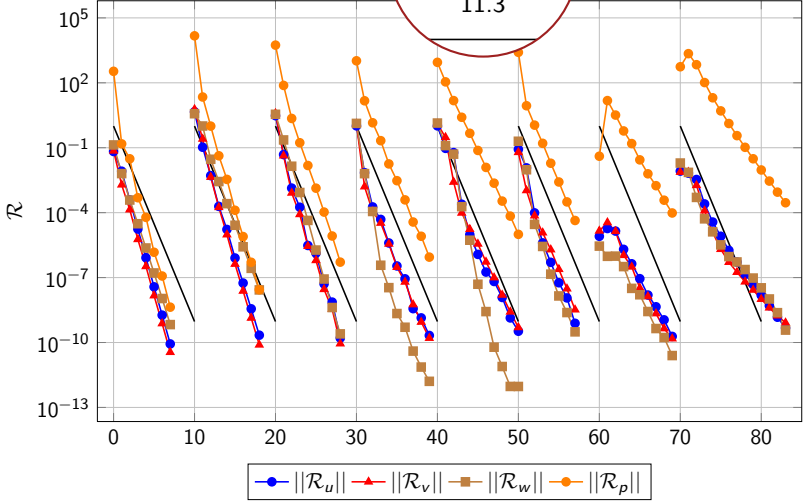


Base Flow — Computation

root grid =	129 × 33	129 × 33	129 × 33	129 × 33
finest grid =	257 × 65	1025 × 257	1025 × 257	1025 × 257
$Re_c =$	$1 \cdot 10^3$	$1 \cdot 10^4$	$1 \cdot 10^5$	$1 \cdot 10^6$
time (min) =	1.0	6.8	6.7	2

257×65
 1025×257
 $5 \cdot 10^5$
 11.3

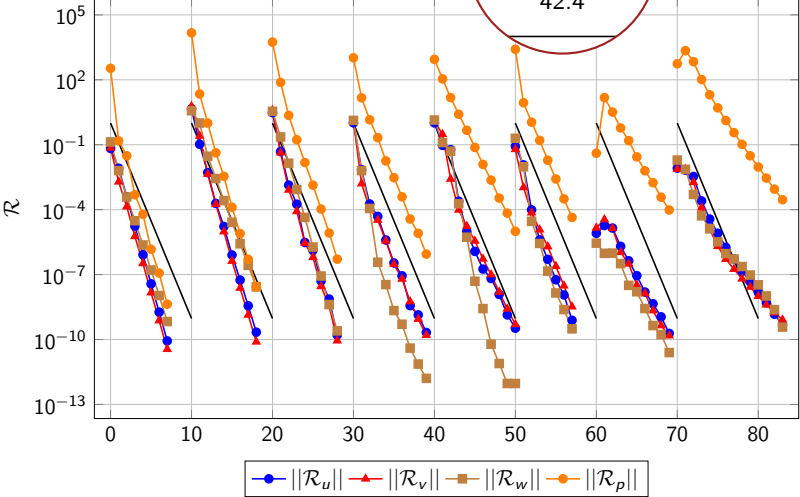
65	257×65	257×65
513	4097×1025	8193×1025
5	$1 \cdot 10^6$	$1 \cdot 10^6$
9	75.7	260.5



Base Flow — Computation

root grid =	129 × 33	129 × 33	129 × 33	129 × 33	257 × 65	257 × 65
finest grid =	257 × 65	1025 × 257	1025 × 257	1025 × 257	1025 × 257	8193 × 1025
$Re_c =$	$1 \cdot 10^3$	$1 \cdot 10^4$	$1 \cdot 10^5$	$2 \cdot 10^5$	$5 \cdot 10^5$	$1 \cdot 10^6$
time (min) =	1.0	6.8	6.7	7.6	42.4	260.5

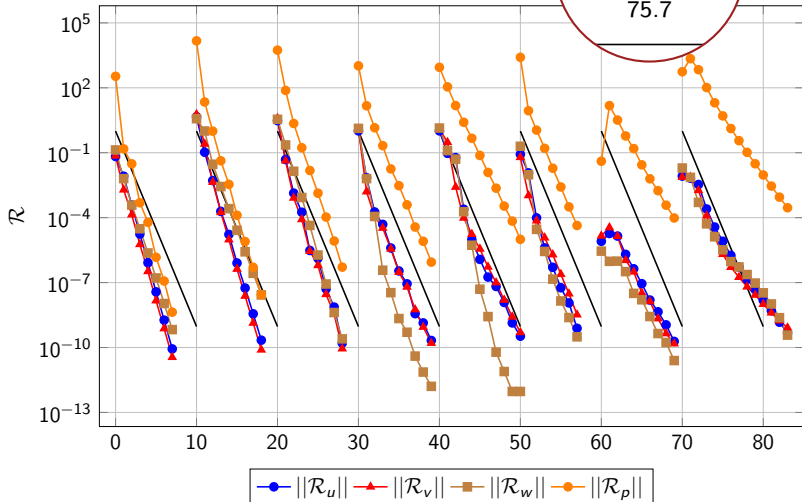
257×65
 2049×513
 $5 \cdot 10^5$
 42.4



Base Flow — Computation

root grid =	129×33	129×33	129×33	129×33	257×65	257×65
finest grid =	257×65	1025×257	1025×257	1025×257	1025×257	4097×1025
$Re_c =$	$1 \cdot 10^3$	$1 \cdot 10^4$	$1 \cdot 10^5$	$2 \cdot 10^5$	$5 \cdot 10^5$	$1 \cdot 10^6$
time (min) =	1.0	6.8	6.7	7.6	11.3	75.7

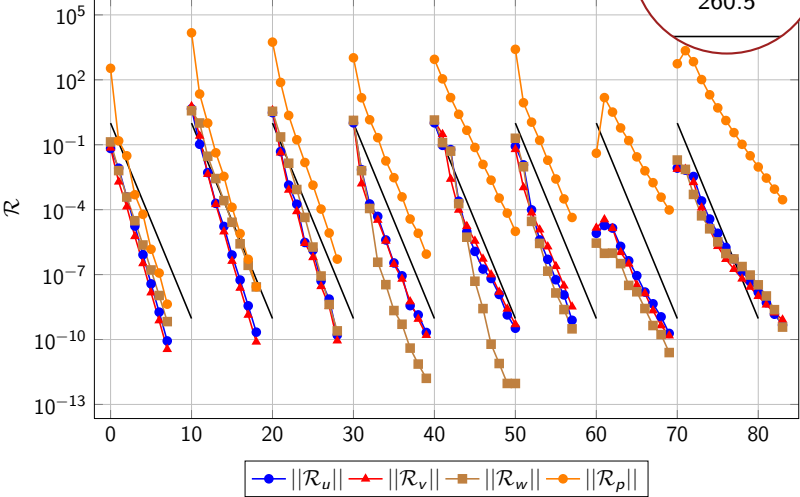
257×65
 4097×1025
 $1 \cdot 10^6$
 75.7



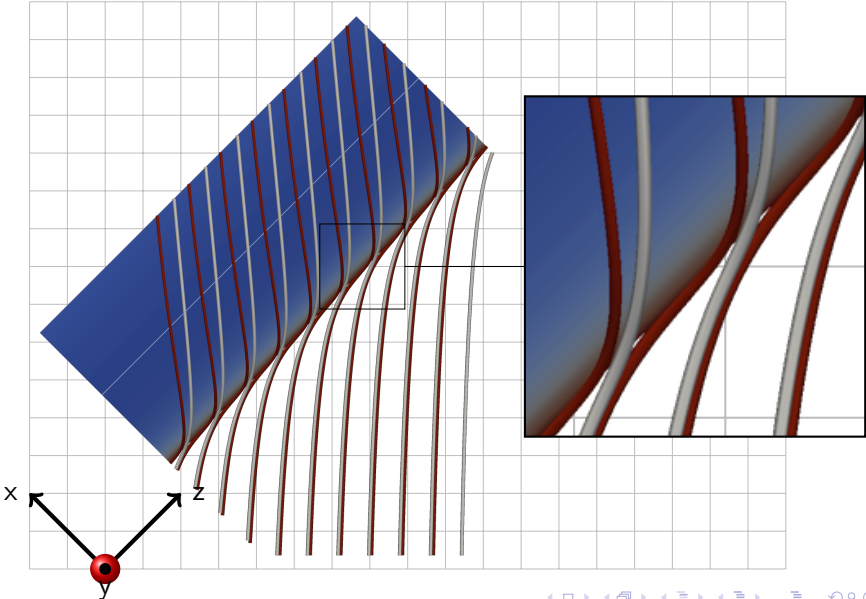
Base Flow — Computation

root grid =	129 × 33	129 × 33	129 × 33	129 × 33	257 × 65	257 × 65	257 × 65
finest grid =	257 × 65	1025 × 257	1025 × 257	1025 × 257	1025 × 257	2049 × 513	4097 × 1025
$Re_c =$	$1 \cdot 10^3$	$1 \cdot 10^4$	$1 \cdot 10^5$	$2 \cdot 10^5$	$5 \cdot 10^5$	$5 \cdot 10^5$	$1 \cdot 10^6$
time (min) =	1.0	6.8	6.7	7.6	11.3	42.4	160.0

257×65
 8193×1025
 $1 \cdot 10^6$
 260.5



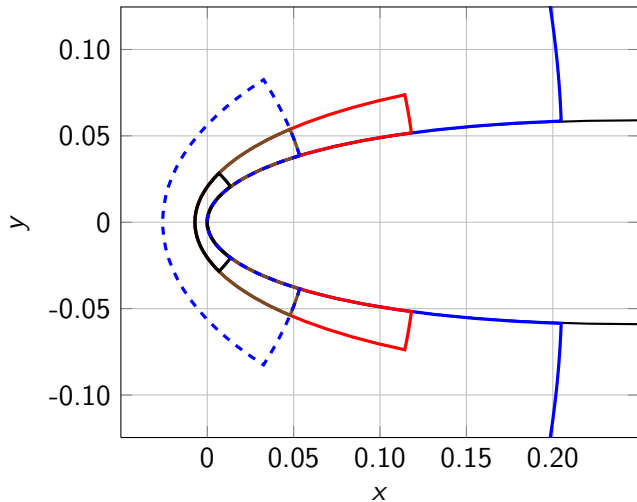
Base Flow — Streamlines



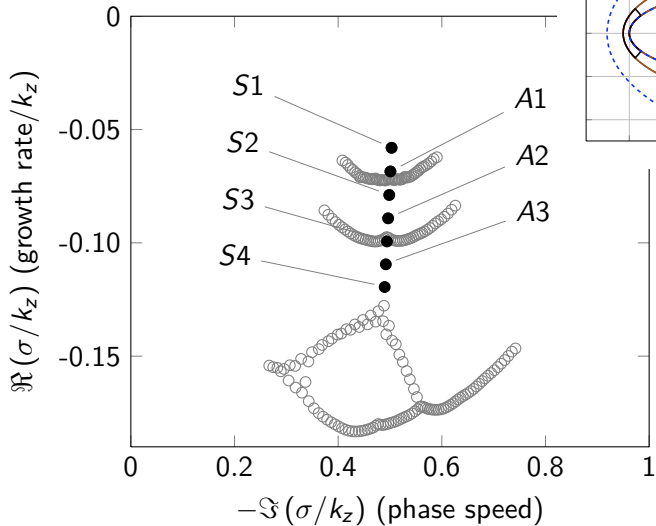
Outline

- ▶ Swept-wing instabilities
- ▶ Procedure and tools
- ▶ Multigrid
- ▶ Base flow
- ▶ **Perturbations: a modal description**
- ▶ Receptivity and sensitivity
- ▶ Conclusions and future work

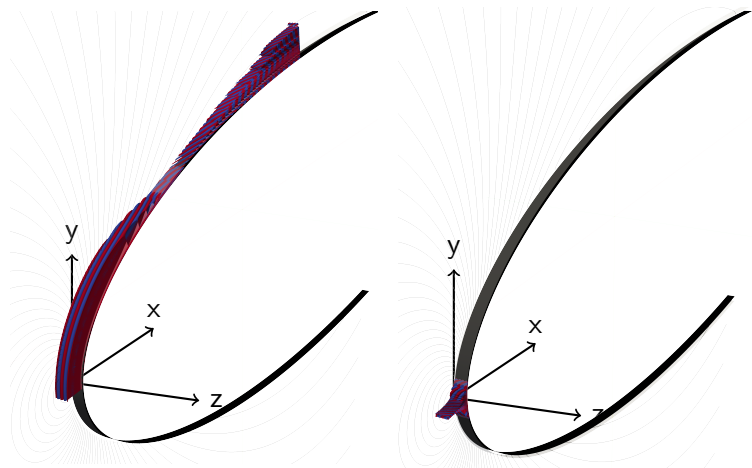
Perturbations — Domains



Perturbations — Spectrum

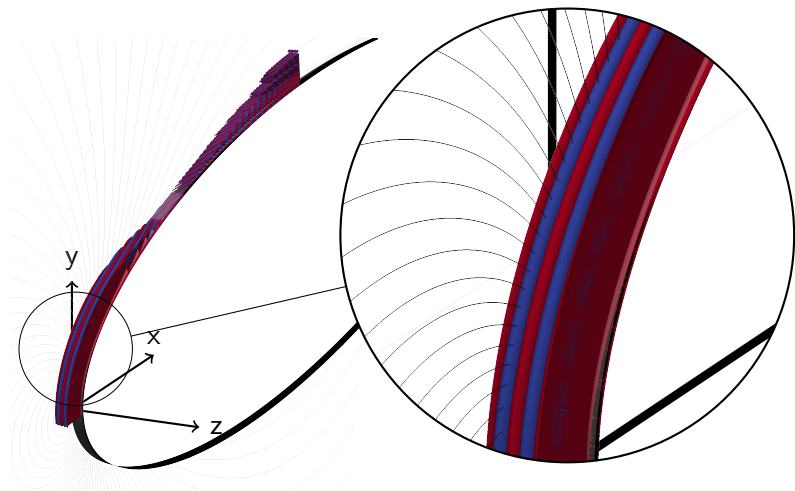


Perturbations — Direct and adjoint eigenvectors



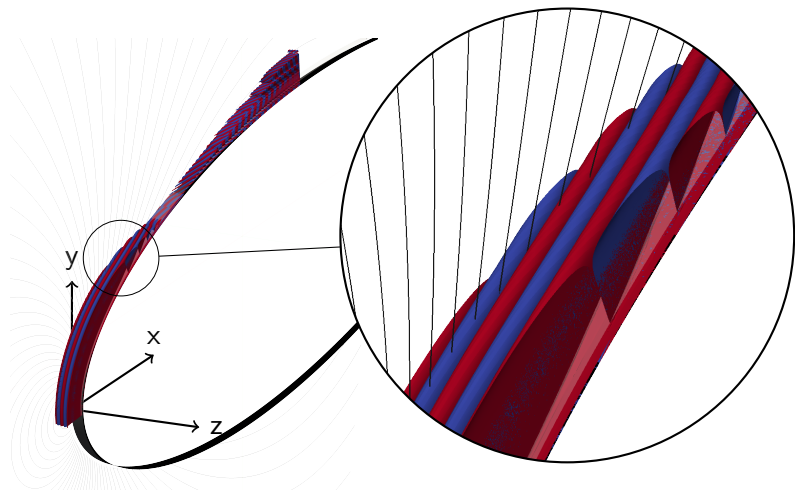
Iso-contours of the u -component of the perturbation. Blue: negative, red: positive

Perturbations — Direct and adjoint eigenvectors



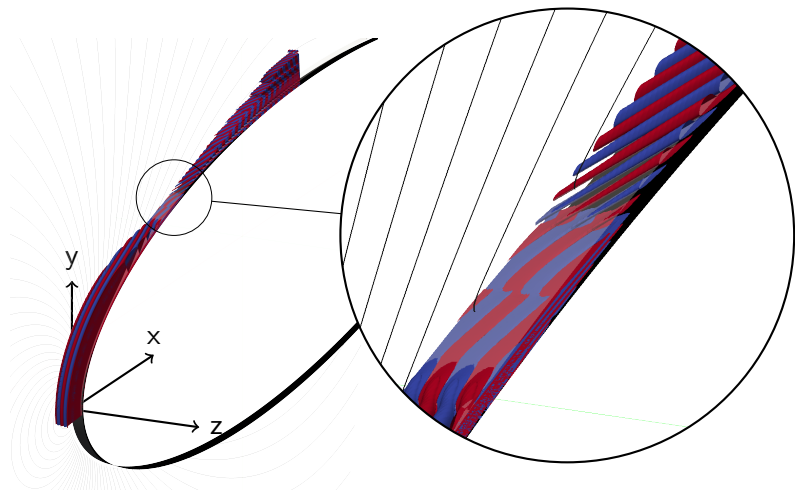
Direct eigenvector, attachment-line like structures

Perturbations — Direct and adjoint eigenvectors



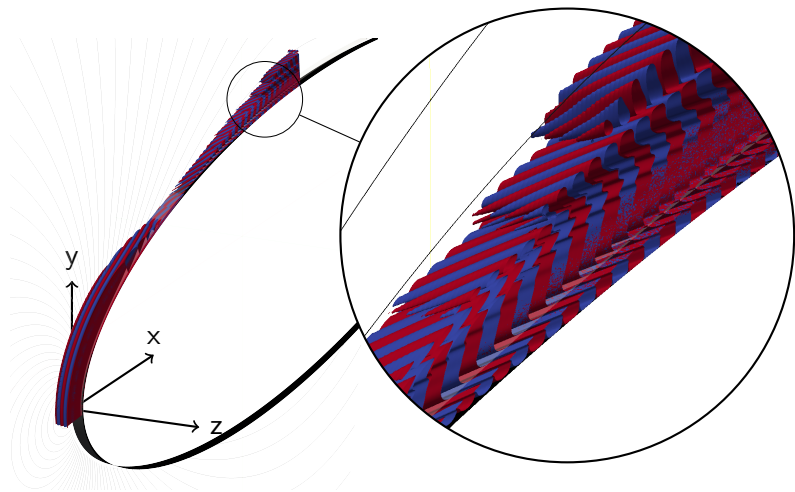
Direct eigenvector, structure orientation is against streamlines

Perturbations — Direct and adjoint eigenvectors



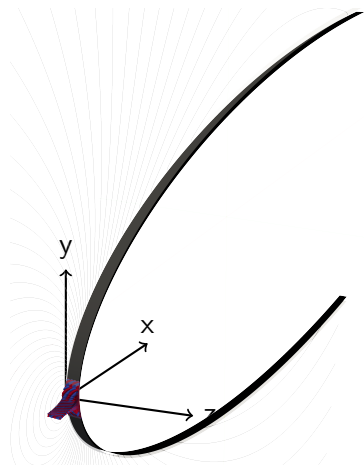
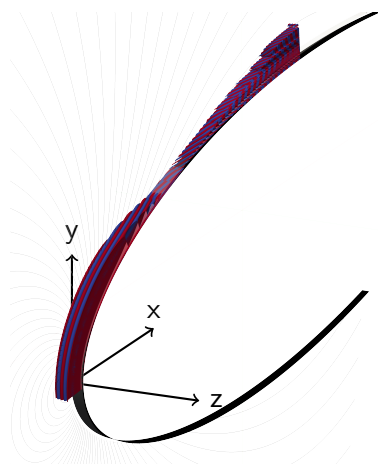
Direct eigenvector, transition from attachment line to crossflow like structures

Perturbations — Direct and adjoint eigenvectors

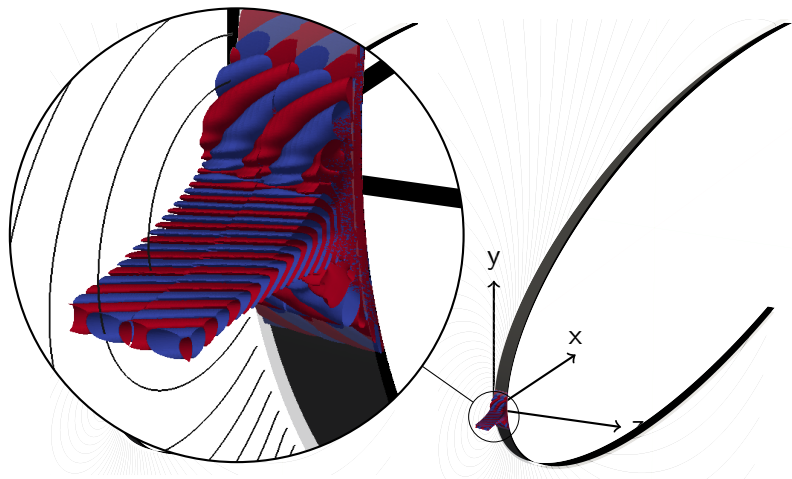


Direct eigenvector, crossflow like structures

Perturbations — Direct and adjoint eigenvectors

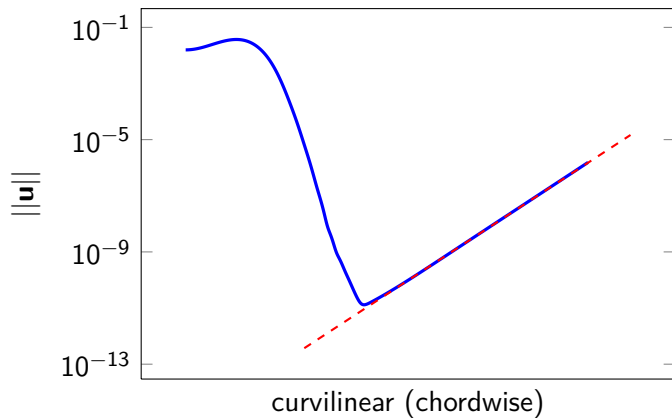


Perturbations — Direct and adjoint eigenvectors



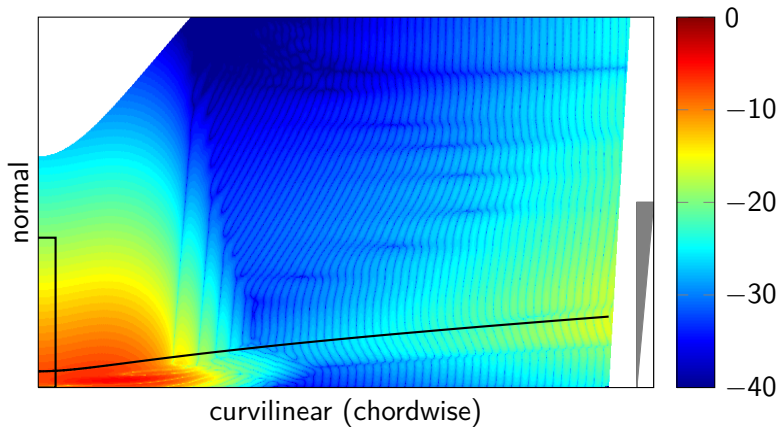
Adjoint eigenvector, localized upstream of and close to the attachment line region

Perturbations — Direct eigenvector



Direct eigenvector, norm of the velocity as a function of the chordwise coordinate

Perturbations — Direct eigenvector



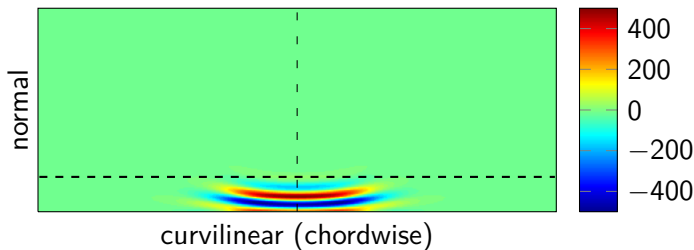
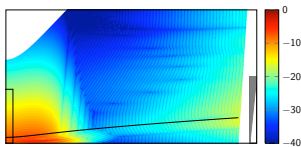
Direct eigenvector, u component at fixed spanwise position;
logarithmic scale in color

Outline

- ▶ Swept-wing instabilities
- ▶ Procedure and tools
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- ▶ Conclusions and future work

Perturbations — Receptivity

$$\delta(\text{amp}) = -\langle \hat{\mathbf{q}}^+, \delta \hat{\mathbf{f}} \rangle$$



Adjoint eigenvector, u^+ component at fixed spanwise location

Perturbations — Wavemaker

Sensitivity

$$\mathcal{I}(\text{obj}, \hat{\mathbf{q}}, \hat{\mathbf{q}}^+, \hat{\mathbf{f}}, A) = \text{obj} - \langle \hat{\mathbf{q}}^+, (\sigma B + A) \hat{\mathbf{q}} - \hat{\mathbf{f}} \rangle$$

$$\frac{\partial \mathcal{I}}{\partial A} \delta A = 0 \quad \Longrightarrow \quad \delta(\sigma) = \langle \hat{\mathbf{q}}^+, \delta A \hat{\mathbf{q}} \rangle$$

δA — variation in the operator

- ▶ base-flow change
- ▶ boundary conditions
- ▶ feedback forcing
- ▶ generic structural change

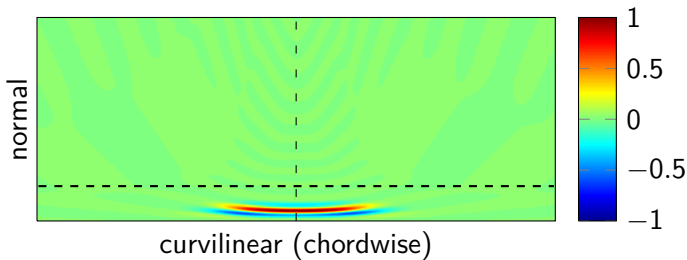
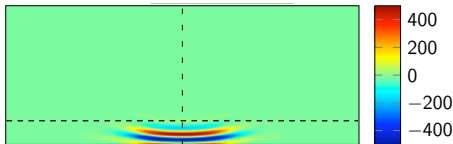
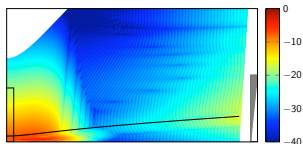
Feedback forcing

$$\delta A = C_0 \delta(x - x_0)$$

localized feedback forcing
dependent on the perturbation
e.g. small cylinder

Perturbations — Wavemaker

$$\delta\sigma = \langle \hat{\mathbf{q}}^+, \delta A \hat{\mathbf{q}} \rangle$$



Wavemaker at fixed spanwise location, $u^+ u$

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Conclusions

Numerics

- ▶ Multigrid has been proven as an extremely efficient solver for the non-linear Navier-Stokes equations on stretched grids and with adaptive refinement
- ▶ The Krylov subspace method implemented in SLEPc has been used to identify the flow's coherent structures
- ▶ The location of the outflow boundary has no influence on the stability results

Physics

- ▶ Direct modes show features of both attachment line and crossflow vortices: they share the same growth rate and phase speed
- ▶ The amplitude and growth rate of the global eigenvectors are dependent on a very small region a few boundary layer thicknesses across the attachment line
- ▶ The localization of the most sensitive region suggests local stability results are valid

Future work

Numerics

- ▶ Extension of multigrid to complex-valued linear problems will provide a fast and memory efficient solution strategy for eigenvector computations
- ▶ Multigrid refinement strategies for eigenvector computations
- ▶ Provide documentation and clean up/optimize the code (parallelization?)

Physics

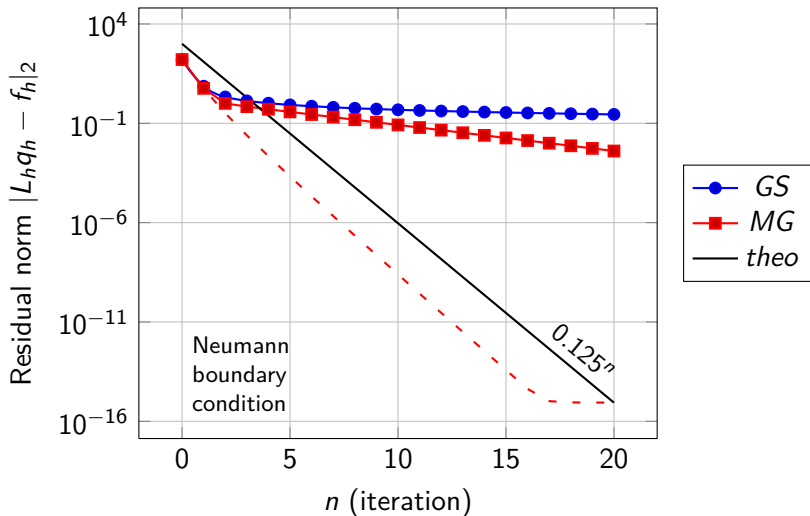
- ▶ The role of the adjoint's boundary values regarding receptivity to boundary conditions
- ▶ Identification of physical mechanisms leading to the transition between attachment line and crossflow vortices
- ▶ Parametric study and identification of the critical Reynolds number

Thanks



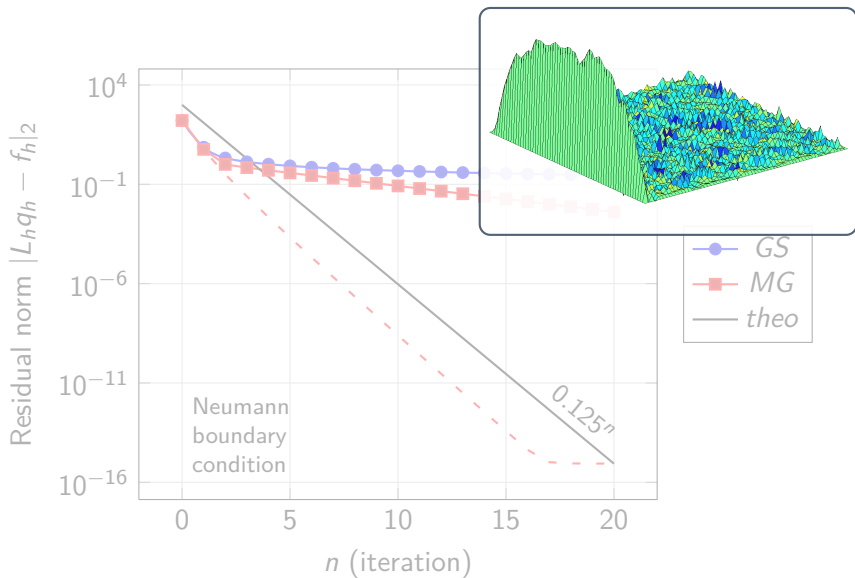
Gauss-Seidel vs. Multigrid

$$\Delta q = f, \quad \partial_n q = 0$$



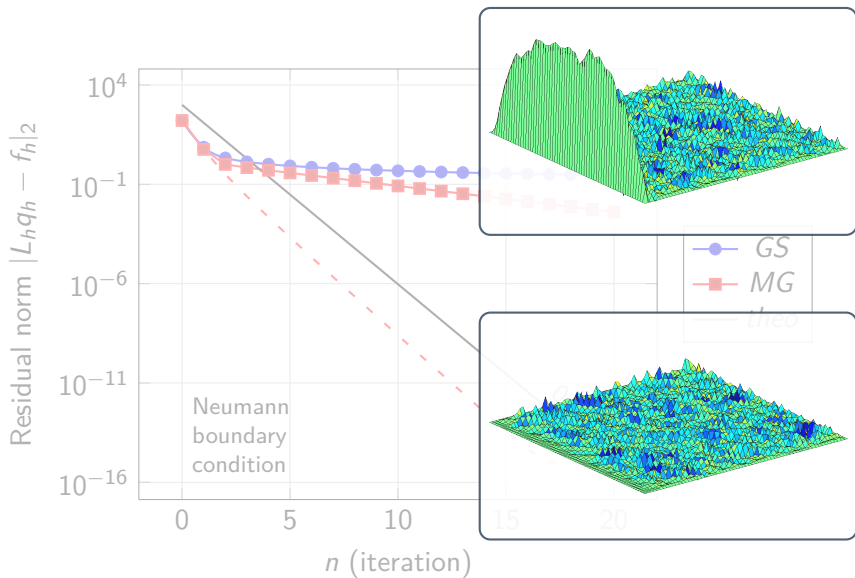
Gauss-Seidel vs. Multigrid

$$\Delta q = f, \quad \partial_n q = 0$$



Gauss-Seidel vs. Multigrid

$$\Delta q = f, \quad \partial_n q = 0$$



Gauss-Seidel vs. Multigrid

$$\Delta q = f, \quad \partial_n q = 0$$

