

Mathematical Methods for Marine Energy Extraction

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Time-parallelization of sequential DA problems

- Luenberger observer
- Time-parallelization setting
- Parareal algorithm
- Diamond strategy (Parareal case)

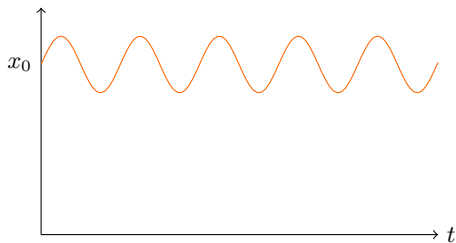
Bathymetry optimization

- Derivation of the wave model
- Description of the optimization problem
- Continuous optimization problem
- Numerical examples

Mathematical analysis of the BEM theory

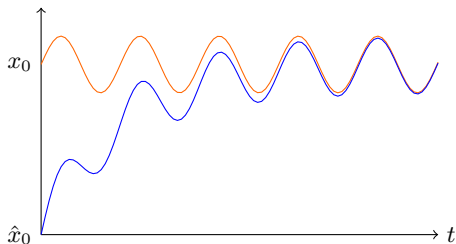
- Glauert's modeling
- Simplified and corrected models
- Solving algorithms
- Optimization

$$(1) \quad \begin{cases} \dot{x}(t) = Ax(t) + Bu(t), \\ x(0) = x_0 \text{ unknown}, \\ y(t) = Cx(t). \end{cases}$$



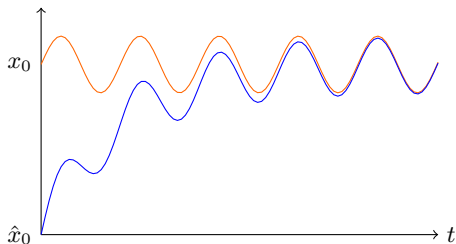
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Is there a *convenient* way to choose the *observer gain* L ?

Note that

$$(2) \iff \begin{cases} \dot{\hat{x}}(t) = (A - LC)\hat{x}(t) + (Bu(t) + Ly(t)), \\ \hat{x}(0) = \hat{x}_0. \end{cases}$$

and then $x(t) - \hat{x}(t) = e^{(A-LC)t} (x(0) - \hat{x}(0))$

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THEOREM ► IDENTITY OBSERVER THEOREM [LUENBERGER]

Given a completely *observable* system (1), an identity observer of the form (2) can be constructed, and the coefficients of the **characteristic polynomial of the observer** can be selected arbitrarily.

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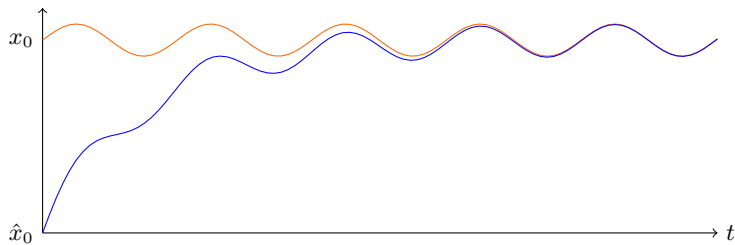
PROPOSITION 1.1

We assume System (1) is observable and the eigenvalues of $A - LC$ are negative and simple. Then, we have

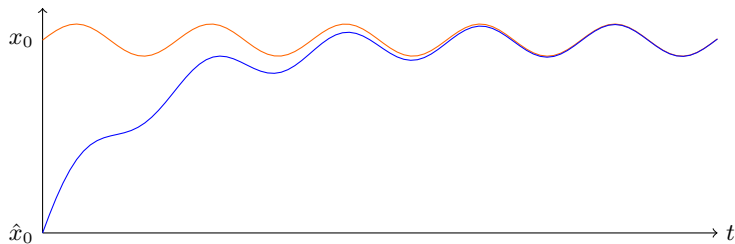
$$\|e^{(A-LC)t}\| \leq \gamma e^{-\mu t}$$

with $\mu := \min_{\nu \in \sigma(A-LC)} |\nu|$ and $\gamma := \text{cond}(V) = \|V^{-1}\| \|V\|$, where V is the matrix whose rows are the eigenvectors of $A - LC$ and $\|\cdot\|$ represents the induced 2-norm of a matrix.

DIAMOND STRATEGY



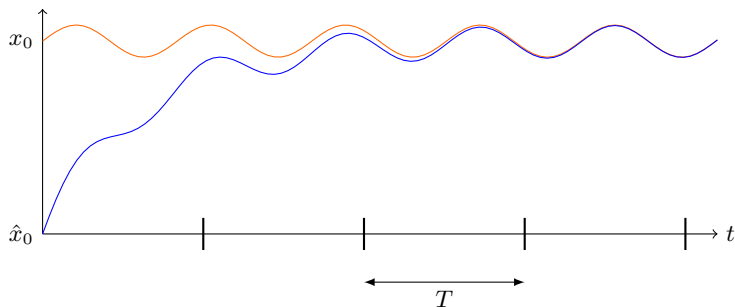
DIAMOND STRATEGY



Luenberger observer

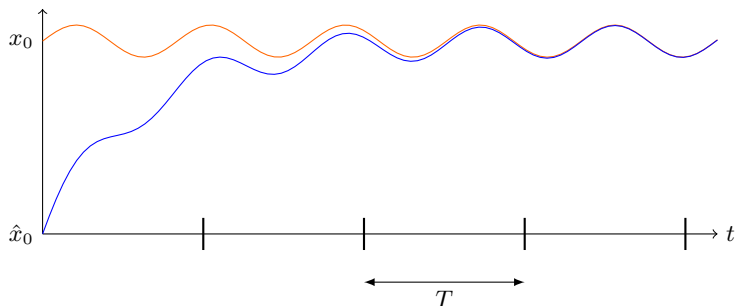
Time-parallel method

DIAMOND STRATEGY



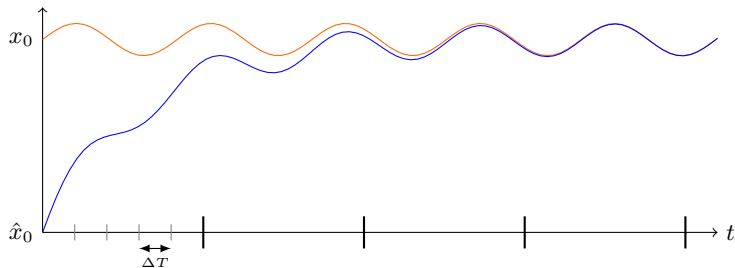
- Divide the time interval into *windows* W_ℓ of a given length $T > 0$.

DIAMOND STRATEGY



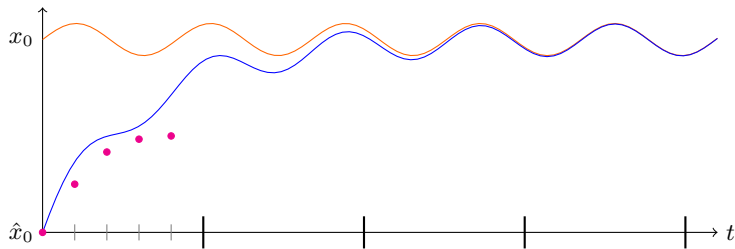
- ▶ Divide the time interval into *windows* W_ℓ of a given length $T > 0$.
- ▶ Solve Equation (2) on each window, in a sequential order, using a time-parallel algorithm.

DIAMOND STRATEGY (TIME-PARALLELIZATION)



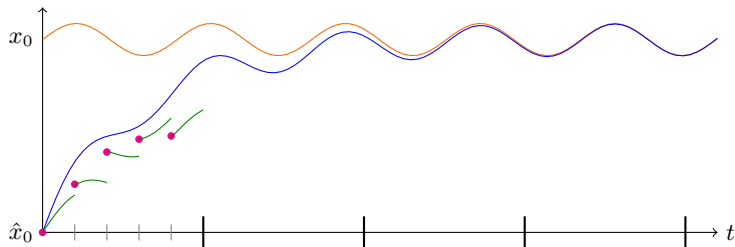
- Decompose W_ℓ into N *subintervals* of length ΔT .

DIAMOND STRATEGY (TIME-PARALLELIZATION)



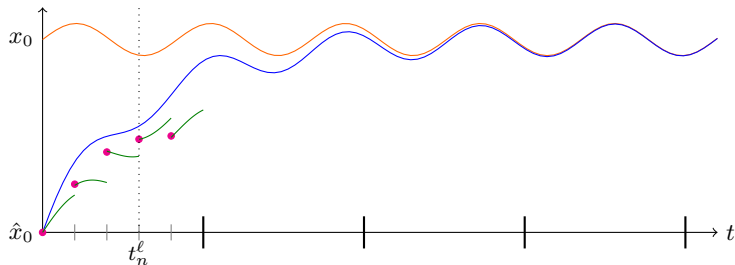
- Parallelizing in time requires the introduction of initial conditions $\hat{X}_{\ell,n}^h$.

DIAMOND STRATEGY (TIME-PARALLELIZATION)



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- ▶ We then construct a parallel version $\hat{x}_{\parallel}(t)$ of Equation (2) in each subinterval.

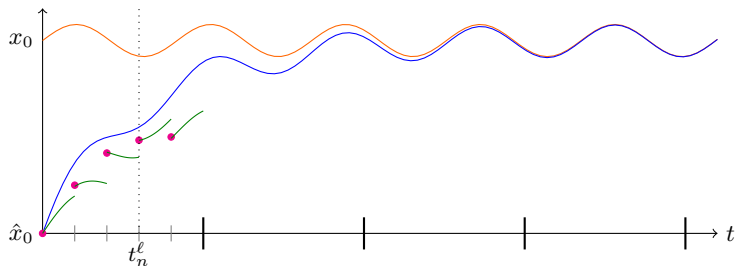
DIAMOND STRATEGY (TIME-PARALLELIZATION)



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$$J_{\ell,n}^h := \hat{X}_{\ell,n}^h - \hat{x}_{\parallel}(t_n^{\ell-}).$$

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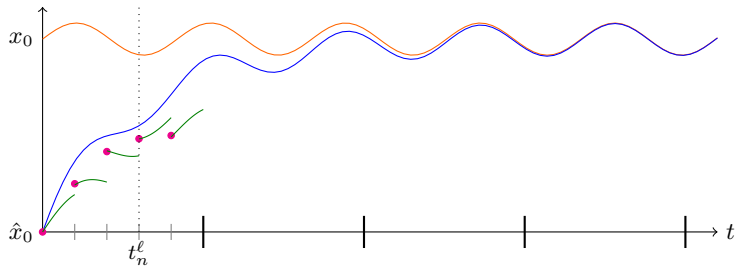


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Next step : define a suitable stopping criterion !

DIAMOND STRATEGY (STOPPING CRITERION)



LEMMA

Under the assumptions of Proposition 1.1, we have

$$\left\| (x - \hat{x}_\parallel)(t_n^{\ell-}) \right\| \leq \gamma \left(e^{-\mu t_n^\ell} \|x(0) - \hat{x}(0)\| + e^{-\mu \Delta T} \|J_{\ell, n-1}^h\| \right).$$

PROPOSITION 1.2 ► A POSTERIORI ESTIMATE

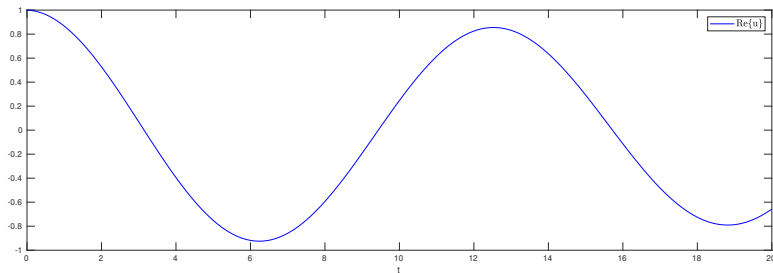
Let us assume that h is obtained from the stopping criterion in W_ℓ

$$\max_{1 \leq n \leq N} \|J_{\ell,n}^h\| \leq \tilde{\gamma} e^{-\mu \ell T}$$

where $\tilde{\gamma}$ is an arbitrary parameter. Then, the rate of convergence of $\hat{x}_\parallel(t)$ to $x(t)$ is bounded by μ , i.e.

$$\left\| (x - \hat{x}_\parallel)(t_n^\ell) \right\| \leq \gamma e^{-\mu \Delta T} (\|x(0) - \hat{x}(0)\| + \tilde{\gamma}) e^{-\mu \ell T}.$$

PARAREAL ALGORITHM



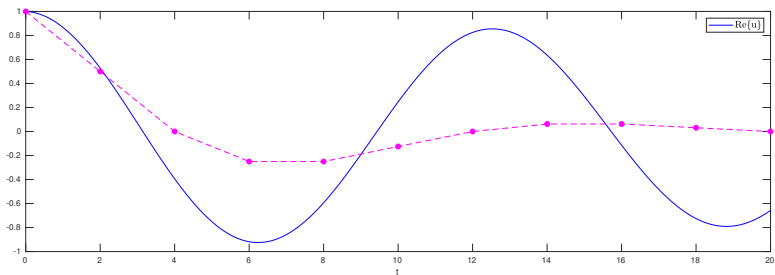
Dahlquist equation $\dot{u}(t) = -\frac{i}{2}u$ in $[0, 20]$

To solve the problem

$$\begin{cases} \dot{u}(t) = f(u(t)), & t \in [0, T] \\ u(0) = u_0 \end{cases}$$

we decompose the time interval on N subintervals, denoted by (t_{n-1}, t_n) .

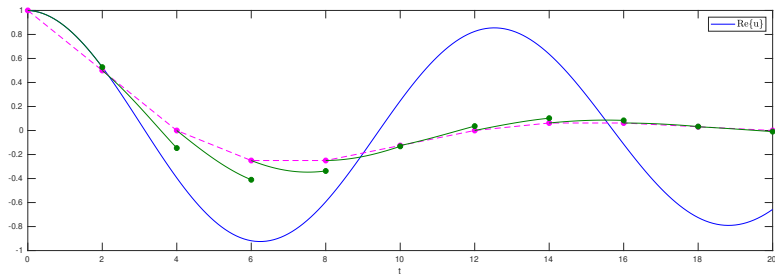
PARAREAL ALGORITHM



- Impose arbitrary values on the subintervals by using the **coarse solver** \mathcal{G} :

$$U_0^0 = u_0, \quad U_n^0 = \mathcal{G}(t_n, t_{n-1}, U_{n-1}^0).$$

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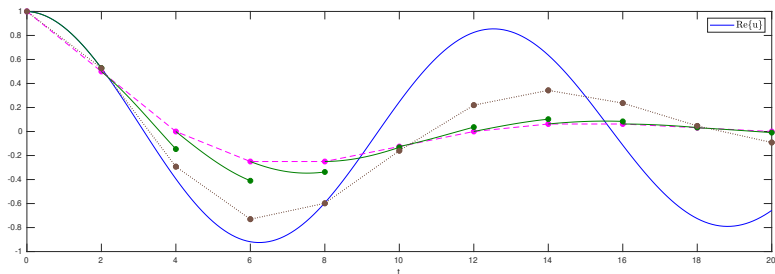
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- Using the **fine solver** \mathcal{F} , solve in parallel

$$\begin{cases} \dot{u}(t) = f(u(t)), & t \in [t_{n-1}, t_n] \\ u(t_{n-1}) = U_{n-1}^0. \end{cases}$$

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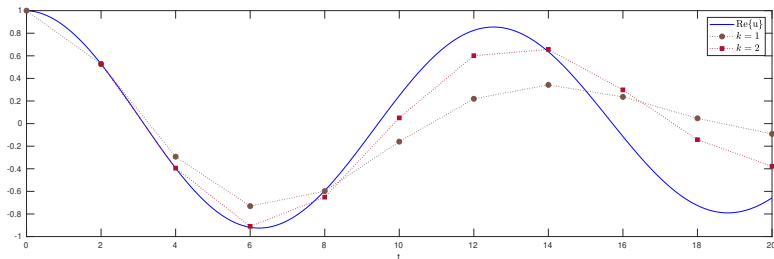
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- Smooth the discontinuities previously introduced by defining

$$U_n^1 := \mathcal{F}(t_n, t_{n-1}, U_{n-1}^0) + \mathcal{G}(t_n, t_{n-1}, U_{n-1}^1) - \mathcal{G}(t_n, t_{n-1}, U_{n-1}^0).$$

PARAREAL ALGORITHM



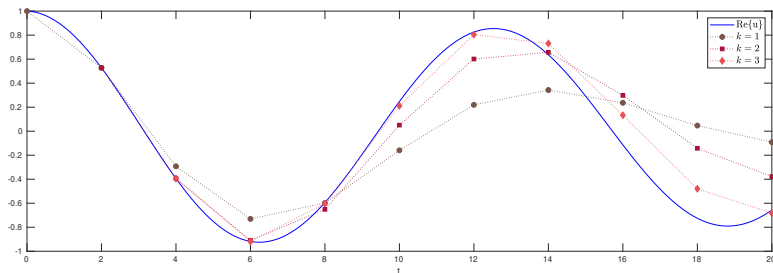
At iteration k :

- compute $\{\mathcal{F}(t_n, t_{n-1}, U_{n-1}^{k-1})\}_{n=1}^N$ in parallel.
- Update the sequence

$$U_n^k := \mathcal{F}(t_n, t_{n-1}, U_{n-1}^{k-1}) + \mathcal{G}(t_n, t_{n-1}, U_{n-1}^k) - \mathcal{G}(t_n, t_{n-1}, U_{n-1}^{k-1})$$

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What about its convergence ?

THEOREM ► CONVERGENCE OF PARAREAL [GANDER AND HAIRER]

(...) at iteration k of the Parareal algorithm, we have the bound

$$\|u(t_n) - U_n^k\| \leq \frac{C_3}{C_1} \frac{(C_1 \Delta T^{p+1})^{k+1}}{k!} (1 + C_2 \Delta T)^{n-(k+1)} \prod_{j=0}^k (n - j).$$

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- **Superlinear** rate of convergence.

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- Superlinear rate of convergence.
- Among other assumptions, $\mathcal{F}(t_n, t_{n-1}, U_{n-1}^k)$ is **the exact solution** on (t_{n-1}, t_n) , and \mathcal{G} must satisfy

$$\|\mathcal{G}(t + \Delta T, t, x) - \mathcal{G}(t + \Delta T, t, y)\| \leq (1 + C_2 \Delta T) \|x - y\|.$$

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- The result is well suited for **non-decaying problems**.

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THEOREM 1.3 ► CONVERGENCE OF PARAREAL FOR DECAYING PROBLEMS [KWOK, RIFFO AND SALOMON]

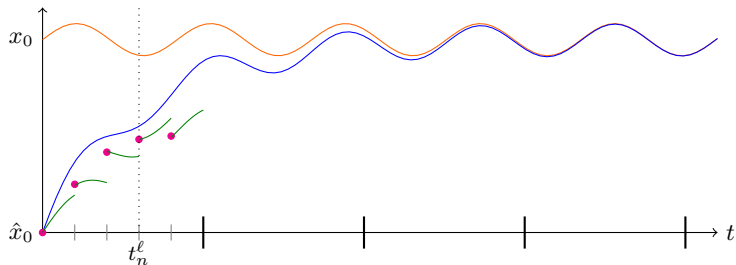
(...) We also assume that \mathcal{F} and \mathcal{G} are Lipschitz with respect to the initial conditions:

$$\max \{ \|\mathcal{F}(t_n, t_{n-1}, y) - \mathcal{F}(t_n, t_{n-1}, z)\|, \|\mathcal{G}(t_n, t_{n-1}, y) - \mathcal{G}(t_n, t_{n-1}, z)\| \} \leq \varepsilon \|y - z\|,$$

for a constant $\varepsilon \in (0, 1)$. Then, after k iterations of the Parareal algorithm, we have

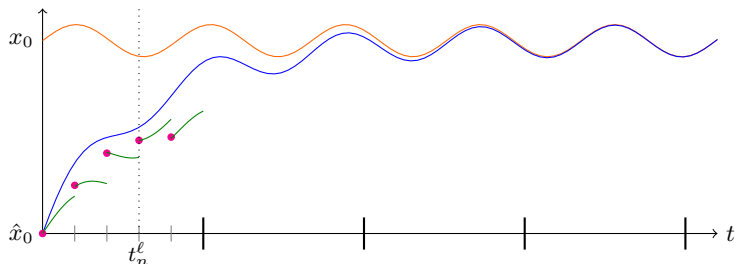
$$\|U_n^k - u(t_n)\| \leq \begin{cases} 0 & n \leq k \\ \alpha \beta^k \sum_{i=0}^{n-k-1} \binom{k+i}{k} \varepsilon^i & n > k. \end{cases}$$

DIAMOND STRATEGY (PARAREAL CASE)



- The Luenberger observer $\hat{x}(t)$ is a decaying problem (Proposition 1.1).

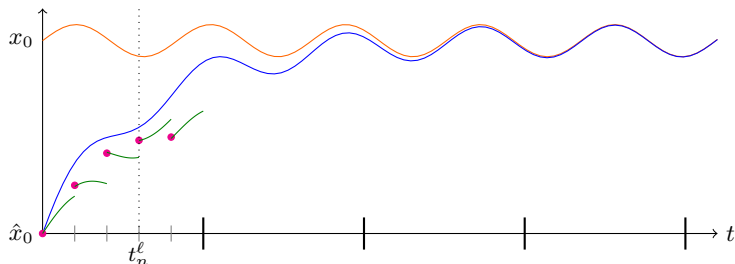
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- The number of parareal iterations $\{k_\ell\}_\ell$ can be determined from
 - (a) Proposition 1.2 (a *a posteriori* estimate)

$$\max_{1 \leq n \leq N} \left\| \hat{X}_{\ell,n}^{k_\ell} - \hat{x}_\parallel(t_n^\ell) \right\| \leq \tilde{\gamma} e^{-\mu \ell T}, \quad \tilde{\gamma} \text{ arbitrary.}$$

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- (b) Theorem 1.3 (*a priori* bound)

$$\left\| \hat{X}_{\ell,n}^{k_\ell} - \hat{x}_\parallel(t_n^\ell) \right\| \leq \alpha \beta^{k_\ell} \sum_{i=0}^{n-k_\ell-1} \binom{k_\ell+i}{k_\ell} \varepsilon^i, \quad n > k_\ell.$$

DIAMOND STRATEGY (PARAREAL CASE)

THEOREM 1.4

We keep the assumptions of Proposition 1.1 and Theorem 1.3. For a window W_ℓ and $\tilde{\gamma} > 0$, we define

$$k_\ell = \begin{cases} \min S_\ell & S_\ell \neq \emptyset \\ k_{\ell-1} & S_\ell = \emptyset \end{cases}$$

where

$$S_\ell = \left\{ k \in \mathbb{N}^*, k \leq N-1 : \alpha\beta^k \sum_{i=0}^{N-k-1} \binom{k+i}{k} \varepsilon^i \leq \tilde{\gamma} e^{-\mu \ell T} \cdot \frac{1-\varepsilon}{\alpha(1-\varepsilon^N)} \right\}.$$

Suppose that we apply the *Diamond strategy* using k_ℓ iterations of the Parareal algorithm. Then, the stopping criterion is satisfied.

We define the efficiency of the algorithm as

$$E = \frac{\tau_s}{N\tau_p}$$

where τ_s , τ_p are the CPU times required to reach a given tolerance Tol by using a sequential and parallel solver, respectively; and N represents the number of available processors.

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THEOREM

The estimated efficiency of the *Diamond Strategy* is given by

$$E^{th} = \frac{\ell^* \tau_{\Delta T}^{\mathcal{F}}}{\tau_{\Delta T}^{\mathcal{F}} + N\tau_{\Delta T}^{\mathcal{G}}} \left(\sum_{\ell=0}^{\ell^*-1} k_{\ell} \right)^{-1},$$

where $\tau_{\Delta T}^{\mathcal{F}}$, $\tau_{\Delta T}^{\mathcal{G}}$ represents the amount of time spent in solving (2) over an interval of size ΔT with \mathcal{F} and \mathcal{G} , respectively, and

$$\ell^* := \min \left\{ \ell \in \mathbb{N} : \left(\tilde{\gamma}(1 + e^{-\mu\Delta T}) + \gamma e^{-\mu\Delta T} \|x(0) - \hat{x}(0)\| \right) e^{-\mu\ell T} \leq \text{Tol} \right\}.$$

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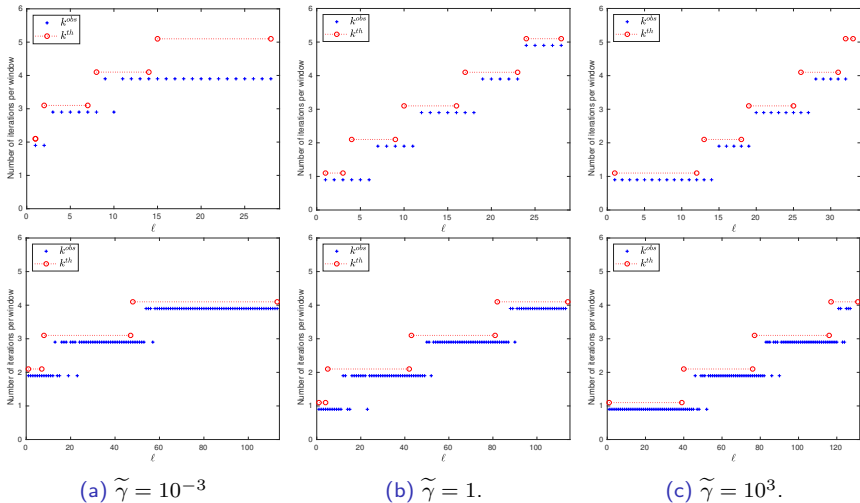
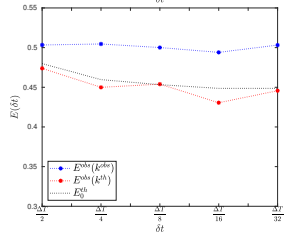
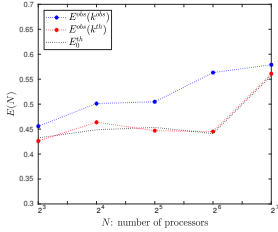
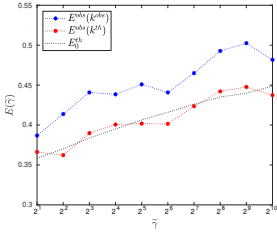
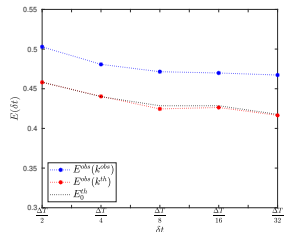
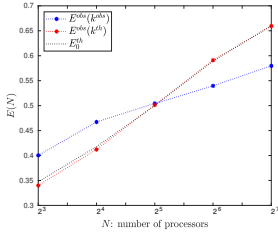
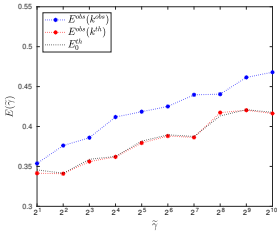


Figure: Comparison between k^{th} and k^{obs} , for $N = 16$ and $\delta t = \frac{\Delta T}{2^5}$. The eigenvalues of $A - LC$ are $\{-0.8, -1\}$ (top) and $\{-0.2, -0.25\}$ (bottom).



(a) $E(\tilde{\gamma})$, for $N = 16$ and $\delta t = \frac{\Delta T}{2^5}$.

(b) $E(N)$, for $\delta t = \frac{\Delta T}{2^5}$ and $\tilde{\gamma} = 2^{10}$.

(c) $E(\delta t)$, for $N = 16$ and $\tilde{\gamma} = 2^{10}$.

Figure: Comparison between $E^{obs}(k^{obs})$, $E^{obs}(k^{th})$ and E_0^{th} . The eigenvalues of $A - LC$ are $\{-0.8, -1\}$ (top) and $\{-0.2, -0.25\}$ (bottom).

- ▶ Use of other time-parallelization algorithms (e.g. ParaExp).
- ▶ Extension to a stochastic framework (continuous Kalman filter).
- ▶ Considering a *variable* window approach.

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Mathematical analysis of the BEM theory

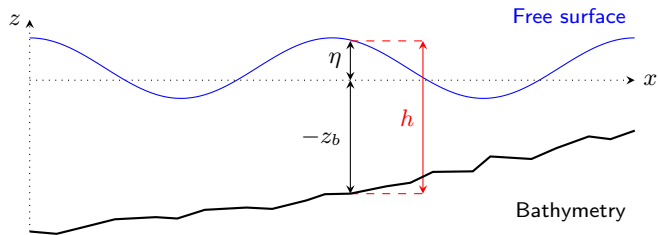
- Glauert's modeling

- Simplified and corrected models

- Solving algorithms

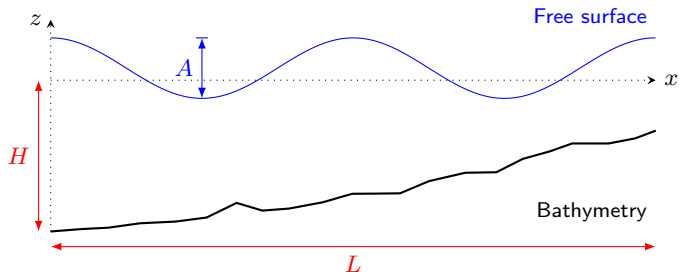
- Optimization

DERIVATION OF THE WAVE MODEL



► $\Omega_t = \{(x, z) \in \Omega \times \mathbb{R} \mid -z_b(x) \leq z \leq \eta(x, t)\}, \quad t \geq 0.$

DERIVATION OF THE WAVE MODEL



- ▶ $\Omega_t = \{(x, z) \in \Omega \times \mathbb{R} \mid -z_b(x) \leq z \leq \eta(x, t)\}, \quad t \geq 0.$
- ▶ Asymptotic derivation:

$$\varepsilon := \frac{H}{L}, \quad \delta := \frac{A}{H}.$$

$$(3) \quad \left\{ \begin{array}{ll} \frac{\partial \mathbf{u}}{\partial t} + (\mathbf{u} \cdot \nabla) \mathbf{u} = \operatorname{div}(\sigma_T) + \mathbf{g} & \text{in } \Omega_t, \\ \operatorname{div}(\mathbf{u}) = 0 & \text{in } \Omega_t, \\ \mathbf{u} = \mathbf{u}_0 & \text{in } \Omega_0. \end{array} \right.$$

- ▶ Incompressible fluid,
- ▶ $\mathbf{u} = (u, w)^\top$ denotes its velocity,
- ▶ $\sigma_T = -p\mathbb{I} + \mu(\nabla \mathbf{u} + \nabla \mathbf{u}^\top)$ is the total stress tensor, p denotes the pressure,
- ▶ gravity $\mathbf{g} = (0, -g)^\top$, atmospheric pressure p_0 , viscosity μ and density are constants.

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Change of variables

$$x' = \frac{x}{L}, \quad z' = \frac{z}{H}, \quad t' = \frac{C_0}{L} t,$$

and

$$u' = \frac{u}{\delta C_0}, \quad w' = \frac{w}{\delta \varepsilon C_0}, \quad \eta' = \frac{\eta}{A}, \quad z'_b = \frac{z_b}{H},$$

where $C_0 = \sqrt{gH}$. The dimensionless coefficients are given by

$$\mu' = \frac{\mu}{C_0 L}, \quad p' = \frac{p}{gH}, \quad p'_a = \frac{p_a}{gH}.$$

DEPTH-AVERAGED MASS EQUATION

Due to the **Leibnitz integral rule** and the boundary conditions, integrating the mass equation gives

$$\int_{-z_b}^{\delta\eta} \left(\frac{\partial u}{\partial x} + \frac{\partial w}{\partial z} \right) dz = 0$$
$$\frac{\partial}{\partial x} \left(\int_{-z_b}^{\delta\eta} u dz \right) - \delta u(x, \delta\eta, t) \frac{\partial \eta}{\partial x} - u(x, -z_b, t) \frac{\partial z_b}{\partial x} + w(x, \delta\eta, t) - w(x, -z_b, t) = 0$$

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$$(4) \quad \begin{cases} -\delta u \frac{\partial \eta}{\partial x} + w = \frac{\partial \eta}{\partial t} \sqrt{1 + (\varepsilon \delta)^2 \left| \frac{\partial \eta}{\partial x} \right|^2} & \text{on } (x, \delta\eta(x, t), t), \\ u \frac{\partial z_b}{\partial x} + w = 0 & \text{on } (x, -z_b(x), t). \end{cases}$$

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$$\frac{\partial(h_\delta \bar{u})}{\partial x} + \frac{\partial \eta}{\partial t} \sqrt{1 + (\varepsilon \delta)^2 \left| \frac{\partial \eta}{\partial x} \right|^2} = 0$$

We denote the depth-averaged velocity by

$$\bar{u}(x, t) = \frac{1}{h_\delta(x, t)} \int_{-z_b}^{\delta\eta} u(x, z, t) dz,$$

where $h_\delta = \delta\eta + z_b$.

The Momentum Equation (in w) yields

$$(4) \quad \varepsilon^2 \delta \left(\frac{\partial w}{\partial t} + \delta \left(u \frac{\partial w}{\partial x} + w \frac{\partial w}{\partial z} \right) \right) = -\frac{\partial p}{\partial z} - 1 + \delta \frac{\partial}{\partial x} \left(\mu \left(\frac{\partial u}{\partial z} + \varepsilon^2 \frac{\partial w}{\partial x} \right) \right) \\ + 2\delta \frac{\partial}{\partial z} \left(\mu \frac{\partial w}{\partial z} \right).$$

We assume a small viscosity coefficient $\mu = \varepsilon\mu_0$.

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After **rearranging terms** of order ε^2 and **integrating in z** , we get

$$p(x, z, t) = \mathcal{O}(\varepsilon^2 \delta) + (\delta\eta - z) + \varepsilon\delta\mu_0 \left(\frac{\partial u}{\partial x} + 2 \frac{\partial w}{\partial z} - \frac{\partial u}{\partial x}(x, \delta\eta, t) \right) + p(x, \delta\eta, t) - 2\varepsilon\delta\mu_0 \frac{\partial w}{\partial z}(x, \delta\eta, t)$$

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DEPTH-AVERAGED MOMENTUM EQUATION

The Momentum Equation (in u) yields

$$\frac{\partial u}{\partial t} + \delta \left(u \frac{\partial u}{\partial x} + w \frac{\partial u}{\partial z} \right) = -\frac{1}{\delta} \frac{\partial p}{\partial x} + 2 \frac{\partial}{\partial x} \left(\varepsilon \mu_0 \frac{\partial u}{\partial x} \right) + \frac{\partial}{\partial z} \left(\varepsilon \mu_0 \left(\frac{1}{\varepsilon^2} \frac{\partial u}{\partial z} + \frac{\partial w}{\partial x} \right) \right).$$

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Since $\text{div}(\mathbf{u}) = 0$, integrating its left-hand side gives

$$\int_{-z_b}^{\delta\eta} \left[\frac{\partial u}{\partial t} + \delta \left(u \frac{\partial u}{\partial x} + w \frac{\partial u}{\partial z} \right) \right] dz = \int_{-z_b}^{\delta\eta} \frac{\partial u}{\partial t} dz + \delta \int_{-z_b}^{\delta\eta} \left(\frac{\partial u^2}{\partial x} + \frac{\partial uw}{\partial z} \right) dz$$

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PROPOSITION

The hydrostatic pressure, combined with the stress boundary conditions, implies that $u(x, z, t) = \bar{u}(x, t) + \mathcal{O}(\varepsilon)$. In particular, we have the approximation

$$h_\delta \overline{u^2} = h_\delta \bar{u}^2 + \mathcal{O}(\varepsilon^2).$$

DEPTH-AVERAGED MOMENTUM EQUATION

To treat the right-hand side, we use the **hydrostatic pressure**

$$\begin{aligned} \int_{-z_b}^{\delta\eta} \left[-\frac{1}{\delta} \frac{\partial p}{\partial x} + \varepsilon \mu_0 \left(2 \frac{\partial}{\partial x} \left(\frac{\partial u}{\partial x} \right) + \frac{\partial}{\partial z} \left(\frac{\partial w}{\partial x} \right) \right) + \frac{\mu_0}{\varepsilon} \frac{\partial}{\partial z} \left(\frac{\partial u}{\partial z} \right) \right] dz \\ = -h_\delta \frac{\partial \eta}{\partial x} + \mathcal{O}(\varepsilon) + \left[\frac{\mu_0}{\varepsilon} \frac{\partial u}{\partial z}(x, \delta\eta, t) - \frac{\mu_0}{\varepsilon} \frac{\partial u}{\partial z}(x, -z_b, t) \right]. \end{aligned}$$

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In summary, we have

$$\begin{aligned} \frac{\partial(h_\delta \bar{u})}{\partial t} + \delta \frac{\partial(h_\delta \bar{u}^2)}{\partial x} &= -h_\delta \frac{\partial \eta}{\partial x} + \mathcal{O}(\varepsilon) \\ &+ \left[\frac{\mu_0}{\varepsilon} \frac{\partial u}{\partial z}(x, \delta\eta, t) - \frac{\mu_0}{\varepsilon} \frac{\partial u}{\partial z}(x, -z_b, t) \right] \\ &+ \delta u(x, \delta\eta, t) \frac{\partial \eta}{\partial t} \left(\sqrt{1 + (\varepsilon\delta)^2 \left| \frac{\partial \eta}{\partial x} \right|^2} - 1 \right). \end{aligned}$$

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In summary, we have


$$\begin{aligned} \frac{\partial(h_\delta \bar{u})}{\partial x} + \frac{\partial \eta}{\partial t} \sqrt{1 + (\varepsilon \delta)^2 \left| \frac{\partial \eta}{\partial x} \right|^2} &= 0, \\ \frac{\partial(h_\delta \bar{u})}{\partial t} + \delta \frac{\partial(h_\delta \bar{u}^2)}{\partial x} &= -h_\delta \frac{\partial \eta}{\partial x} + \mathcal{O}(\varepsilon) \\ &\quad + \left[\frac{\mu_0}{\varepsilon} \frac{\partial u}{\partial z}(x, \delta\eta, t) - \frac{\mu_0}{\varepsilon} \frac{\partial u}{\partial z}(x, -z_b, t) \right] \\ &\quad + \delta u(x, \delta\eta, t) \frac{\partial \eta}{\partial t} \left(\sqrt{1 + (\varepsilon \delta)^2 \left| \frac{\partial \eta}{\partial x} \right|^2} - 1 \right). \end{aligned}$$

Navier-Stokes


$$\left\{ \begin{array}{ll} \frac{\partial \mathbf{u}}{\partial t} + (\mathbf{u} \cdot \nabla) \mathbf{u} = \operatorname{div}(\sigma_T) + \mathbf{g} & \text{in } \Omega_t, \\ \operatorname{div}(\mathbf{u}) = 0 & \text{in } \Omega_t, \\ \mathbf{u} = \mathbf{u}_0 & \text{in } \Omega_0. \end{array} \right.$$

Saint-Venant

$$\left\{ \begin{array}{l} \frac{\partial h\bar{u}}{\partial x} + \frac{\partial \eta}{\partial t} = 0, \\ \frac{\partial h\bar{u}}{\partial t} + \frac{\partial h\bar{u}^2}{\partial x} + gh \frac{\partial \eta}{\partial x} = 0. \end{array} \right.$$


$$\varepsilon = \frac{H}{L} \longrightarrow 0$$

(Long wave theory)



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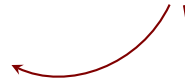
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Wave Equation

$$\frac{\partial^2 \eta}{\partial t^2} - \frac{\partial}{\partial x} \left(gz_b \frac{\partial \eta}{\partial x} \right) = 0.$$


$$\varepsilon = \frac{H}{L} \longrightarrow 0$$

(Long wave theory)


$$\delta = \frac{A}{H} \longrightarrow 0$$

(Small amplitude theory)



In a two-dimensional setting, $\eta(x, t) = \text{Re}\{\psi_{tot}(x)e^{-i\omega t}\}$ is a solution of the Wave Equation, where the amplitude ψ_{tot} satisfies

$$(5) \quad \omega^2 \psi_{tot} + \text{div}(gz_b \nabla \psi_{tot}) = 0.$$

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The bathymetry can be decomposed as $z_b(x) := z_0 + \delta z_b(x)$, with z_0 constant and δz_b has compact support in Ω .

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We then reformulate (5) as

Total wave

$$\begin{cases} \text{div}((1+q)\nabla \psi_{tot}) + k_0^2 \psi_{tot} = 0 & \text{in } \Omega, \\ \nabla(\psi_{tot} - \psi_0) \cdot \hat{n} - ik_0(\psi_{tot} - \psi_0) = 0 & \text{on } \partial\Omega, \end{cases}$$

where $q(x) := \frac{\delta z_b(x)}{z_0}$ is compactly supported in Ω , $k_0 := \frac{\omega}{\sqrt{gz_0}}$, \hat{n} is the unit normal to $\partial\Omega$ and $\psi_0(x) = e^{ik_0 x \cdot \vec{d}}$ (s.t. $|\vec{d}| = 1$).

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Scattered wave ($\psi_{tot} = \psi_0 + \psi_{sc}$)

$$\begin{cases} \text{div}((1+q)\nabla \psi_{sc}) + k_0^2 \psi_{sc} = -\text{div}(q\nabla \psi_0) & \text{in } \Omega, \\ \nabla \psi_{sc} \cdot \hat{n} - ik_0 \psi_{sc} = 0 & \text{on } \partial\Omega. \end{cases}$$

where $q(x) := \frac{\delta z_b(x)}{z_0}$ is compactly supported in Ω , $k_0 := \frac{\omega}{\sqrt{gz_0}}$, \hat{n} is the unit normal to $\partial\Omega$ and $\psi_0(x) = e^{ik_0 x \cdot \vec{d}}$ (s.t. $|\vec{d}| = 1$).

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The bathymetry can be decomposed as $z_b(x) := z_0 + \delta z_b(x)$, with z_0 constant and δz_b has compact support in Ω .

We then consider the following problem

Helmholtz formulation

$$(6) \quad \begin{cases} -\operatorname{div}((1+q)\nabla \psi) - k_0^2 \psi = \operatorname{div}(q\nabla \psi_0) & \text{in } \Omega, \\ (1+q)\nabla \psi \cdot \hat{n} - ik_0 \psi = g - q\nabla \psi_0 \cdot \hat{n} & \text{on } \partial\Omega, \end{cases}$$

where $\Omega \subset \mathbb{R}^2$ is a bounded open set with Lipschitz boundary, $q \in L^\infty(\Omega)$ satisfying

$$(7) \quad (\exists \alpha > 0) \quad 1 + q(x) \geq \alpha \quad \text{a.e. } x \in \Omega.$$

A weak formulation for (6) is given by

$$(8) \quad a(q; \psi, \phi) = b(q; \phi) \quad \forall \phi \in H^1(\Omega),$$

where

$$\begin{aligned} a(q; \psi, \phi) &:= \int_{\Omega} \left((1+q) \nabla \psi \cdot \nabla \bar{\phi} - k_0^2 \psi \bar{\phi} \right) dx - ik_0 \int_{\partial\Omega} \psi \bar{\phi} d\sigma, \\ b(q; \phi) &:= - \int_{\Omega} q \nabla \psi_0 \cdot \nabla \bar{\phi} dx + \langle g, \bar{\phi} \rangle_{H^{-1/2}, H^{1/2}}. \end{aligned}$$

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$$b(q; \phi) := - \int_{\Omega} q \nabla \psi_0 \cdot \nabla \bar{\phi} dx + \langle g, \bar{\phi} \rangle_{H^{-1/2}, H^{1/2}}.$$

The sesquilinear form a :

- is continuous under the norm

$$\|\psi\|_{1,k_0}^2 := k_0^2 \|\psi\|_{L^2(\Omega)}^2 + \alpha \|\nabla \psi\|_{L^2(\Omega)}^2.$$

- Satisfies a Gårding inequality

$$\operatorname{Re}\{a(q; \psi, \psi)\} + 2k_0^2 \|\psi\|_{L^2(\Omega)}^2 \geq \|\psi\|_{1,k_0}^2.$$

We are interested in solving the next PDE-constrained optimization problem

$$(9) \quad \begin{aligned} & \min_{(q, \psi) \in U_{\Lambda} \times H^1(\Omega)} J(q, \psi) \\ & \text{s.t.} \quad (8). \end{aligned}$$

where $U_{\Lambda} = \{q \in BV(\Omega) \mid \alpha - 1 \leq q(x) \leq \Lambda \text{ a.e. } x \in \Omega\}$ is a closed, weakly* closed and convex subset of $BV(\Omega)$.

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Space of functions of Bounded Variations

- ▶ Banach space for the norm $\|q\|_{BV(\Omega)} := \|q\|_{L^1(\Omega)} + |Dq|(\Omega)$, where D is the distributional gradient and $|Dq|(\Omega)$ is the variation of q .
- ▶ The weak* convergence means

$$q_n \rightarrow q \text{ in } L^1(\Omega) \text{ and } Dq_n \rightharpoonup Dq \text{ in } \mathcal{M}_b(\Omega, \mathbb{R}^N).$$

- ▶ The application $q \in BV(\Omega) \mapsto |Dq|(\Omega) \in \mathbb{R}^+$ is lower semi-continuous with respect to the weak* topology of BV .

THEOREM

Assume that $q \in U_\Lambda$. Then there exists a constant $C_s(k_0, \Omega) > 0$ such that

$$\|\psi\|_{1,k_0} \leq C_s(k_0, \Omega) \sup_{\|\phi\|_{1,k_0}=1} |a(q; \psi, \phi)|.$$

In addition, the solution to (8) satisfies the bound

$$\begin{aligned} \|\psi\|_{1,k_0} &\leq C_s(k_0, \Omega) C(\Omega) \max\{k_0^{-1}, \alpha^{-1/2}\} \\ &\quad \times \left(\|q\|_{L^\infty(\Omega)} \|\nabla \psi_0\|_{L^2(\Omega)} + \|g\|_{H^{-1/2}(\partial\Omega)} \right), \end{aligned}$$

with $C(\Omega) > 0$.

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with $C(\Omega) > 0$.

As a result of this theorem and the continuity of the trace, we have

$$\begin{aligned} \|\psi_{tot}\|_{1,k_0} &\leq C(\Omega) C_s(k_0, \Omega) k_0 \max\{k_0^{-1}, \alpha^{-1/2}\}, \\ \|\psi_{sc}\|_{1,k_0} &\leq k_0 C_s(k_0, \Omega) \alpha^{-1/2} \|q\|_{L^\infty(\Omega)} \sqrt{|\Omega|}. \end{aligned}$$

THEOREM

Let $(q_n)_n \subset U$ be a sequence satisfying $\|q_n\|_{BV(\Omega)} \leq M$ and whose weak* limit in $BV(\Omega)$ is denoted by q_∞ . Let $(\psi(q_n))_n$ be the sequence of weak solution to Problem (8). Then $\psi(q_n)$ converges strongly in $H^1(\Omega)$ towards $\psi(q_\infty)$. In other words, the mapping

$$q \in (U_\Lambda, \text{weak}^*) \mapsto \psi(q) \in (H^1(\Omega), \text{strong}),$$

is continuous.

THEOREM ► EXISTENCE OF OPTIMAL SOLUTION [COCQUET, RIFFO AND SALOMON]

Assume that the cost function $(q, \psi) \in U_\Lambda \mapsto J(q, \psi) \in \mathbb{R}$ satisfies:

(A1) There exists $\beta > 0$ such that

$$J(q, \psi) = J_0(q, \psi) + \beta |Dq|(\Omega).$$

(A2) $\forall (q, \psi) \in U_\Lambda \times H^1(\Omega)$, $J_0(q, \psi) \geq m > -\infty$.

(A3) $(q, \psi) \mapsto J_0(q, \psi)$ is lower-semi-continuous with respect to the (weak*, weak) topology of $BV(\Omega) \times H^1(\Omega)$.

Then the optimization problem (9) has at least one optimal solution in $U_\Lambda \times H^1(\Omega)$.

THEOREM

Assume that $q \in L^\infty(\Omega)$ and satisfies (7) and $g \in L^2(\partial\Omega)$. Then the solution to Problem (8) satisfies

$$\|\psi\|_{C^0(\Omega)} \leq \tilde{C}(\Omega)\tilde{C}_s(k_0, \alpha) \left(\|q\|_{L^\infty(\Omega)} \|\nabla\psi_0\|_{L^\infty(\Omega)} + \|g\|_{L^2(\partial\Omega)} \right),$$

where $\tilde{C}(\Omega) > 0$ and

$$\tilde{C}_s(k_0, \alpha) = 1 + \left((1 + k_0^2)k_0^{-1} + \alpha^{-1/2} \right) \max\{k_0^{-1}, \alpha^{-1/2}\} C_s(k_0, \Omega).$$

THEOREM

Assume that $q \in L^\infty(\Omega)$ and satisfies (7) and $g \in L^2(\partial\Omega)$. Then the solution to Problem (8) satisfies

$$\|\psi\|_{C^0(\Omega)} \leq \tilde{C}(\Omega) \tilde{C}_s(k_0, \alpha) \left(\|q\|_{L^\infty(\Omega)} \|\nabla \psi_0\|_{L^\infty(\Omega)} + \|g\|_{L^2(\partial\Omega)} \right),$$

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Consequently,

$$\|\psi_{tot}\|_{C^0(\Omega)} \leq k_0 \tilde{C}(\Omega) \tilde{C}_s(k_0, \alpha),$$

$$\|\psi_{sc}\|_{C^0(\Omega)} \leq k_0 \tilde{C}(\Omega) \left[\left((1 + k_0^2) k_0^{-1} + \alpha^{-1/2} \right) \alpha^{-1/2} C_s(k_0, \Omega) + 1 \right] \|q\|_{L^\infty(\Omega)}.$$

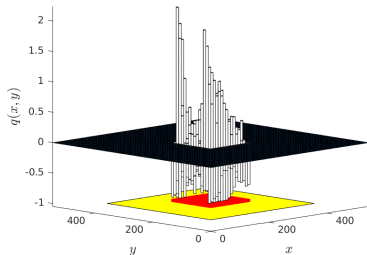
Minimization of the cost functional

$$J(q, \psi_{tot}) = \frac{\omega_0^2}{2} \int_{\Omega_0} |\psi_{tot}(x, y)|^2 dx dy,$$
$$\Omega_0 = [\frac{L}{6}, \frac{5L}{6}]^2$$

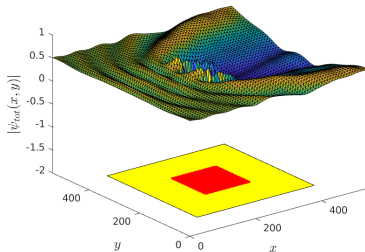
The bathymetry is only optimized on a subset $\Omega_q = [\frac{L}{4}, \frac{3L}{4}]^2 \subset \Omega_0$.

OPTIMAL BATHYMETRY FOR A WAVE DAMPING PROBLEM

Optimal topography



(a) View from above.



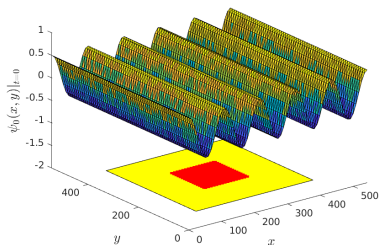
(b) Norm of the numerical solution.

Minimization of the cost functional

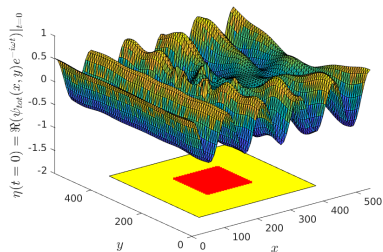
$$J(q, \psi_{tot}) = \frac{\omega_0^2}{2} \int_{\Omega_0} |\psi_{tot}(x, y)|^2 dx dy,$$
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OPTIMAL BATHYMETRY FOR A WAVE DAMPING PROBLEM



(a) Real part of the incident wave.



(b) Real part of the numerical solution.

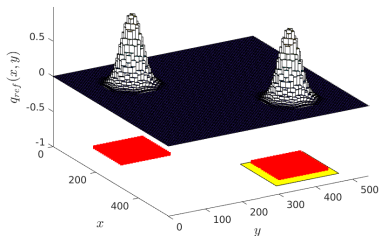
Minimization of the cost functional

$$J(q, \psi_{tot}) = \frac{\omega_0^2}{2} \int_{\Omega_0} |\psi_{tot}(x, y) - \psi_{ref}(x, y)|^2 dx dy,$$
$$\Omega_0 = [\frac{3L}{4} - \delta, \frac{3L}{4} + \delta]^2$$

where ψ_{ref} is the amplitude associated with q_{ref} , $\delta = \frac{L}{6}$ and $\Omega_q = [\frac{L}{8}, \frac{3L}{8}]^2 \cup [\frac{5L}{8}, \frac{7L}{8}]^2$.

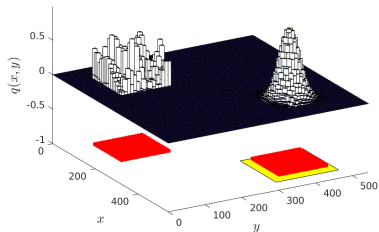
DETECTION OF A BATHYMETRY FROM A WAVEFIELD

Actual topography



(a) Actual bathymetry.

Reconstructed topography



(b) Reconstructed bathymetry.

Minimization of the cost functional

$$J(q, \psi_{tot}) = \frac{\omega_0^2}{2} \int_{\Omega_0} |\psi_{tot}(x, y) - \psi_{ref}(x, y)|^2 dx dy,$$

$$\Omega_0 = [\frac{3L}{4} - \delta, \frac{3L}{4} + \delta]^2$$

where ψ_{ref} is the amplitude associated with q_{ref} , $\delta = \frac{L}{6}$ and $\Omega_q = [\frac{L}{8}, \frac{3L}{8}]^2 \cup [\frac{5L}{8}, \frac{7L}{8}]^2$.

- ▶ A natural extension is to consider a polychromatic wave.
- ▶ Several questions have to be addressed in between, regarding first a possible decomposition of the cost functional and then the convergence of the whole procedure.
- ▶ This idea cannot be extended to nonlinear wave propagation models as Saint-Venant or Boussinesq.

Time-parallelization of sequential DA problems

- Luenberger observer

- Time-parallelization setting

- Parareal algorithm

- Diamond strategy (Parareal case)

Bathymetry optimization

- Derivation of the wave model

- Description of the optimization problem

- Continuous optimization problem

- Numerical examples

Mathematical analysis of the BEM theory

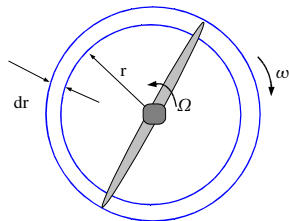
- Glauert's modeling

- Simplified and corrected models

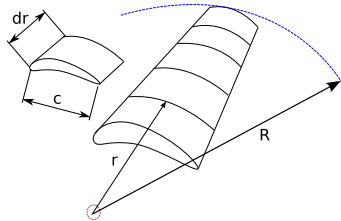
- Solving algorithms

- Optimization

GLAUERT'S MODELING (GLOBAL VARIABLES)



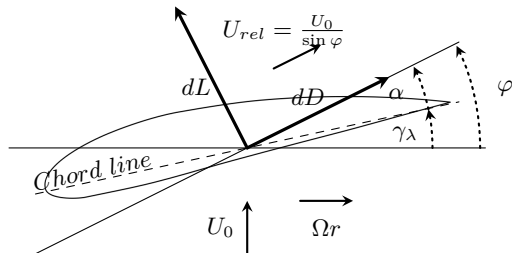
(a) Radial decomposition



(b) Blade element model

$$a = \frac{U_{-\infty} - U_0}{U_{-\infty}},$$
$$a' = \frac{\omega}{2\Omega},$$
$$\tan \varphi = \frac{1 - a}{\lambda(1 + a')}.$$

GLAUERT'S MODELING (LOCAL VARIABLES)

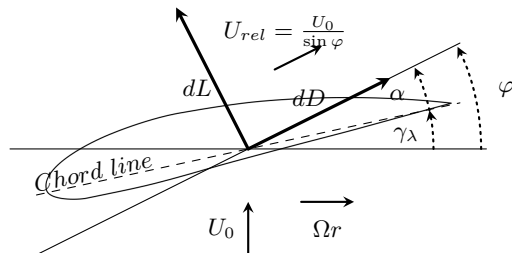


Blade element profile and associated angles, velocities and forces.

$$dL = C_L(\alpha) \frac{\rho}{2} U_{rel}^2 c_\lambda dr,$$

$$dD = C_D(\alpha) \frac{\rho}{2} U_{rel}^2 c_\lambda dr.$$

GLAUERT'S MODELING (LOCAL VARIABLES)



$$dL = C_L(\alpha) \frac{\rho}{2} U_{rel}^2 c_\lambda dr,$$
$$dD = C_D(\alpha) \frac{\rho}{2} U_{rel}^2 c_\lambda dr.$$

Blade element profile and associated angles, velocities and forces.

ASSUMPTION 3.1

In what follows, we assume that C_L is well-defined and continuous on an interval

$$I_\beta := [-\beta, \beta]$$

for some $\beta \in [0, \alpha_s)$ and positive on $I_\beta \cap \mathbb{R}^+$. The coefficient C_D is well-defined and positive on \mathbb{R} .

We denote by dT the infinitesimal thrust and dQ the infinitesimal torque that apply on the blade element under consideration.

Macroscopic approach (Momentum Theory)

$$\begin{aligned}dT &= 4a(1-a)U_{-\infty}^2\rho\pi r dr, \\dQ &= 4a'(1-a)\lambda U_{-\infty}^2\rho\pi r^2 dr.\end{aligned}$$

Local expressions (Blade Element Theory)

$$\begin{aligned}dT &= \sigma_\lambda \frac{(1-a)^2}{\sin^2 \varphi} (C_L(\varphi - \gamma_\lambda) \cos \varphi + C_D(\varphi - \gamma_\lambda) \sin \varphi) U_{-\infty}^2 \rho \pi r dr, \\dQ &= \sigma_\lambda \frac{(1-a)^2}{\sin^2 \varphi} (C_L(\varphi - \gamma_\lambda) \sin \varphi - C_D(\varphi - \gamma_\lambda) \cos \varphi) U_{-\infty}^2 \rho \pi r^2 dr,\end{aligned}$$

$$\text{with } \sigma_\lambda = \frac{Bc_\lambda}{2\pi r}.$$

Glauert's relations

$$\tan \varphi = \frac{1 - a}{\lambda(1 + a')},$$

$$\frac{a}{1 - a} = \frac{\sigma_\lambda}{4 \sin^2 \varphi} (C_L(\varphi - \gamma_\lambda) \cos \varphi + C_D(\varphi - \gamma_\lambda) \sin \varphi),$$

$$\frac{a'}{1 - a} = \frac{\sigma_\lambda}{4 \lambda \sin^2 \varphi} (C_L(\varphi - \gamma_\lambda) \sin \varphi - C_D(\varphi - \gamma_\lambda) \cos \varphi).$$

Glauert's relations become

$$(10) \quad \tan \varphi = \frac{1 - a}{\lambda(1 + a')},$$

$$(11) \quad \frac{a}{1 - a} = \mu_L(\varphi) \frac{\cos \varphi}{\sin^2 \varphi},$$

$$(12) \quad \frac{a'}{1 - a} = \mu_L(\varphi) \frac{1}{\lambda \sin \varphi},$$

with $\mu_L(\varphi) := \frac{\sigma_\lambda}{4} C_L(\varphi - \gamma_\lambda)$ defined on $I_{\beta, \gamma_\lambda} := [-\beta + \gamma_\lambda, \beta + \gamma_\lambda]$.

SIMPLIFIED MODEL ($C_D = 0$)

Glauert's relations become

$$(10) \quad \tan \varphi = \frac{1 - a}{\lambda(1 + a')},$$

$$(11) \quad \frac{a}{1 - a} = \mu_L(\varphi) \frac{\cos \varphi}{\sin^2 \varphi},$$

$$(12) \quad \frac{a'}{1 - a} = \mu_L(\varphi) \frac{1}{\lambda \sin \varphi},$$

with $\mu_L(\varphi) := \frac{\sigma_\lambda}{4} C_L(\varphi - \gamma_\lambda)$ defined on $I_{\beta, \gamma_\lambda} := [-\beta + \gamma_\lambda, \beta + \gamma_\lambda]$.

THEOREM 3.2 ► REFORMULATION OF THE SIMPLIFIED MODEL

Suppose that Assumption 3.1 holds and that $(\varphi, a, a') \in I - \{0, \frac{\pi}{2}\} \times \mathbb{R} - \{1\} \times \mathbb{R} - \{-1\}$ satisfies Eqs (10–12). Then φ satisfies

$$(13) \quad \mu_L(\varphi) = \mu_G(\varphi),$$

where $\mu_G(\varphi) := \sin \varphi \tan(\theta_\lambda - \varphi)$. Reciprocally, suppose that $\varphi \in I - \{0, \frac{\pi}{2}\}$ satisfies Eq. (13) and define a and a' as the corresponding solutions of Eqs. (11) and (12), respectively. Then $(\varphi, a, a') \in I - \{0, \frac{\pi}{2}\} \times \mathbb{R} - \{1\} \times \mathbb{R} - \{-1\}$ satisfies Eqs. (10–12).

$$(14) \quad \tan \varphi = \frac{1-a}{\lambda(1+a')},$$

$$(15) \quad \frac{a}{1-a} = \frac{1}{\sin^2 \varphi} (\mu_L^c(\varphi) \cos \varphi + \mu_D^c(\varphi) \sin \varphi) - \frac{\psi((a-a_c)_+)}{(1-a)^2},$$

$$(16) \quad \frac{a'}{1-a} = \frac{1}{\lambda \sin^2 \varphi} (\mu_L^c(\varphi) \sin \varphi - \mu_D^c(\varphi) \cos \varphi),$$

where $\mu_L^c(\varphi) := \frac{\sigma_\lambda}{4F_\lambda(\varphi)} C_L(\varphi - \gamma_\lambda)$, $\mu_D^c(\varphi) := \frac{\sigma_\lambda}{4F_\lambda(\varphi)} C_D(\varphi - \gamma_\lambda)$, defined respectively on $I_{\beta, \gamma_\lambda}$ and \mathbb{R} .

$$(14) \quad \tan \varphi = \frac{1-a}{\lambda(1+a')},$$

$$(15) \quad \frac{a}{1-a} = \frac{1}{\sin^2 \varphi} (\mu_L^c(\varphi) \cos \varphi + \mu_D^c(\varphi) \sin \varphi) - \frac{\psi((a-a_c)_+)}{(1-a)^2},$$

$$(16) \quad \frac{a'}{1-a} = \frac{1}{\lambda \sin^2 \varphi} (\mu_L^c(\varphi) \sin \varphi - \mu_D^c(\varphi) \cos \varphi),$$

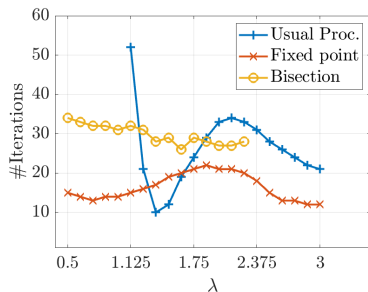
where $\mu_L^c(\varphi) := \frac{\sigma_\lambda}{4F_\lambda(\varphi)} C_L(\varphi - \gamma_\lambda)$, $\mu_D^c(\varphi) := \frac{\sigma_\lambda}{4F_\lambda(\varphi)} C_D(\varphi - \gamma_\lambda)$, defined respectively on $I_{\beta, \gamma_\lambda}$ and \mathbb{R} .

THEOREM 3.3 ► REFORMULATION OF THE CORRECTED MODEL

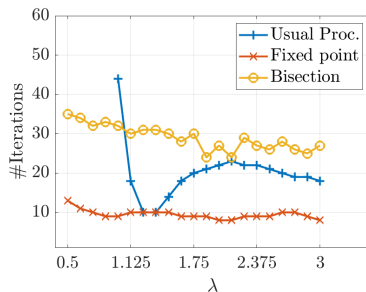
(...) Suppose also that $(\varphi, a, a') \in I^+ - \{\frac{\pi}{2}\} \times \mathbb{R} - \{1\} \times \mathbb{R}$ satisfies Equations (14–16). Then φ satisfies

$$(17) \quad \mu_L^c(\varphi) - \tan(\theta_\lambda - \varphi) \mu_D^c(\varphi) = \mu_G^c(\varphi),$$

where $\mu_G^c(\varphi) := \mu_G(\varphi) + \frac{\cos \theta_\lambda \sin^2 \varphi}{\cos(\theta_\lambda - \varphi)} \frac{\psi((\tau(\varphi) - a_c)_+)}{(1 - \tau(\varphi))^2}$. (...)



(a) $a_c = 0.2$



(b) $a_c = 1$

Number of iterations of usual (Usual Proc.), fixed-point (Fixed point) and bisection (Bisection) algorithms required to solve Equation (17), according to the criterium

$$\left| \mu_L^c(\varphi^k) - \tan(\theta_\lambda - \varphi^k) \mu_D^c(\varphi^k) - \mu_G^c(\varphi^k) \right| \leq \text{Tol} = 10^{-10}$$

and for various values of λ .

The design procedure mainly consists in optimizing the power coefficient

$$\begin{aligned} \max C_p(\gamma_\lambda, c_\lambda, \varphi) &= \frac{8}{\lambda_{\max}^2} \int_{\lambda_{\min}}^{\lambda_{\max}} \lambda^3 a'(1-a) \left(1 - \frac{C_D}{C_L} \tan^{-1} \varphi\right) d\lambda, \\ \text{s.t. } \begin{cases} \tan \varphi = \frac{1-a}{\lambda(1+a')}, \\ \frac{a}{1-a} = \frac{1}{\sin^2 \varphi} (\mu_L^c(\varphi) \cos \varphi + \mu_D^c(\varphi) \sin \varphi) - \frac{\psi((a-a_c)_+)}{(1-a)^2}, \\ \frac{a'}{1-a} = \frac{1}{\lambda \sin^2 \varphi} (\mu_L^c(\varphi) \sin \varphi - \mu_D^c(\varphi) \cos \varphi). \end{cases} \end{aligned}$$

The design procedure mainly consists in optimizing the power coefficient

$$\begin{aligned} \max C_p(\gamma_\lambda, c_\lambda, \varphi) &= \frac{8}{\lambda_{\max}^2} \int_{\lambda_{\min}}^{\lambda_{\max}} \lambda^3 a' (1-a) \left(1 - \frac{C_D}{C_L} \tan^{-1} \varphi\right) d\lambda, \\ \text{s.t. } \begin{cases} \tan \varphi = \frac{1-a}{\lambda(1+a')}, \\ \frac{a}{1-a} = \frac{1}{\sin^2 \varphi} (\mu_L(\varphi) \cos \varphi + \mu_D^c(\varphi) \sin \varphi), \\ \frac{a'}{1-a} = \frac{1}{\lambda \sin^2 \varphi} (\mu_L(\varphi) \sin \varphi - \mu_D^c(\varphi) \cos \varphi). \end{cases} \end{aligned}$$

Several simplifications are taken into account:

- Assume $F_\lambda(\varphi) = 1$ and $\psi((a - a_c)_+) = 0$.

The design procedure mainly consists in optimizing the power coefficient

$$\begin{aligned} \max C_p(\gamma_\lambda, c_\lambda, \varphi) &= \frac{8}{\lambda_{\max}^2} \int_{\lambda_{\min}}^{\lambda_{\max}} \lambda^3 a' (1-a) \left(1 - \frac{C_D}{C_L} \tan^{-1} \varphi\right) d\lambda, \\ \text{s.t. } \begin{cases} \tan \varphi = \frac{1-a}{\lambda(1+a')}, \\ \frac{a}{1-a} = \frac{1}{\sin^2 \varphi} (\mu_L(\varphi) \cos \varphi + \mu_D^c(\varphi) \sin \varphi), \\ \frac{a'}{1-a} = \frac{1}{\lambda \sin^2 \varphi} (\mu_L(\varphi) \sin \varphi - \mu_D^c(\varphi) \cos \varphi). \end{cases} \end{aligned}$$

Several simplifications are taken into account:

- ▶ Assume $F_\lambda(\varphi) = 1$ and $\psi((a - a_c)_+) = 0$.
- ▶ Define $\bar{\alpha} = \varphi - \gamma_\lambda$ that minimizes $\frac{C_D}{C_L}$.

The design procedure mainly consists in optimizing the power coefficient

$$\begin{aligned} \max C_p(\gamma_\lambda, c_\lambda, \varphi) &= \frac{8}{\lambda_{\max}^2} \int_{\lambda_{\min}}^{\lambda_{\max}} \lambda^3 a' (1 - a) d\lambda, \\ \text{s.t. } \begin{cases} \tan \varphi = \frac{1 - a}{\lambda(1 + a')}, \\ \frac{a}{1 - a} = \frac{1}{\sin^2 \varphi} \mu_L(\varphi) \cos \varphi, \\ \frac{a'}{1 - a} = \frac{1}{\lambda \sin^2 \varphi} \mu_L(\varphi) \sin \varphi. \end{cases} \end{aligned}$$

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- ▶ Assume $F_\lambda(\varphi) = 1$ and $\psi((a - a_c)_+) = 0$.
- ▶ Define $\bar{\alpha} = \varphi - \gamma_\lambda$ that minimizes $\frac{C_D}{C_L}$.
- ▶ Then the coefficient C_D is simply neglected.

The design procedure mainly consists in optimizing the power coefficient

$$\begin{aligned} \max C_p(\gamma_\lambda, \varphi) &= \frac{8}{\lambda_{\max}^2} \int_{\lambda_{\min}}^{\lambda_{\max}} \lambda^3 a' (1 - a) d\lambda, \\ \text{s.t. } \begin{cases} \tan \varphi = \frac{1 - a}{\lambda(1 + a')}, \\ \frac{a}{1 - a} = \frac{1}{\sin^2 \varphi} \mu_L(\varphi) \cos \varphi, \\ \frac{a'}{1 - a} = \frac{1}{\lambda \sin^2 \varphi} \mu_L(\varphi) \sin \varphi. \end{cases} \end{aligned}$$

Several simplifications are taken into account:

- ▶ Assume $F_\lambda(\varphi) = 1$ and $\psi((a - a_c)_+) = 0$.
- ▶ Define $\bar{\alpha} = \varphi - \gamma_\lambda$ that minimizes $\frac{C_D}{C_L}$.
- ▶ Then the coefficient C_D is simply neglected.
- ▶ Theorem 3.2 allows us to replace $\mu_L(\varphi) := \frac{\sigma_\lambda}{4} C_L(\varphi - \gamma_\lambda)$ by $\mu_G(\varphi)$.

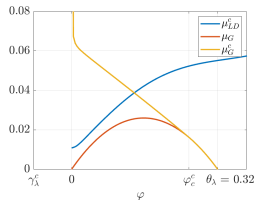
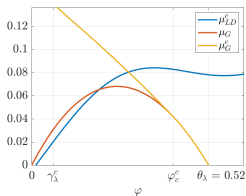
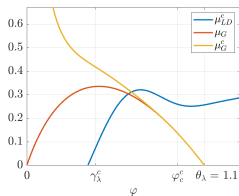
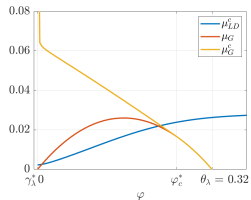
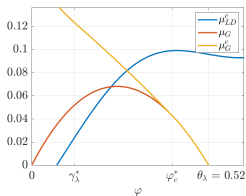
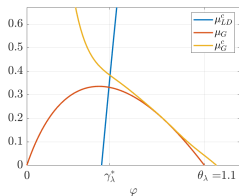
The design procedure mainly consists in optimizing the power coefficient

$$\begin{aligned} \max_{\varphi} \quad C_p(\varphi) &= \frac{8}{\lambda_{\max}} \int_{\lambda_{\min}}^{\lambda_{\max}} \lambda^3 a'(1-a) d\lambda \\ \text{s.t.} \quad &\begin{cases} \tan \varphi = \frac{1-a}{\lambda(1+a')} \\ \frac{a}{1-a} = \mu_G(\varphi) \frac{\cos \varphi}{\sin^2 \varphi} \\ \frac{a'}{1-a} = \mu_G(\varphi) \frac{1}{\lambda \sin \varphi}. \end{cases} \end{aligned}$$

Several simplifications are taken into account:

- ▶ Assume $F_\lambda(\varphi) = 1$ and $\psi((a - a_c)_+) = 0$.
- ▶ Define $\bar{\alpha} = \varphi - \gamma_\lambda$ that minimizes $\frac{C_D}{C_L}$.
- ▶ Then the coefficient C_D is simply neglected.
- ▶ Theorem 3.2 allows us to replace $\mu_L(\varphi) := \frac{\sigma_\lambda}{4} C_L(\varphi - \gamma_\lambda)$ by $\mu_G(\varphi)$.
- ▶ The solution of the new problem is $\gamma_\lambda^* = \varphi^* - \bar{\alpha}$, $c_\lambda^* = \frac{8\pi r \mu_G(\varphi^*)}{BC_L(\bar{\alpha})}$.

NUMERICAL EXPERIMENTS



(a) $\lambda_1 = 0.5$

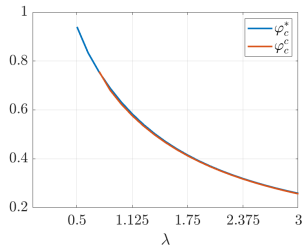
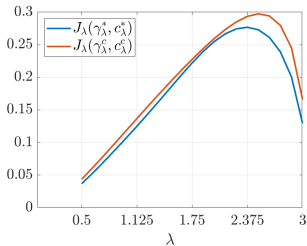
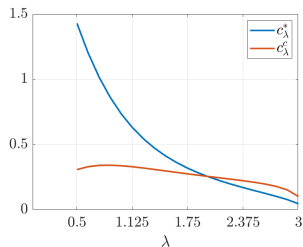
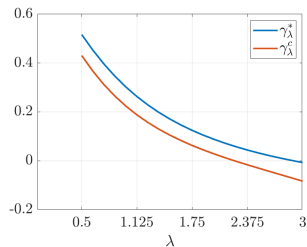
(b) $\lambda_2 = 1.75$

(c) $\lambda_3 = 3$

Graphs of the functions $\mu_{LD}^c : \varphi \mapsto \mu_L^c(\varphi) - \tan(\theta_\lambda - \varphi)\mu_D^c(\varphi)$, μ_G^c and μ_G for various values of λ .

Above: $(\gamma_\lambda, c_\lambda) = (\gamma_\lambda^*, c_\lambda^*)$, below: $(\gamma_\lambda, c_\lambda) = (\gamma_\lambda^c, c_\lambda^c)$.

NUMERICAL EXPERIMENTS



Graphs of the functions γ_λ^* and γ_λ^c , c_λ^* and c_λ^c , $J_\lambda(\gamma_\lambda^*, c_\lambda^*)$ and $J_\lambda(\gamma_\lambda^c, c_\lambda^c)$ and the corresponding φ_c .

- ▶ Concerning the convergence of solving algorithms, extending the proof to a more general framework is desirable.
- ▶ An asymptotic analysis gives a general idea of the optimal solution behavior, however the required assumptions seem very restrictive.
- ▶ The question of multiple optima in the corrected model remains open.

Thank you for your attention !