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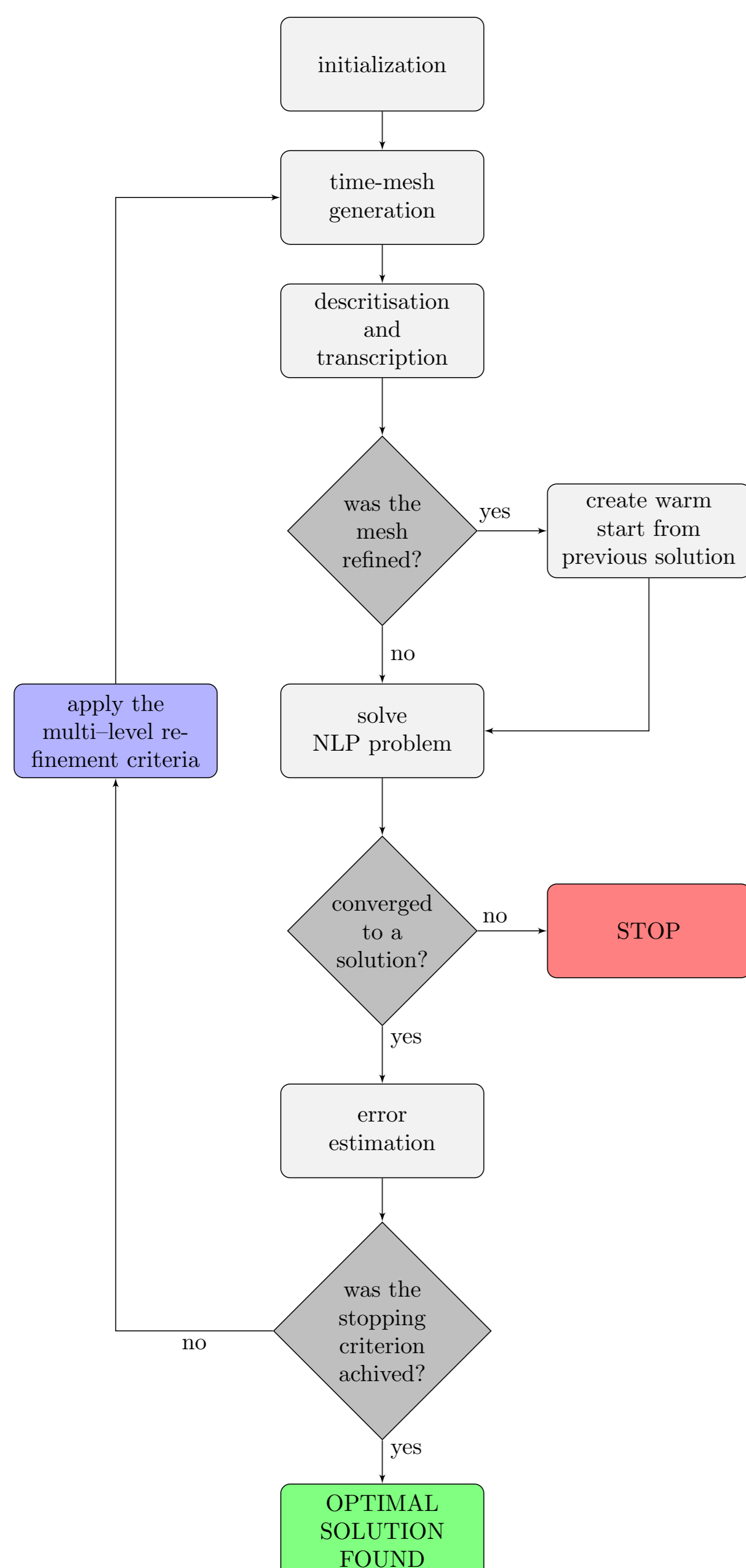


Time–mesh refinement

Continuous time optimal control problem, in Bolza form, with input and state constraints:

$$\begin{aligned} & \text{Minimize } \int_{t_0}^{t_f} L(t, \mathbf{x}(t), \mathbf{u}(t)) dt + G(t_0, \mathbf{x}(t_0), t_f, \mathbf{x}(t_f)) \\ & \text{subject to } \dot{\mathbf{x}}(t) = \mathbf{f}(t, \mathbf{x}(t), \mathbf{u}(t)) \quad \text{a.e. } t \in [t_0, t_f] \\ & \mathbf{u} \in \mathbf{U}(t) \subset \mathbb{R}^m \quad \text{a.e. } t \in [t_0, t_f] \\ & \mathbf{x}(t) \in \mathbb{X} \quad \forall t \in [t_0, t_f] \\ & (\mathbf{x}(t_0), \mathbf{x}(t_f)) \in \mathbb{X}_{01} \end{aligned}$$

Algorithm



Refinement strategy

The original coarse mesh is divided in K mesh intervals

$$\mathcal{S}_k = [\tau_{k-1}, \tau_k], \quad k = 1, \dots, K-1 \quad \text{and} \quad \mathcal{S}_K = [\tau_{K-1}, \tau_K]$$

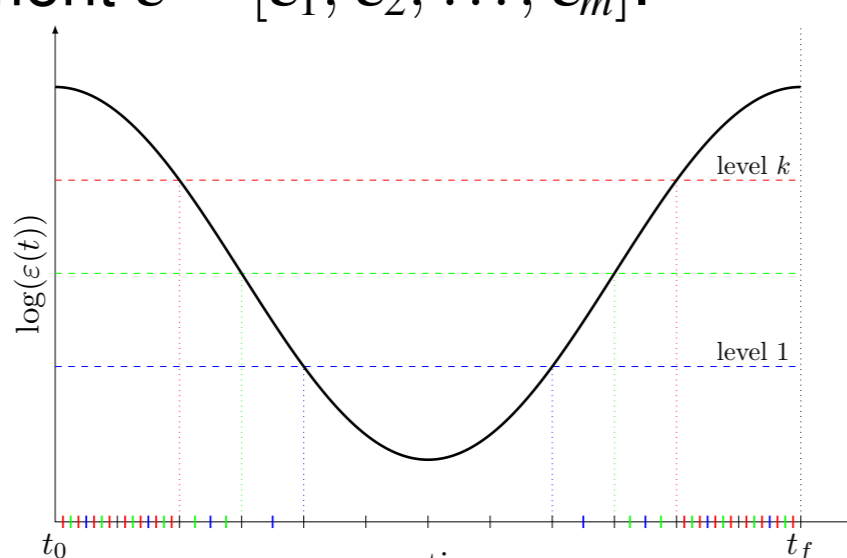
where $\tau_0 < \tau_1 < \dots < \tau_K$ coincide with nodes.

The mesh intervals \mathcal{S}_k form a partition of the time interval,

$$\bigcup_{k=1}^K \mathcal{S}_k = [t_0, t_f] \quad \text{and} \quad \bigcap_{k=1}^K \mathcal{S}_k = \emptyset.$$

After selecting the intervals \mathcal{S}_k that verify the refinement criteria, they are divided into smaller subintervals according to the user–defined levels of refinement $\bar{\varepsilon} = [\varepsilon_1, \varepsilon_2, \dots, \varepsilon_m]$.

A subinterval $\mathcal{S}_{k,i}$ is at the i^{th} level of refinement if $\mathcal{S}_{k,i} \in [\varepsilon_i, \varepsilon_{i+1}]$, for $i = 1, \dots, m$, and it will be refined by adding \mathcal{N}^i of equidistant nodes between each two mesh points.

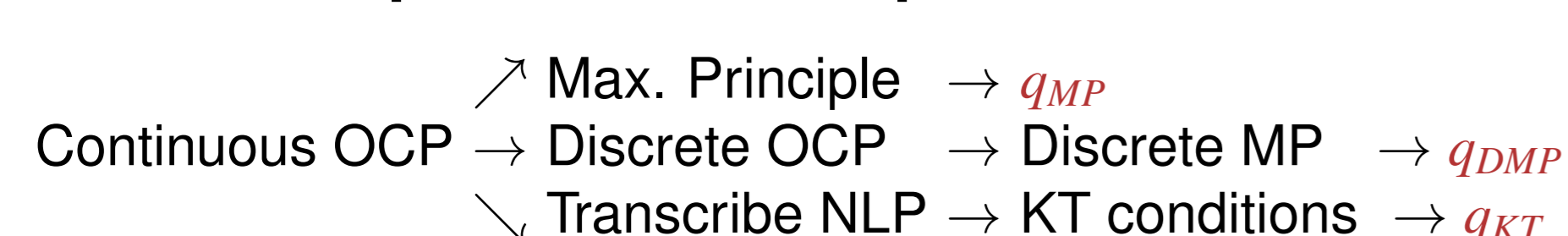


In our case, we define 3 refinement criteria:

- relative error of the trajectory (state variables) (ε_x)
- relative error of the adjoint multipliers (q multipliers) (ε_q)
- a combination of both criteria

and we consider a threshold for the relative error of the trajectory as the **stopping criterion**.

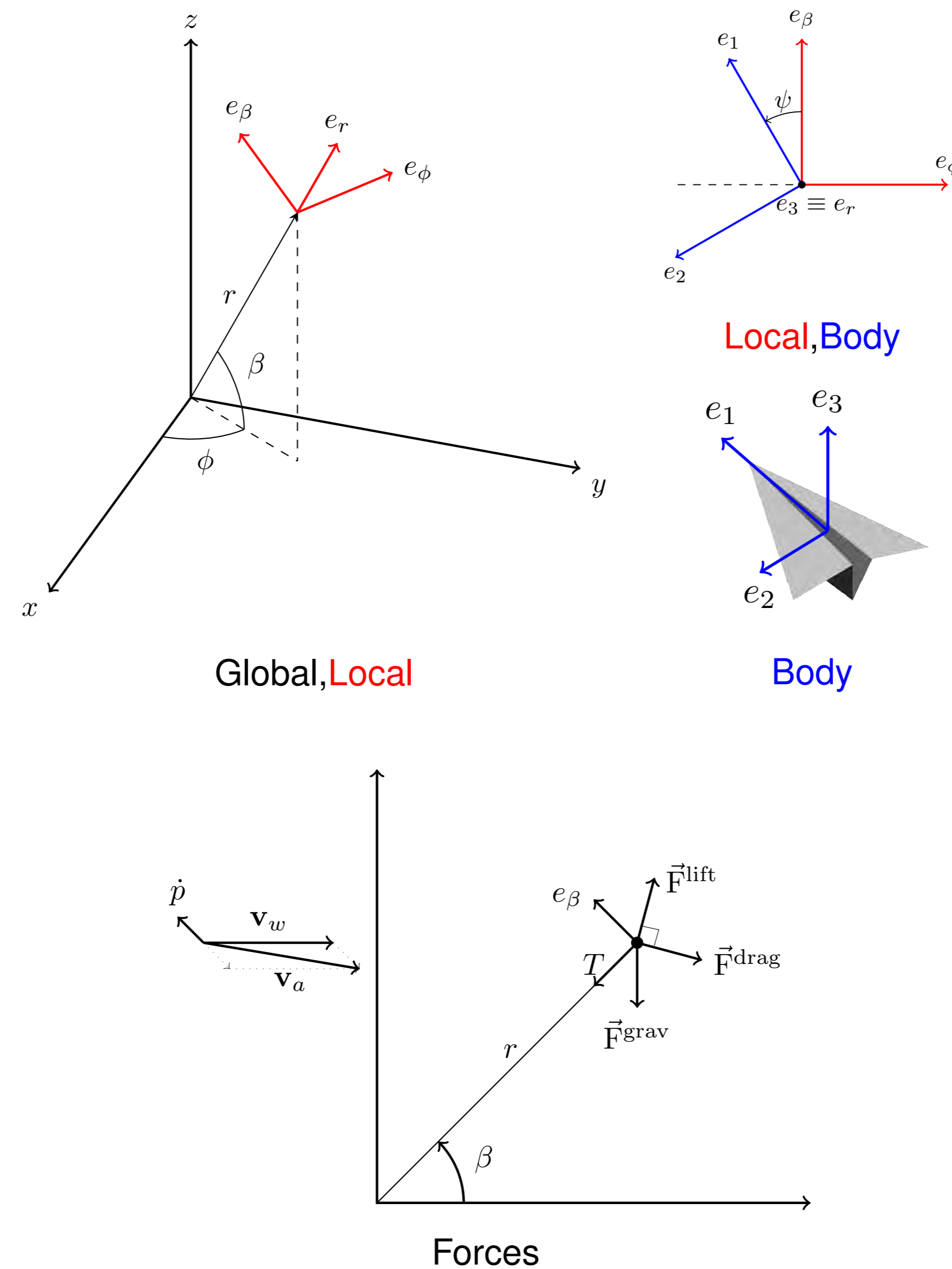
Relationship between multipliers: OCP and NLP



- q_{DMP} and q_{KT} coincide
- q_{DMP} is a discrete approximation of q_{MP}
- q_{KT} is an output of the NLP solver

Kite power system

Coordinate Systems and Forces



Optimal Control Problem

Consider $\mathbf{x} = (r, \phi, \beta, \psi, \dot{\phi}, \dot{\beta})$ and $\mathbf{u} = (v, \alpha, \omega)$:

$$\begin{aligned} & \text{Maximize } \int_0^{t_f} \dot{r}(-T) dt \\ & \text{subject to} \\ & \dot{\mathbf{x}}(t) = \mathbf{f}(\mathbf{x}(t), \mathbf{u}(t)) \quad \text{a.e. } t \in [0, t_f] \\ & \mathbf{x}(0) = \mathbf{x}_0 \\ & \mathbf{x}_{f, \min} \leq \mathbf{x}(t_f) \leq \mathbf{x}_{f, \max} \\ & \mathbf{x}_{\min} \leq \mathbf{x}(t) \leq \mathbf{x}_{\max} \quad \forall t \in [0, t_f] \\ & \mathbf{u}_{\min} \leq \mathbf{u} \leq \mathbf{u}_{\max} \quad \text{a.e. } t \in [0, t_f] \end{aligned}$$

where

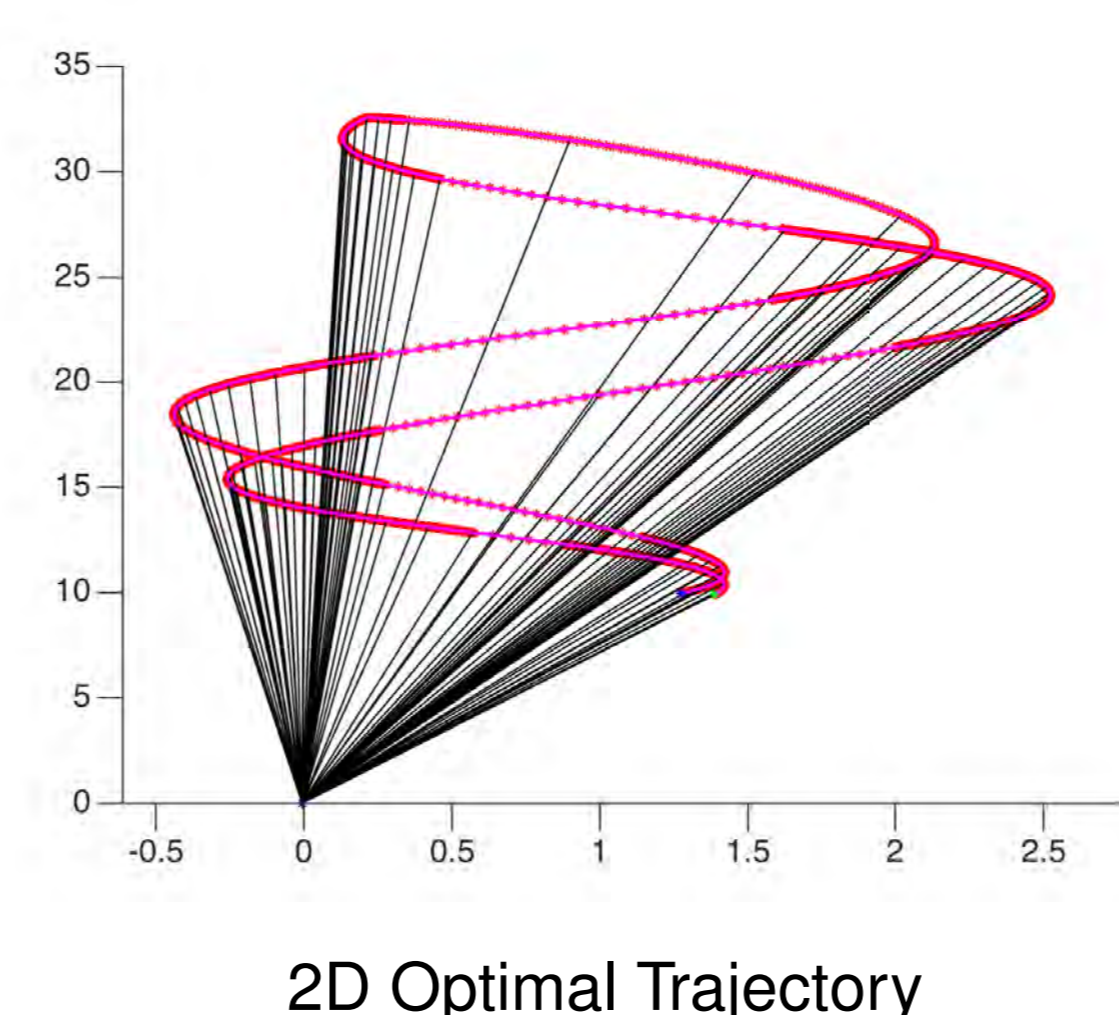
$$\mathbf{f}(\mathbf{x}(t), \mathbf{u}(t)) = \begin{bmatrix} v_t \\ \dot{\phi} \\ \dot{\beta} \\ \omega \\ (F_{\phi}^{aer}(\alpha) + F_{\phi}^{iner}) / (mr \cos(\beta)) \\ (F_{\beta}^{aer}(\alpha) + F_{\beta}^{grav} + F_{\beta}^{iner}) / (mr) \end{bmatrix}, \quad -T = F_r^{aer}(\alpha) + F_r^{grav} + F_r^{iner}.$$

Numerical results

The proposed algorithm is implemented in **MATLAB** using **ICLOCS** and **IPOPT**. The problem is solved using two meshes:

- π_{ML} mesh generate by the adaptive refinement strategy
- π_F equidistant spacing mesh considering the lowest Δt of π_{ML} .

2D Problem

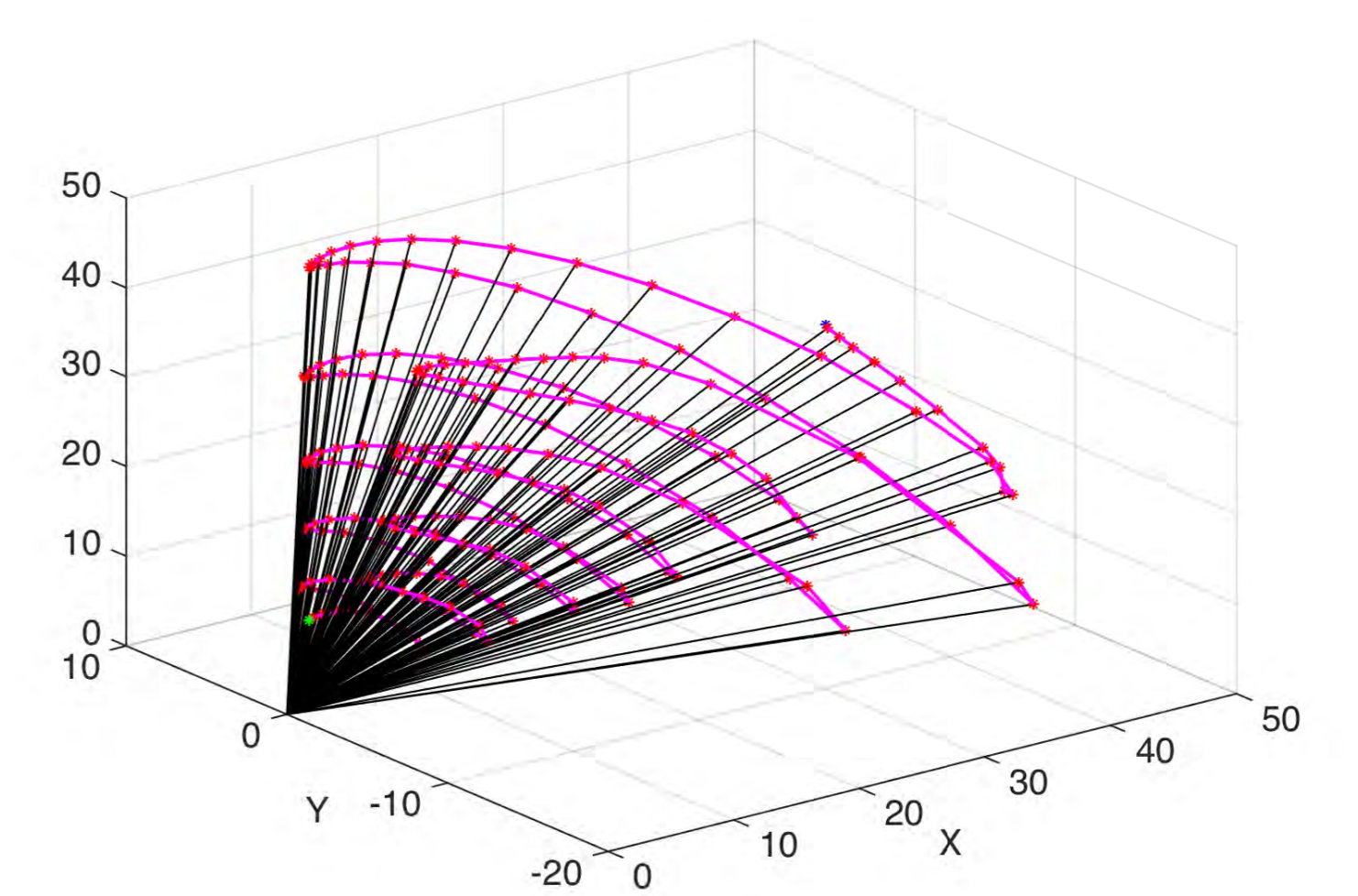


2D Optimal Trajectory

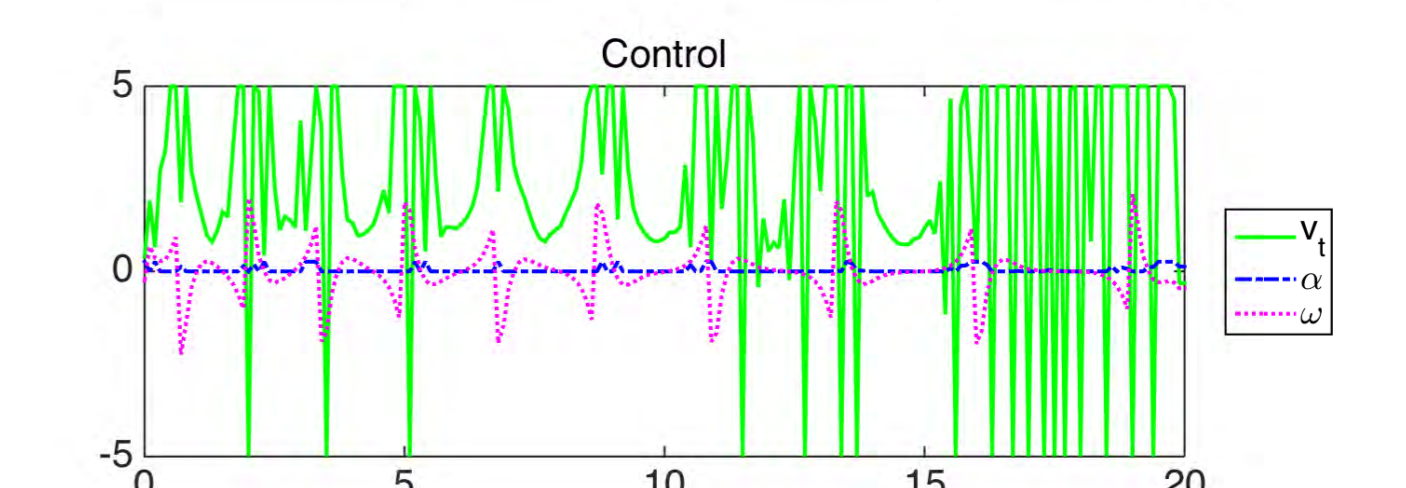
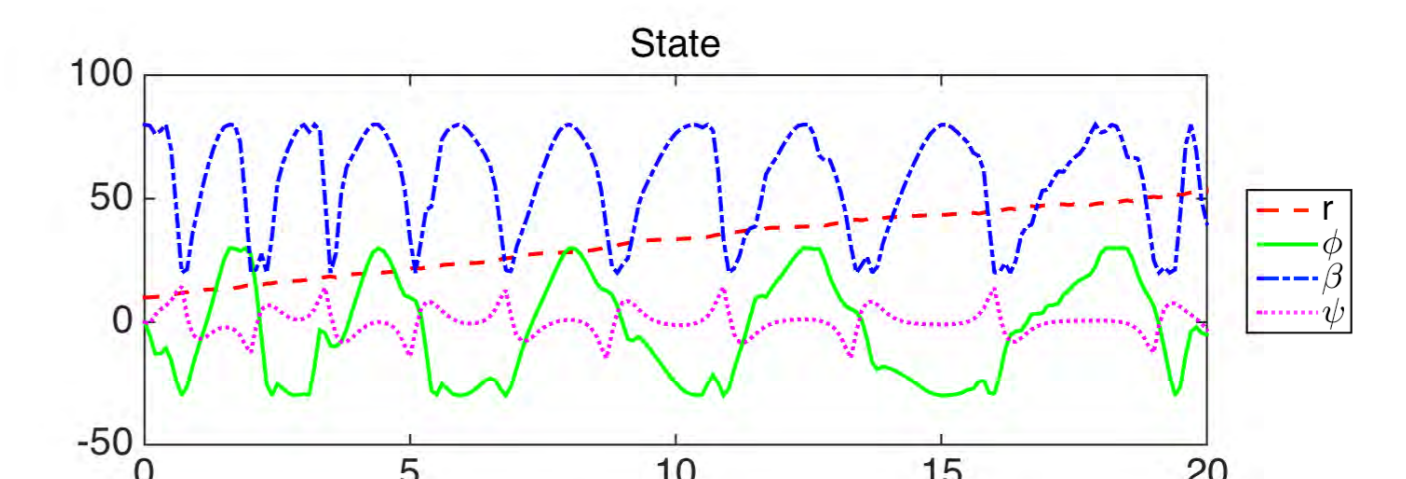
π_j	N_j	Δt_j	I_j	Objective	$\ \varepsilon_x^{(j)}\ _{\infty}$	CPU time (s)	
						Solve	ε_x
π_0	51	1/50	253	53.216968	1.795E ⁻²	4.754	0.107
π_1	474	1/500	122	54.156063	2.328E ⁻³	4.722	0.856
π_2	2877	1/5000	26	54.266839	3.168E ⁻⁴	6.107	3.383
π_{ML}	2877	1/5000	401	54.266839	3.168E ⁻⁴	15.583	4.346
π_F	5001	1/5000	1342	54.266873	2.201E ⁻⁴	431.234	10.913

Mesh comparison

3D Problem



Optimal Trajectory



State and Control

Concluding remarks

Main advantages of refinement strategies

- no need to define *a priori* the most appropriately mesh,
- local mesh resolution only where it's required,
- first solution obtained after few iterations since an initial coarse mesh is used,
- warm–start after refinement leads to faster re–solve,
- less nodes in the overall procedure results in significant savings in memory and computational cost.

... of our refinement algorithm

- multi–level refinement leads to faster convergence,
- local error on q gives a more accurate/faster error estimate,
- a solution is obtained very quickly (if the procedure ends in an early stage).

... in kite power systems

- much faster in the 2D problem,
- optimal solution found without informative initial guess in the 3D problem.

References

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Acknowledgements

Support of FCT – Fundação para a Ciência e Tecnologia – under Grants PTDC/EEA-CRO/116014/2009 and PTDC/EEI-AUT/1450/2012.