

Università di Pavia

## Transient Stability Constrained Optimal Power Flow Problems

HEXAGON 20 June 2024

Valentin Ilea

### *Motivation*

**POLITECNICO MILANO 1863** 

## HEXAGON

Inertia (GW.s)

376

359

340

324

307

290 260

224

190

184

154

140

108

 $-$ case13

 $- \case12$ 

case11

case10

case9 case8

case7

-case6

 $-$ case $5$ 

-case4

 $-$ case3

 $-case2$ 

 $-case1$ 

 $2.5$ 

 $\overline{c}$ 

- Standard Security Constrained OPF (SCOPF) problems give an optimal solution that guarantees steady-state security (control and state variables in-bounds, N-1 security criteria etc.)
- the increase of *converter-based* clean generation resources substitutes the *rotating-machine* based generation lowering the available system inertia and negatively affecting system's dynamic security:



• It is thus necessary to formulate an SCOPF that also considers Transient Stability aspects, in other words, a Transient Stability Constrained OPF (TSC-OPF)

 $\min \mathcal{C}(p) \qquad \rightarrow \quad p$  is the decision variable vector  $p = (P_g, Q_g, V, \theta)^T$ 

such that

 $g_S(p) = 0 \rightarrow \infty$  set of the **steady-state** equality constraints (PF eq, branch currents, etc)

 $h_s(p) \geq 0 \rightarrow \infty$  set of the **steady-state** inequality constraints (capability, box bounds, N-1 security conditions etc)



• It is thus necessary to formulate an SCOPF that also considers Transient Stability aspects, in other words, a Transient Stability Constrained OPF (TSC-OPF)

 $\min \mathcal{C}(p) \qquad \rightarrow \quad p$  is the decision variable vector  $p = (P_g, Q_g, V, \theta)^T$ 

such that

 $\dot{x} = f(x(t), y(t), p) \rightarrow x$  is the vector of state variables (rotor angles, generators frequencies, etc.)  $y$  is the vector of algebraic link variables dot operator is the derivative wrt. time

 $0 = g(x(t), y(t), p) \rightarrow$  link algebraic equations

 $x(t_0) = I_{x0}$   $\rightarrow$  set of initial conditions for x

 $y(t_0) = I_{y0}$   $\rightarrow$  set of initial conditions for y  $h(x(t), y(t)) \leq 0$   $\rightarrow$  stability constraints





• **Constraint transcription** is an algorithmic framework that decouples optimization algorithms and simulation tools [\*]. Differential equations are integrated outside the optimization process and interfaced with NLP solvers:



[\*] S. Abhyankar, G. Geng, M. Anitescu, X. Wang, and V. Dinavahi, "Solution techniques for transient stability-constrained optimal power flow - Part I," IET Gener. Transm. Distrib., vol. 11, no. 12, pp. 3177–3185, 2017





TSC-OPF – Numerical Optimization Methods - Constraint transcription HEXAGON

- 
- *Integration into optimization problem:*

$$
H(x(p, t), y(p, t)) = \sigma \int_0^T [\max(0, h(x(p, t), y(p, t))]^{\eta} dt = 0
$$

- *Remarks:*
	- $\triangleright$   $x(p,t)$  and  $x(p,t)$  are expressed, using trajectory analysis, as an approximation valid in the vicinity of  $p_0$ , the previous NLP optimal solution
	- $\triangleright$  remember that  $h(x(t), y(t))$  is the function quantifying the stability conditions
	- $\triangleright$  !!!  $\sigma$ ,  $\mu$  are user defined parameters to assure smoothness of convergence
	- $\triangleright$  the equality condition is, in general, to hard and may lead to divergence of NLP





TSC-OPF – Numerical Optimization Methods - Constraint transcription HEXAGON

• *Integration into optimization problem - relaxation*

 $H(x(p, t) - \rho, y(p, t) - \rho) \leq 0$ 

 $\triangleright$  where  $\rho$  is a vector of positive slack variables minimized in the objective function





- *Simulation discretization:* the differential equations for all time steps are discretized to non-linear algebraic equations by using a numerical integration scheme.
	- ➢ in *power systems* optimization the few work available have used the Taylor integration scheme for this

$$
x(t) - x(t - \Delta t) = \frac{\Delta t}{2} \cdot [f(x(t), y(t), p) - f(x(t - p), y(t - p), p)]
$$

where  $\Delta t$  is the integration step



## TSC-OPF – Numerical Optimization Methods - Simulation discretization TSC-OPF – Numerical Optimization Methods - Simulation discretization (HEXAGON)

• The solution of the differential equations at discrete time points is approximated by a Lagrange interpolating polynomial



[\*\*] J. D. Hedengren, R.A. Shishavan, K. M. Powell, T. F. Edgar, "Nonlinear modeling, estimation and predictive control in APMonitor," Computers & Chemical Engineering, vol. 70, 2014, pp. 133-148.





• The objective is to determine a matrix M that maps the derivatives to the nonderivative values:

$$
\begin{bmatrix} \dot{x}_1 \\ \dot{x}_2 \\ \dot{x}_3 \end{bmatrix} = M \left( \begin{bmatrix} x_1 \\ x_2 \\ x_3 \end{bmatrix} - \begin{bmatrix} x_0 \\ x_0 \\ x_0 \end{bmatrix} \right)
$$

• Time points for each interval are chosen according to *Lobatto quadrature*; In the case of 4 nodes per horizon step, the internal values are chosen at  $t_{1,2} =$ 1 2  $\pm$ 5 10 ; time points are shifted to a reference time of zero  $(t_0 = 0)$ and a final time of  $t_f = 1$ 

TSC-OPF – Numerical Optimization Methods - Simulation discretization ISC-OPF – Numerical Optimization Methods - Simulation discretization Mexicon Method

• Substituting the polynomial into the mapping relationship:

$$
\begin{bmatrix}\nB + 2Ct_1 + 3Dt_1^2 \\
B + 2Ct_2 + 3Dt_2^2 \\
B + 2Ct_3 + 3Dt_3^2\n\end{bmatrix} = M \begin{bmatrix}\nBt + Ct_1^2 + Dt_1^3 \\
Bt + Ct_2^2 + Dt_2^3 \\
Bt + Ct_3^2 + Dt_3^3\n\end{bmatrix}
$$
\n
$$
\implies M = \begin{bmatrix}\n1 & 2t_1 & 3t_1^2 \\
1 & 2t_2 & 3t_2^2 \\
1 & 2t_3 & 3t_3^2\n\end{bmatrix} \begin{bmatrix}\nt_1 & t_1^2 & t_1^3 \\
t_2 & t_2^2 & t_2^3 \\
t_3 & t_3^2 & t_3^3\n\end{bmatrix}^{-1}
$$

• Now, the mapping relationship is fully determined:

$$
M^{-1} \begin{bmatrix} \dot{x}_1 \\ \dot{x}_2 \\ \dot{x}_3 \end{bmatrix} = \begin{bmatrix} x_1 \\ x_2 \\ x_3 \end{bmatrix} - \begin{bmatrix} x_0 \\ x_0 \\ x_0 \end{bmatrix}
$$





TSC-OPF – Numerical Optimization Methods - Simulation discretization ISC-OPF – Numerical Optimization Methods - Simulation discretization Mexicon Method

• Example:



$$
\tau \cdot M \cdot \begin{pmatrix} x_1 & x_0 \\ x_2 - x_0 \\ x_3 & x_0 \end{pmatrix} = - \begin{pmatrix} x_1 \\ x_2 \\ x_3 \end{pmatrix}
$$





- 
- *GOAL*: find the minimum inertia at control area level so that the frequency stability following major events (short-circuits, line trips, etc.) is guaranteed.



- 
- *GOAL*: find the minimum inertia at control area level so that the frequency stability following major events (short-circuits, line trips, etc.) is guaranteed.

$$
\min \sum_{i \in \text{BUS}} H_i
$$

such that

$$
\frac{dP_{mi}}{dt} = \frac{1}{T_{govi}} \left[ P_{mi}^0 + \frac{1}{R_i} \left( \frac{\Delta \omega_i}{\omega_s} - P_{mi} \right) \right]
$$





### TSC-OPF – Frequency stability model https://www.assetter.com/discreence/stability model

• *Grid model & algebraic link equations*:







IIIIII

min  $\sum H_i$  $i\epsilon$ **EDUS** 

<u>|||||||||||||</u>

such that

$$
P_{e,i}(t) = \sum_{j=1}^{ngen} \left( E_{E_i} * E_{E_j} * |Y_{red,ij}| * \cos\left(\delta_i(t) - \delta_j(t) - \theta_{ij}\right) \right)
$$
\n
$$
\left( \sum_j H_j \cdot A_{n,j} \right) \cdot \omega_a^{Col^t} = \sum_j H_j \cdot A_{n,j} \cdot \omega_j^t \qquad \text{Col frequency calculation}
$$
\n
$$
ROCOF_i^{500} \le 2 \qquad ROCOF_i^{1000} \le 1.5 \qquad ROCOF_i^{2000} \le 1.25 \qquad \text{Stability Constraints}
$$





TSC-OPF – Frequency stability model New York Control of the HEXAGON

• *Modeling events*:

$$
P_{e,i}(t) = \sum_{j=1}^{ngen} \left( E_{E_i} * E_{E_j} * \left| Y_{red,ij} \right| * \cos \left( \delta_i(t) - \delta_j(t) - \theta_{ij} \right) \right)
$$











## TSC-OPF – Frequency stability model Numerical test 1: method performance<br>Numerical test 1: method performance

- **Rueda Network (a CWE equivalent)**:
- 16 buses with generating units
- Optimization model transient simulation results were compared to results of Digsilent, a tool for PS transient simulation
- Shortcircuit at bus A2 is applied and cleared after 200 ms







### TSC-OPF – Frequency stability model<br>Numerical test 1: method performance<br> Ш





## TSC-OPF – Frequency stability model<br>Numerical test 1: method performance<br> IIIIII







## TSC-OPF – Frequency stability model<br>
Numerical test 3: Beal Case Sicily stid Numerical test 2: Real Case – Sicily grid

### **Sicily Network**:

- Analyzed scenario: 8<sup>th</sup> May 2019
- minimum inertia available: 71.5 s
- maximum inertia available: 247 s
- a series of events have been considered and the minimum inertia required was determined as the envelope (max(min))



[Grid Map downloads \(entsoe.eu\)](https://www.entsoe.eu/data/map/downloads/)



## TSC-OPF – Frequency stability model<br>Numerical test 2: Real Case – Sicily grid Numerical test 2: Real Case – Sicily grid

- **Sicily Network**: 90 MW demand disconnection
- the optimizer brings the inertia to the minimum possible
- stability limit is well satisfied  $\rightarrow$  not a critical case





## TSC-OPF – Frequency stability model<br>Numerical test 2: Real Case – Sicily grid Numerical test 2: Real Case – Sicily grid

### **Sicily Network**: Separation from continent



- the optimizer brings the inertia to the minimum possible while ROCOF1000 limit is pushed to its lower bound
- the case is critical in the minimum inertia is  $H_{min} = 108 s$







The minimum inertia requirement is not satisfied by the initial dispatch provided by the energy market. In ASM formulation a minimum zonal inertia constrain is added:

$$
\sum Y_{u,t} \cdot E_u \ge \sum (Y_{u,t} \cdot A_u^n) \cdot H_{min,t}
$$

where:

- $H_u$  is the inertia of generating unit  $u$ ;
- $E_u$  is the kinetic energy of generating unit  $\mu$ ;
- $A_u^n$  is the apparent nominal power of generating unit  $u_i$ ;
- $H_{min,t}$  is the minimum level of inertia required.



# Future Work **HEXAGON**

- test other "orthogonal collocation" approximation functions for  $x(t)$  that could allow for a larger  $\Delta t$
- consider also converter-based DERs dynamic models with synthetic inertia capabilities





## HEXAGON

### **Questions**





