

Panel session on “Handling Uncertainties and Use of Equivalents in Dynamic Security Assessment”  
Ref. 19PESGM2190

# Model Reduction of Active Distribution Networks under Uncertainty

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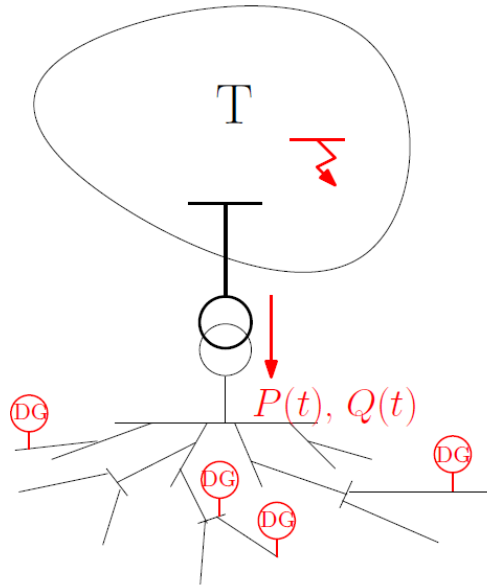
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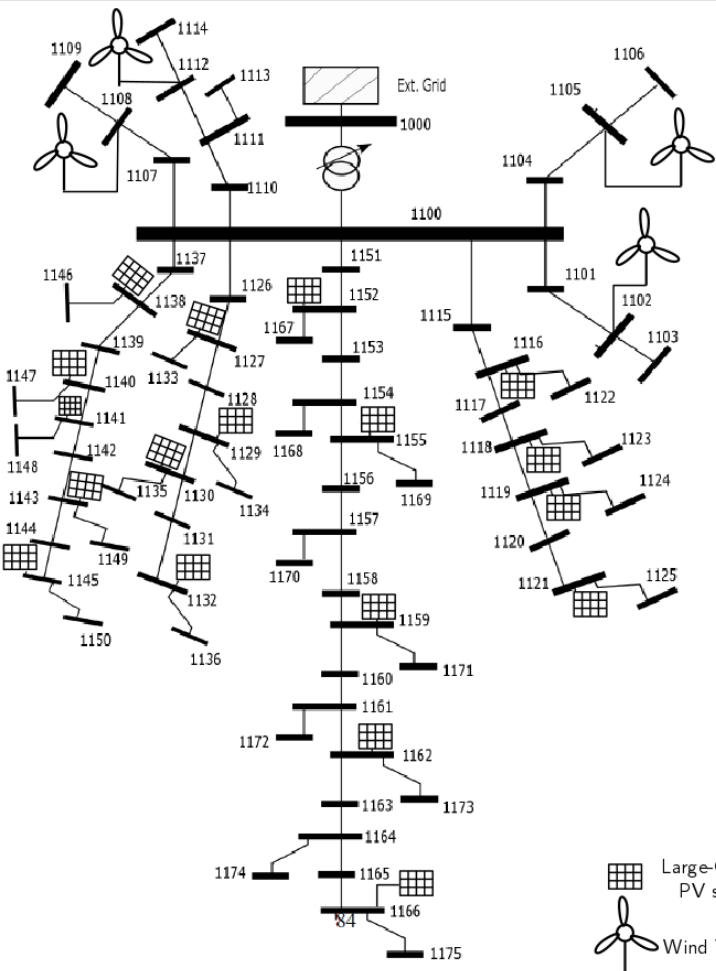


- More and more Inverter-Based Generators connected to distribution grids
- distribution networks become active
- their influence on the whole power system dynamics increases
- it is increasingly important for TSOs to model those Active Distribution Networks (ADNs) in their dynamic simulations

- Dynamic simulations of combined Transmission – Distribution system are impractical
  - large computing times
  - heavy model maintenance
  - confidentiality issue
- DSOs process their own data and transmit to the TSO simplified, reduced-order models of their distribution systems : dynamic equivalents
  - to be attached to the transmission system model
  - no confidentiality issue

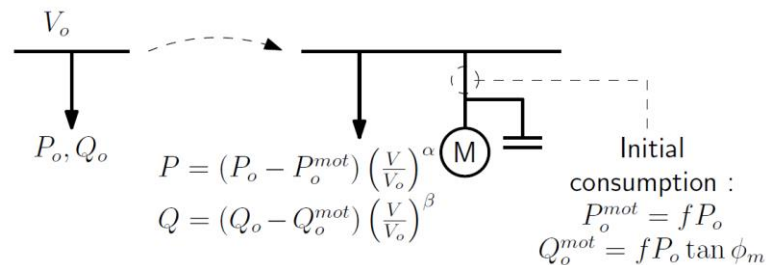


- For use in simulation of **large disturbances** in the transmission system
- accurate in terms of  $P(t)$  and  $Q(t)$  power flows in the distribution transformer
- accounting for **discrete controls** of dispersed units
  - dynamic voltage support, undervoltage tripping, etc.
- compatible with TSO dynamic simulation software
- physically intuitive → **“grey-box” model**
  - includes “physical” components with known models
  - but unknown parameters
- easily updated when operating point changes.



## Loads :

- static part : exponential model
- dynamic part : 3<sup>rd</sup>-order induction motor model

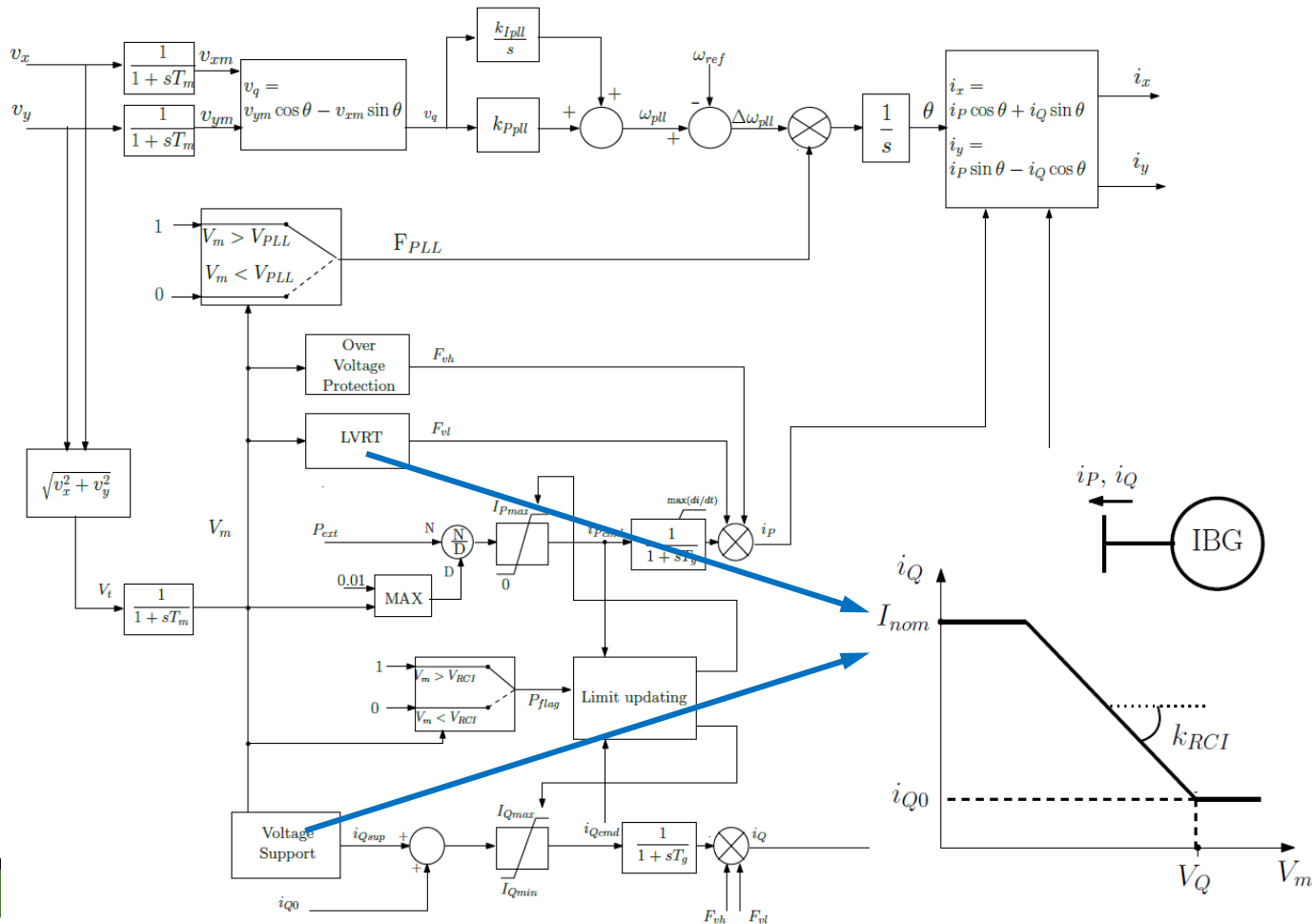


## Inverter-Based Generators (IBGs) :

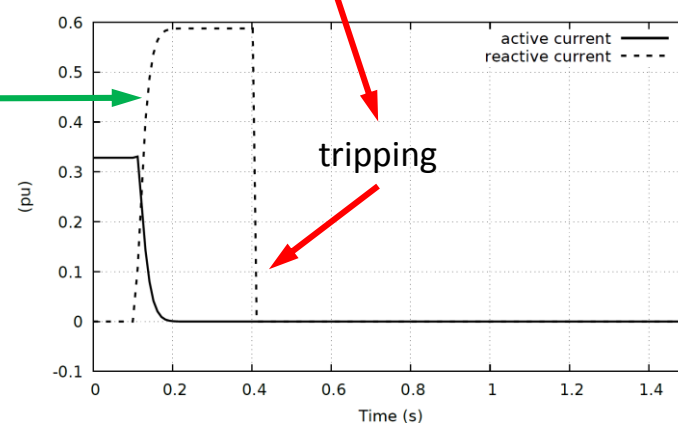
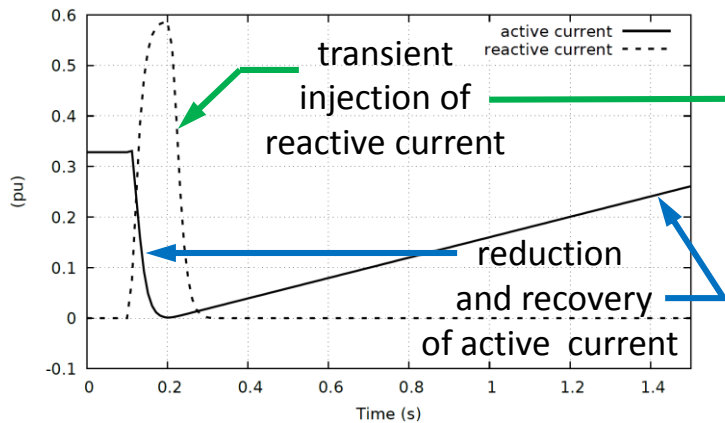
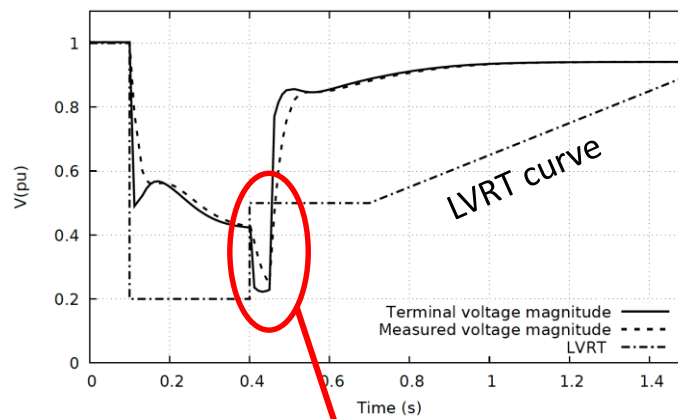
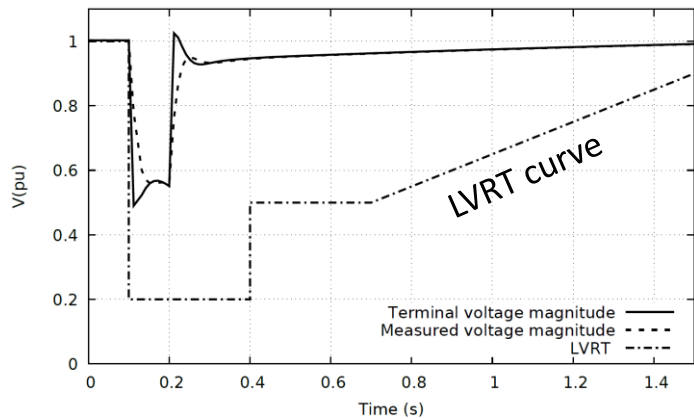
- Phase Locked-Loop (PLL)
- Low Voltage Ride-Through (LVRT)
- dynamic volt. support by reactive current injection
- limited rate of active current recovery after limitation

IBG generic model reproducing the response to voltage variations required by most grid codes

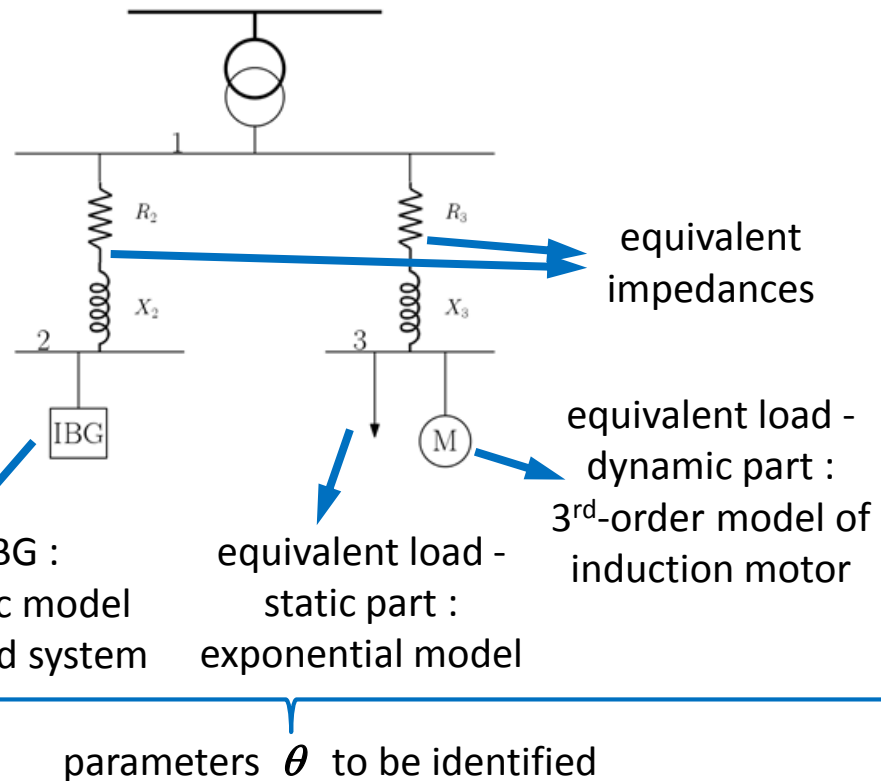
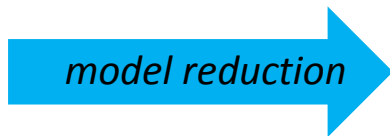
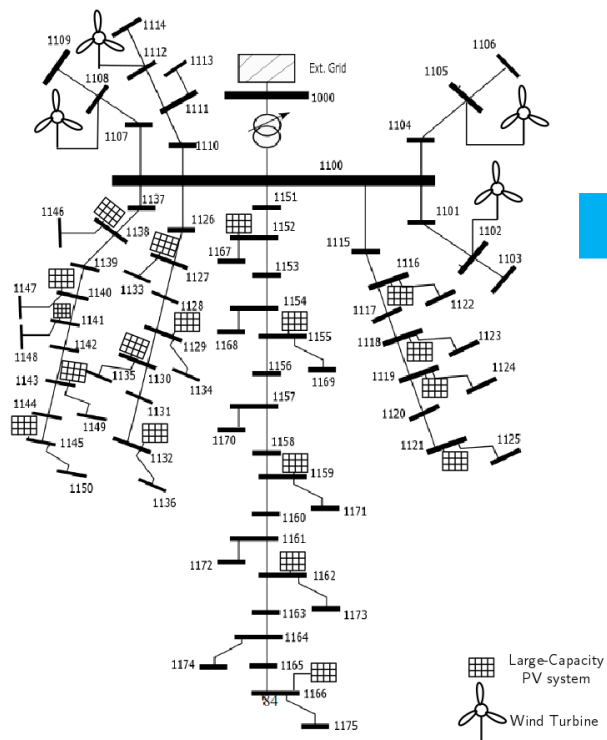
- NC RfG (ENTSO-e)
- VDE AR N 4105/ BDEW MV (Germany)
- IEEE 1547
- etc.



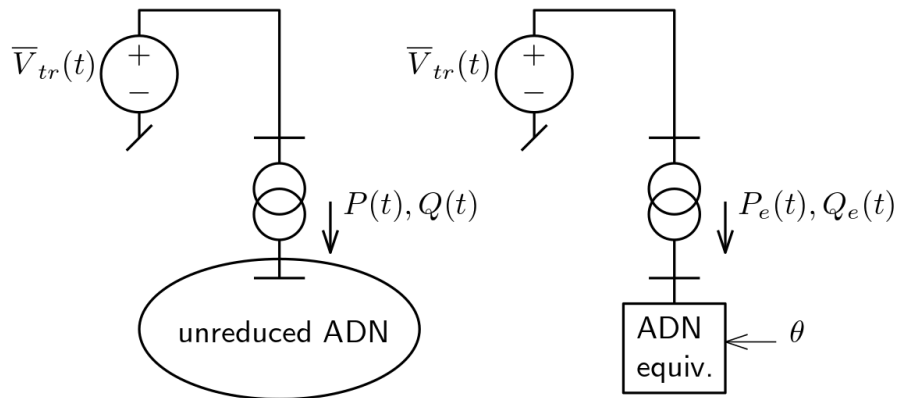
# Example of IBG response to voltage dips



# ADN dynamic equivalent : grey-box model

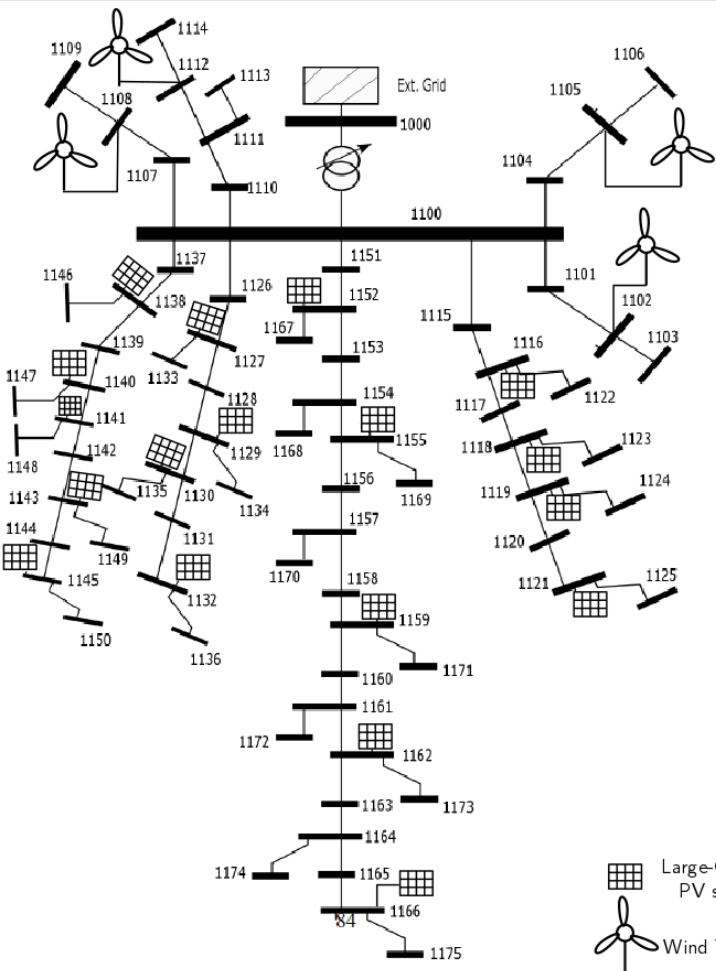






- Measurements not available...
- transmission system replaced by voltage source  $\bar{V}_{tr}(t)$  imposing various disturbances
  - voltage magnitude, phase angle, frequency
- parameters  $\theta$  of the ADN equivalent tuned so that  $(P_e, Q_e)$  approaches  $(P, Q)$  of unreduced system

- Dynamic models involve parameters not known accurately
  - loads : models are already simplified equivalents
  - IBGs : grid codes leave freedom on some parameters
- Impact assessed through Monte-Carlo simulations
  - at a given initial operating point, a disturbance is simulated for  $s$  instances of the same model corresponding to randomly drawn parameter vectors  $\mathbf{p}_1, \dots, \mathbf{p}_s$ .
  - $s$  randomized dynamic responses to the disturbance
  - statistics computed at each point in time

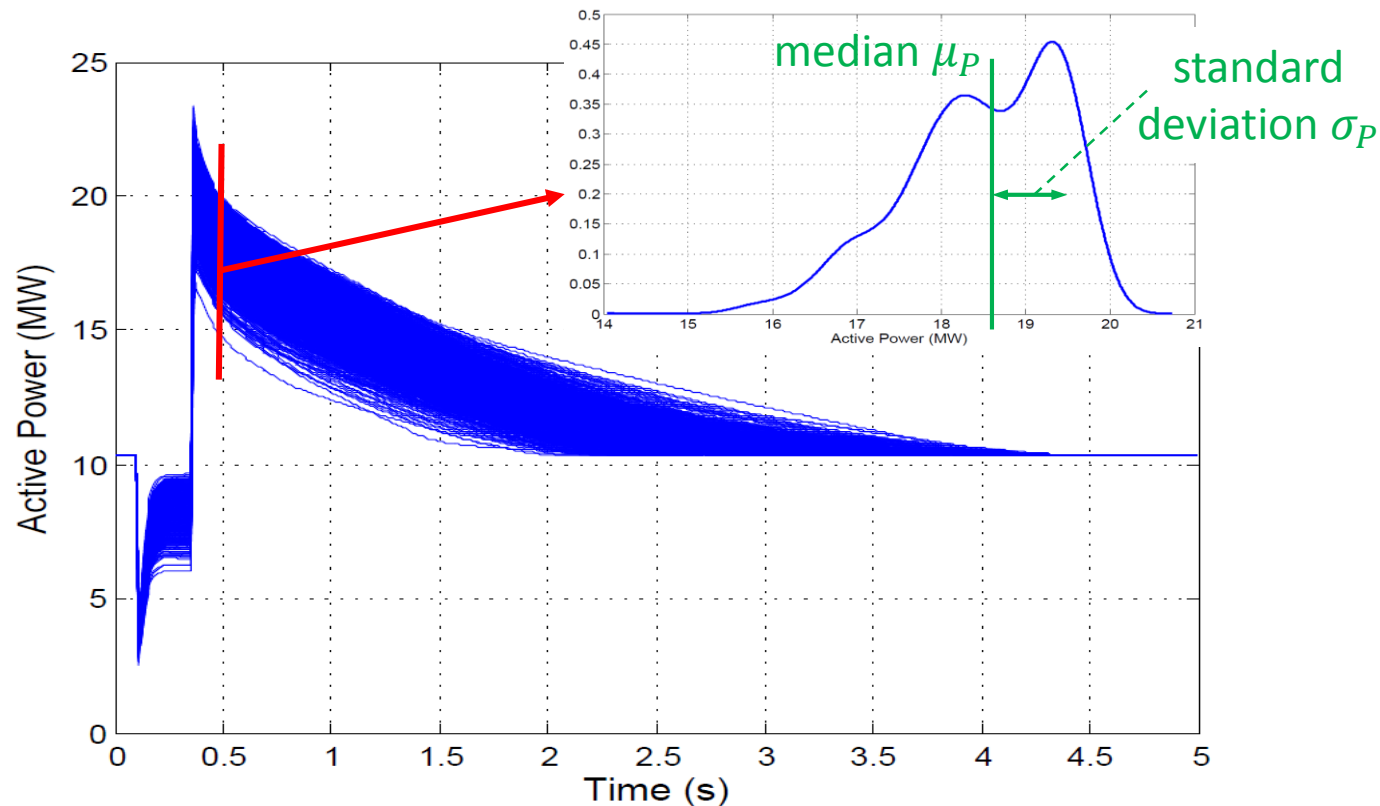


- 75 buses      53 loads      22 IBGs

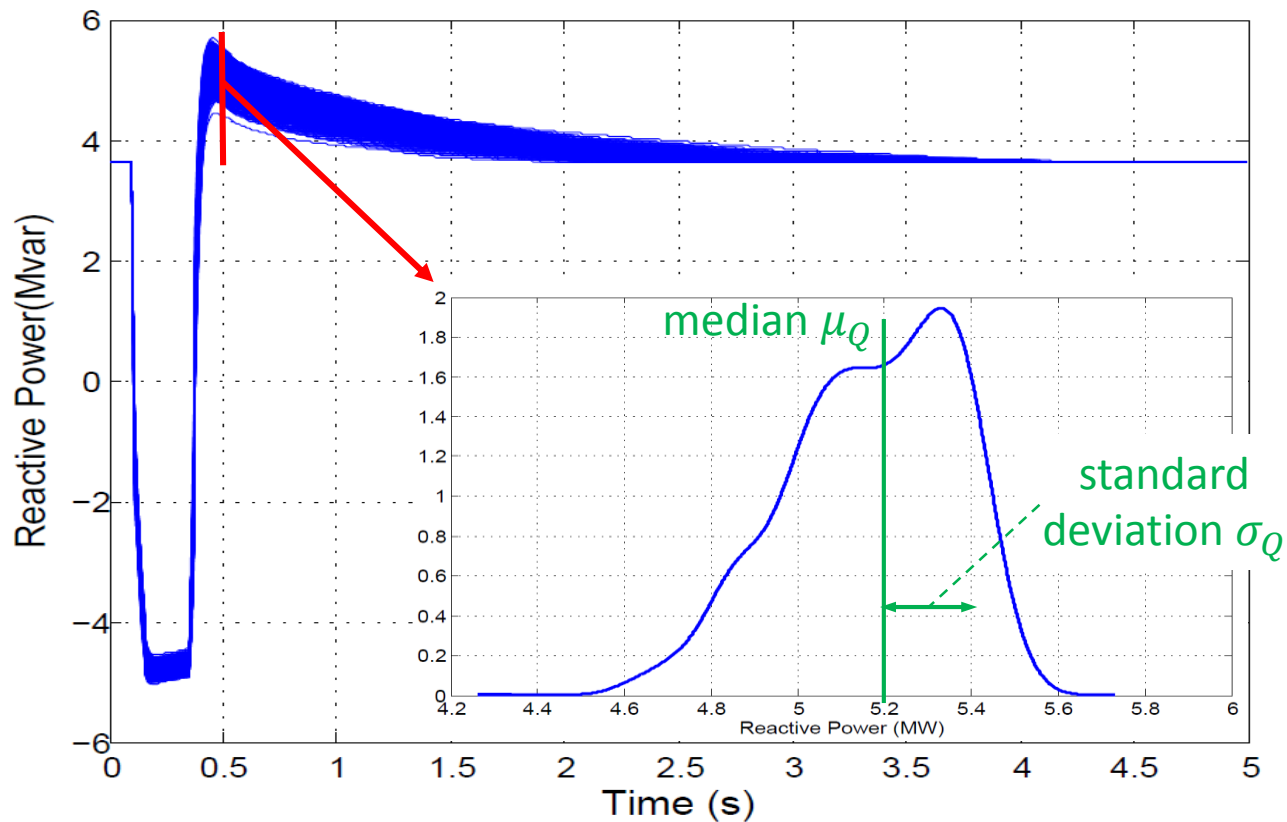
	MW	Mvar
Consumption of loads	19.95	2.83
Production of IBGs	9.80	0
Power flow in transformer	10.33	2.96

- Nb of differential-algebraic equations :
  - unreduced model : 3297
  - equivalent : 117
- Nb of components in  $\theta$  :
  - 17 initially tested
  - 7 removed : negligible impact identified

Responses of active power to a transmission voltage dip of 0.5 pu during 250 ms



Responses of reactive power to a transmission voltage dip of 0.5 pu during 250 ms



derivative-free,  
metaheuristic  
optimization :

*Differential  
Evolution  
algorithm*

$$\min_{\theta} F(\theta) = \frac{1}{d} \sum_{j=1}^d [F_P(\theta, j) + F_Q(\theta, j)]$$

$$\text{with } F_P(\theta, j) = \frac{1}{N} \sum_{k=1}^N \left[ \frac{P_e(\theta, j, k) - \mu_P(j, k)}{\sigma_P(j, k)} \right]^2$$

$$F_Q(\theta, j) = \frac{1}{N} \sum_{k=1}^N \left[ \frac{Q_e(\theta, j, k) - \mu_Q(j, k)}{\sigma_Q(j, k)} \right]^2$$

$$\theta^L \leq \theta \leq \theta^U$$

response of  
equivalent fitted  
to the median

lower weight if  
larger dispersion

$d$  : number of “training” disturbances

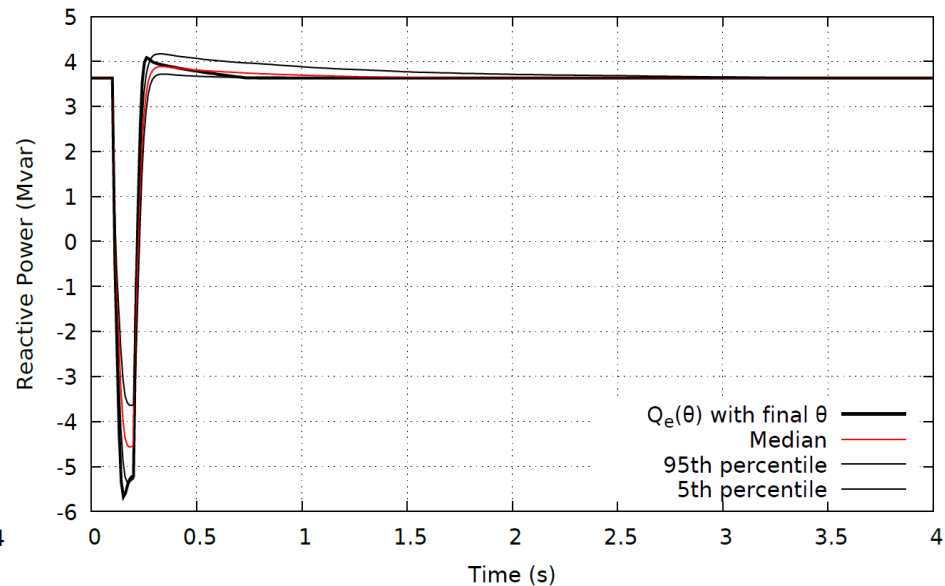
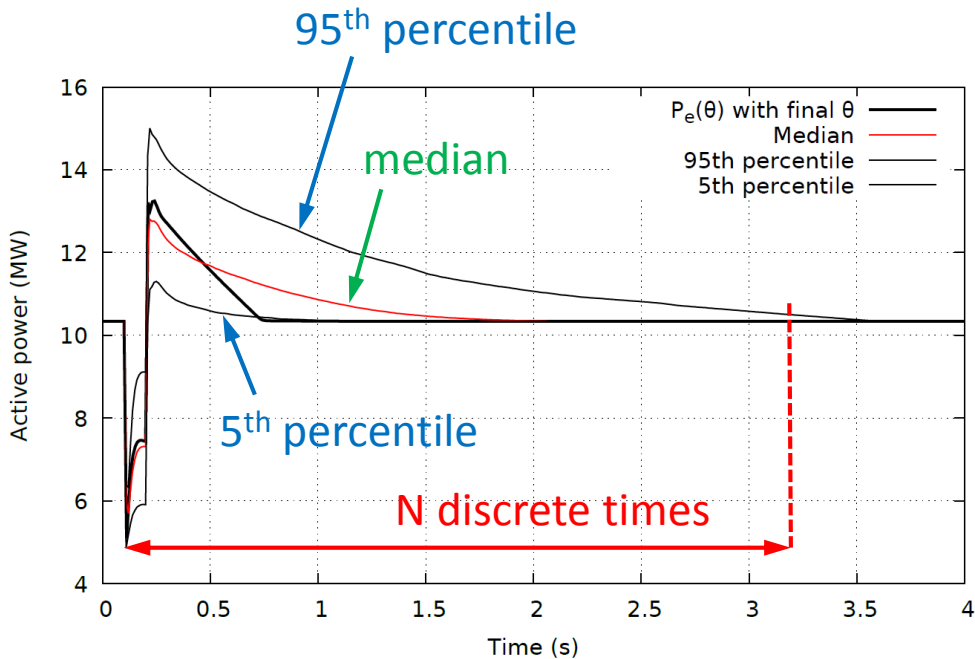
$N$  : number of discrete times of simulation

$\mu_P(j, k)$  : median of distribution of  $P$  at time  $k$  for the  $j$ -th disturbance

$\sigma_P(j, k)$  : corresponding standard deviation

$\mu_Q(j, k)$  and  $\sigma_Q(j, k)$  : same for  $Q$

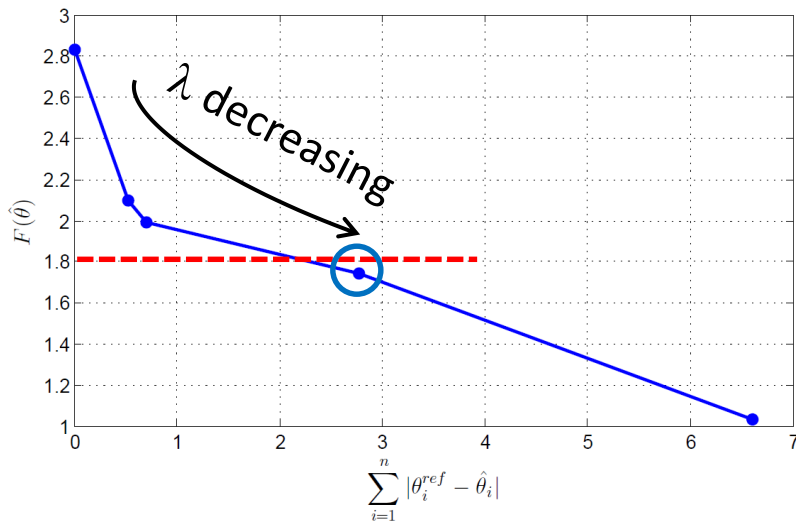
Responses to a transmission voltage dip of 0.3 pu during 100 ms



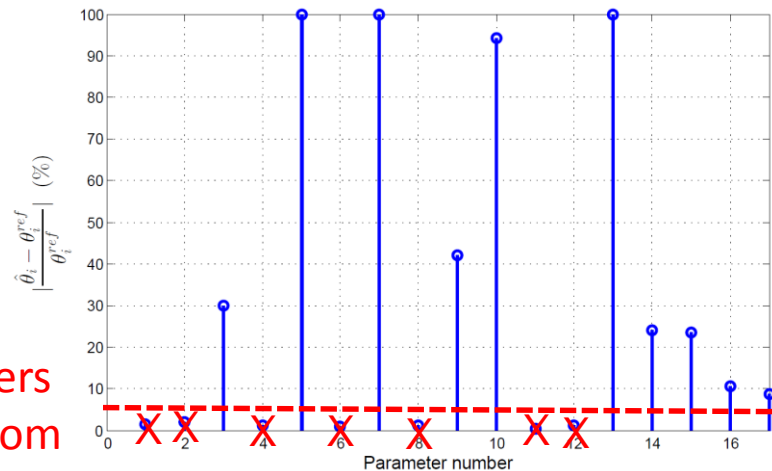
# Keeping the dimension of $\theta$ as small as possible

- To make the reduced model :
  - easier to optimize (faster convergence of DE)
  - more consistent from one case to another
  - easier to interpret
- variant of *Least Absolute Shrinkage and Selection Operator (LASSO)* method

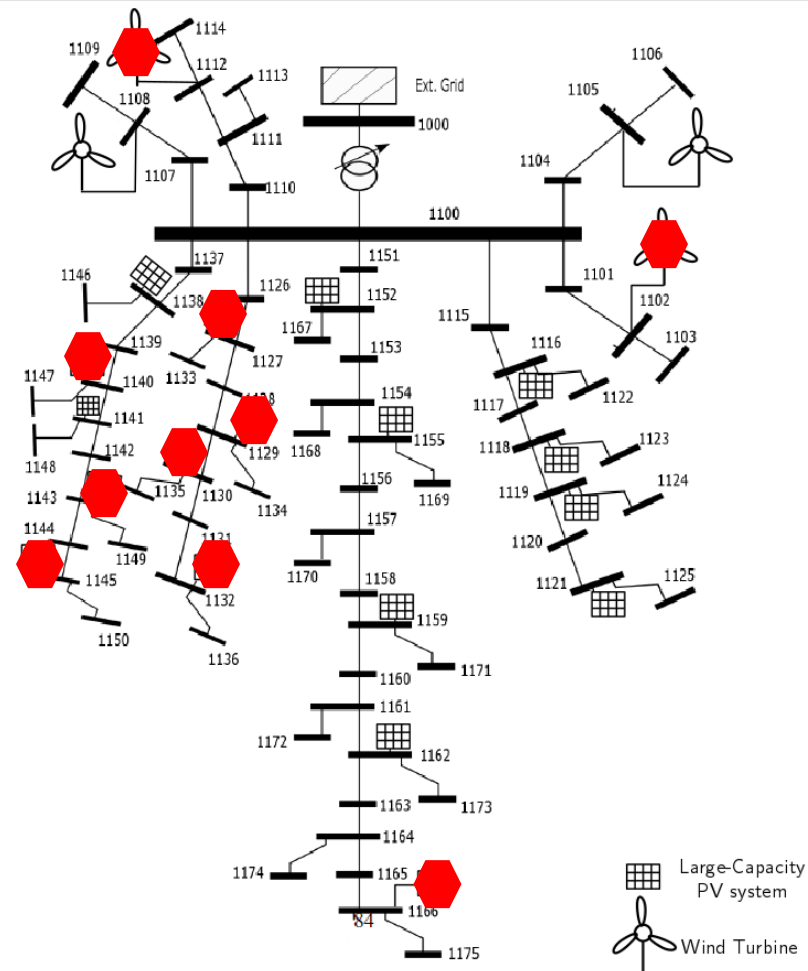
$$\min_{\theta} F(\theta) + \lambda \sum_{i=1}^n |\theta_i^{ref} - \hat{\theta}_i|$$



7 parameters  
removed from  
final  $\theta$

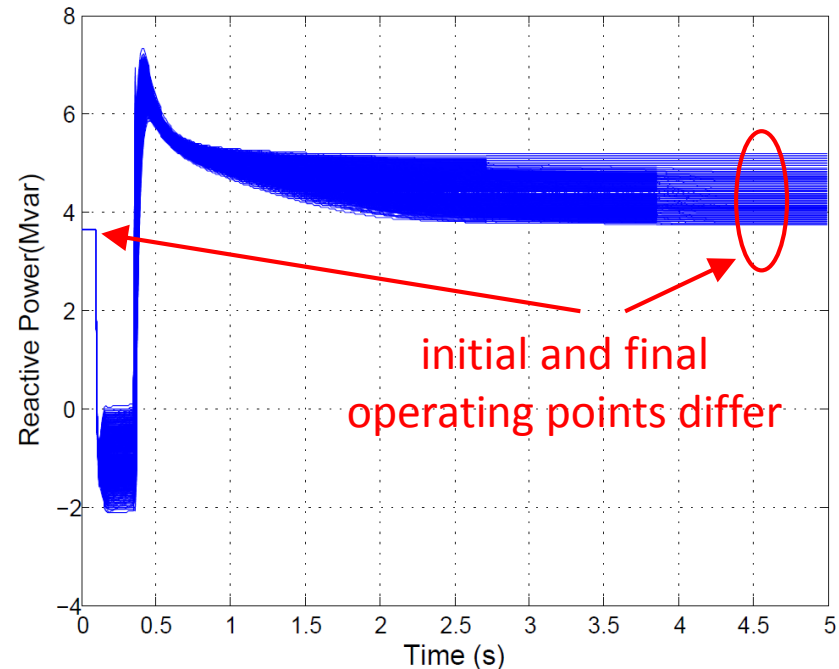
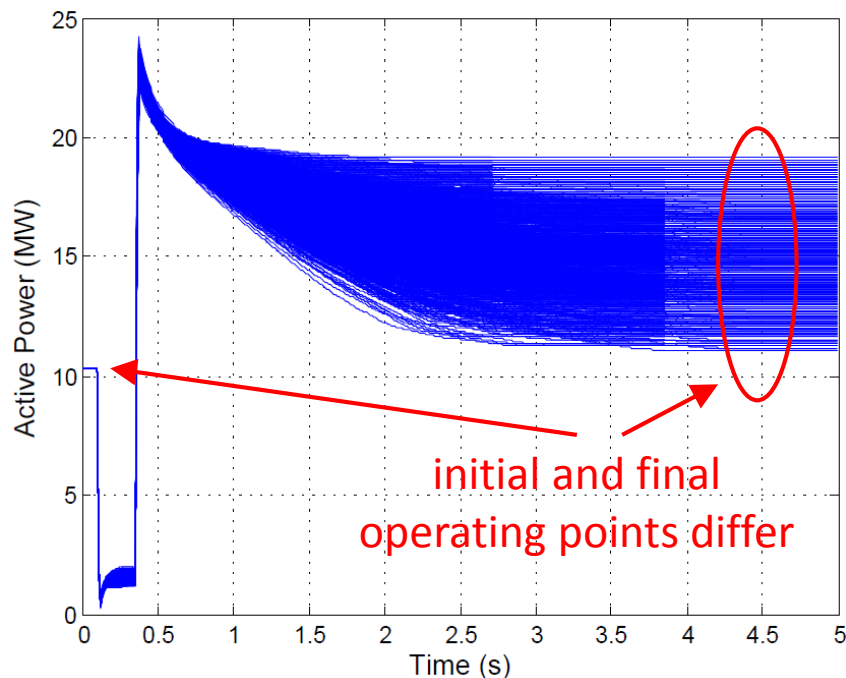






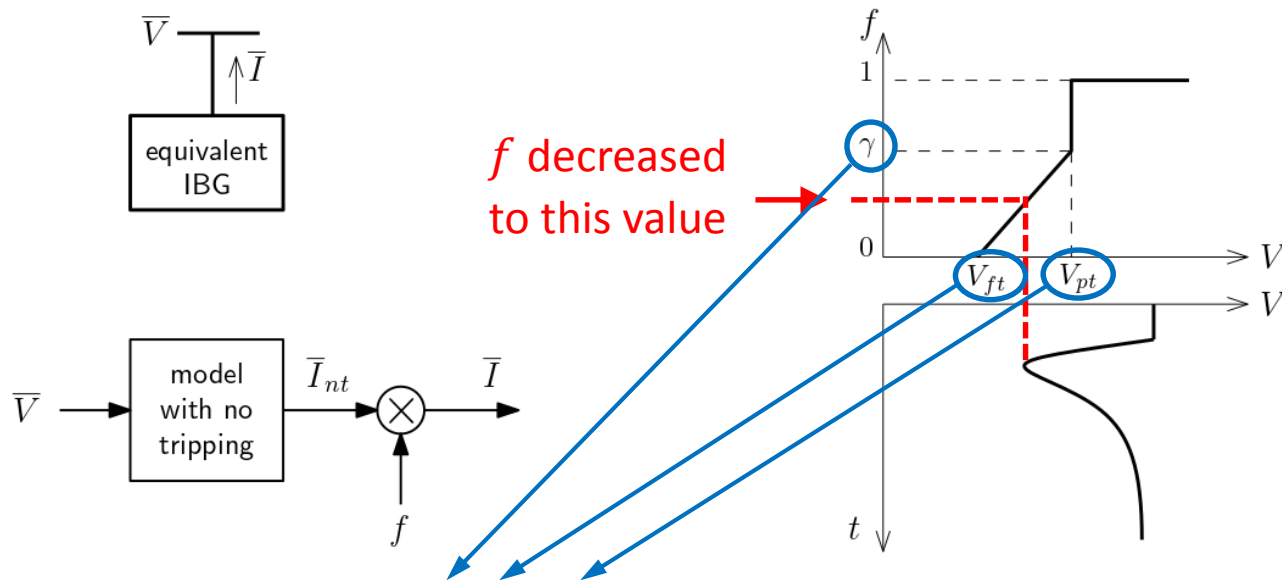
- If the transmission voltage drop is deep enough, some IBGs may disconnect
  - voltage falls below LVRT curve
- Example : transmission voltage drop of 0.8 pu lasting 250 ms

Responses to a transmission voltage dip of 0.8 pu during 250 ms



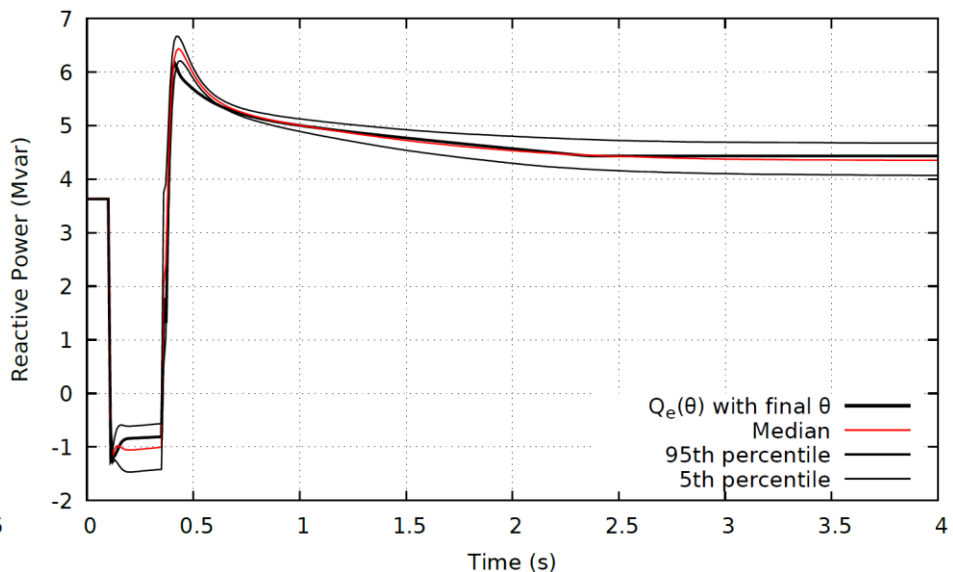
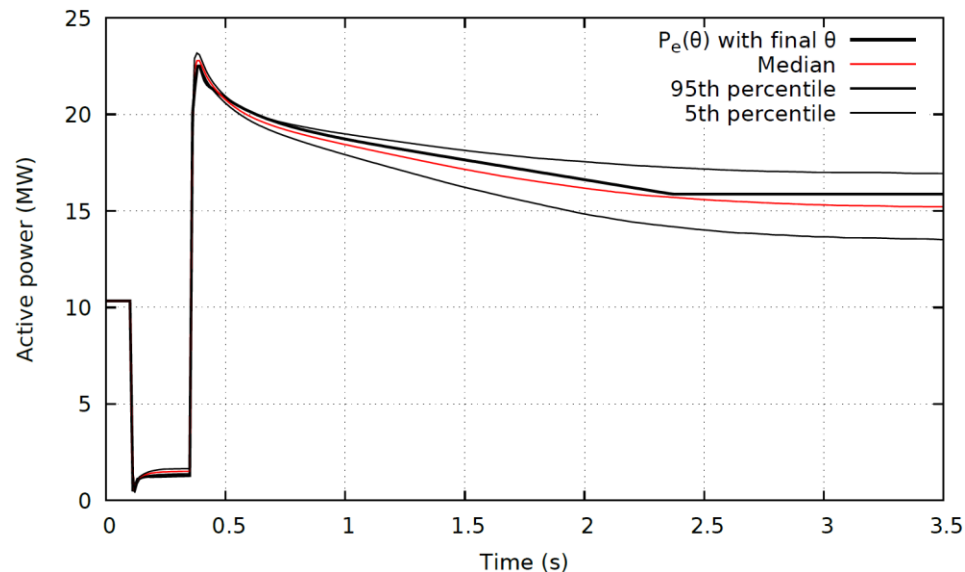
- IBGs with voltage falling below the LVRT curve may trip
- responses with tripping randomized, together with other parameters

- by reducing the current injected by the equivalent IBG



- $V_{pt}$ ,  $V_{ft}$ ,  $\gamma$  are adjusted by weighted least squares
  - after dealing with the other components of  $\theta$

Responses to a transmission voltage dip of 0.8 pu during 250 ms



- ADN equivalent for simulation of large disturbances at transmission level
- grey-box model
- equivalent significantly smaller than unreduced system
- strong nonlinearities and discontinuities considered
  - in particular, partial tripping of IBGs
- weighted-least square identification
  - number of parameters to identify : as small as possible (LASSO)
- impact of model uncertainties identified from Monte-Carlo simulations
  - fitting the “average” response + weighting factors to reflect dispersion
- equivalent trained with multiple disturbances
- good results on a test system with high penetration of renewable energy sources.