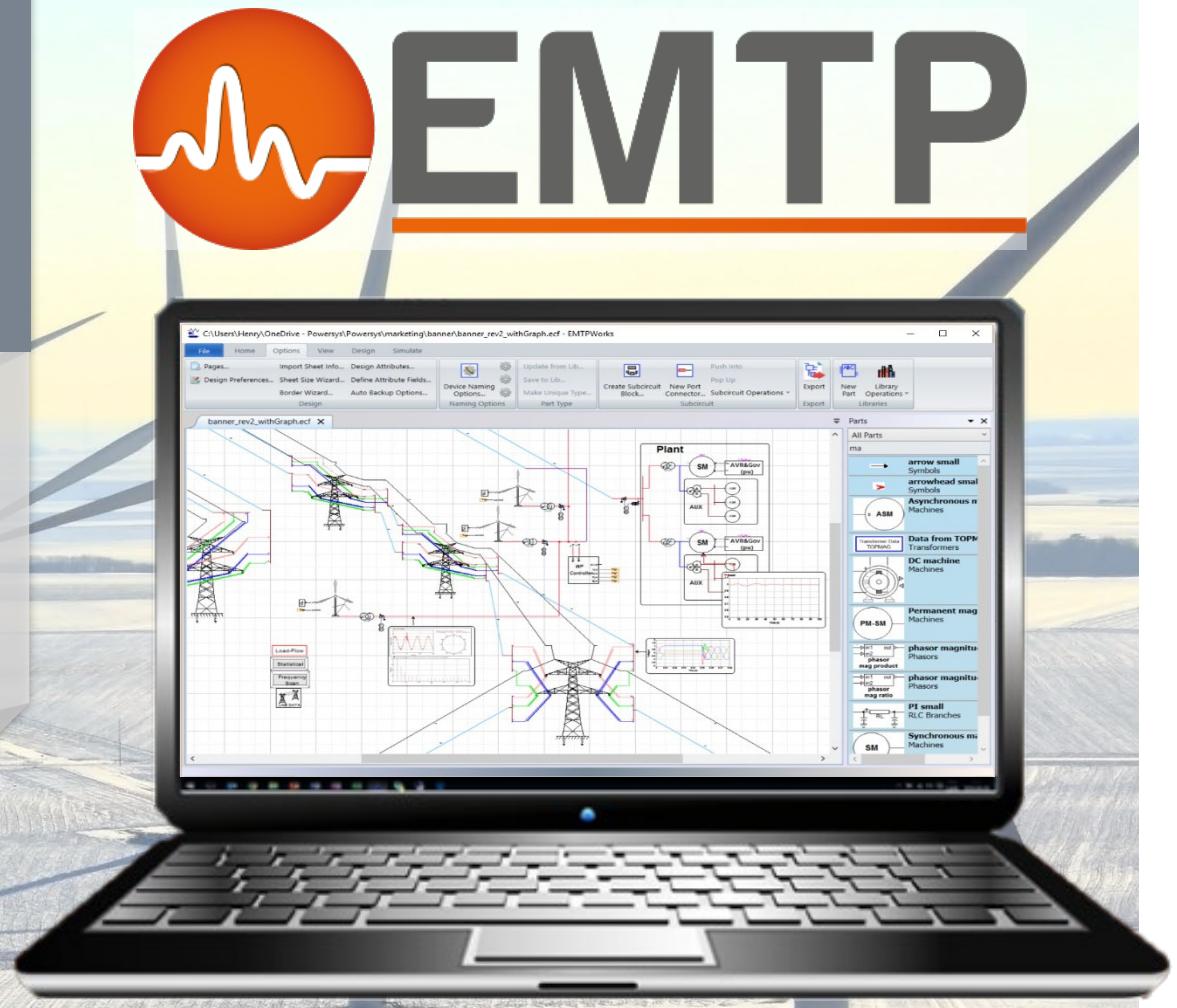


# Automatic model parameter determination to fit IBR behavior.

Willy Nzale, *EMTP*  
Victor Velar, **COORDINADOR**



More info

[www.EMTP.com](http://www.EMTP.com)

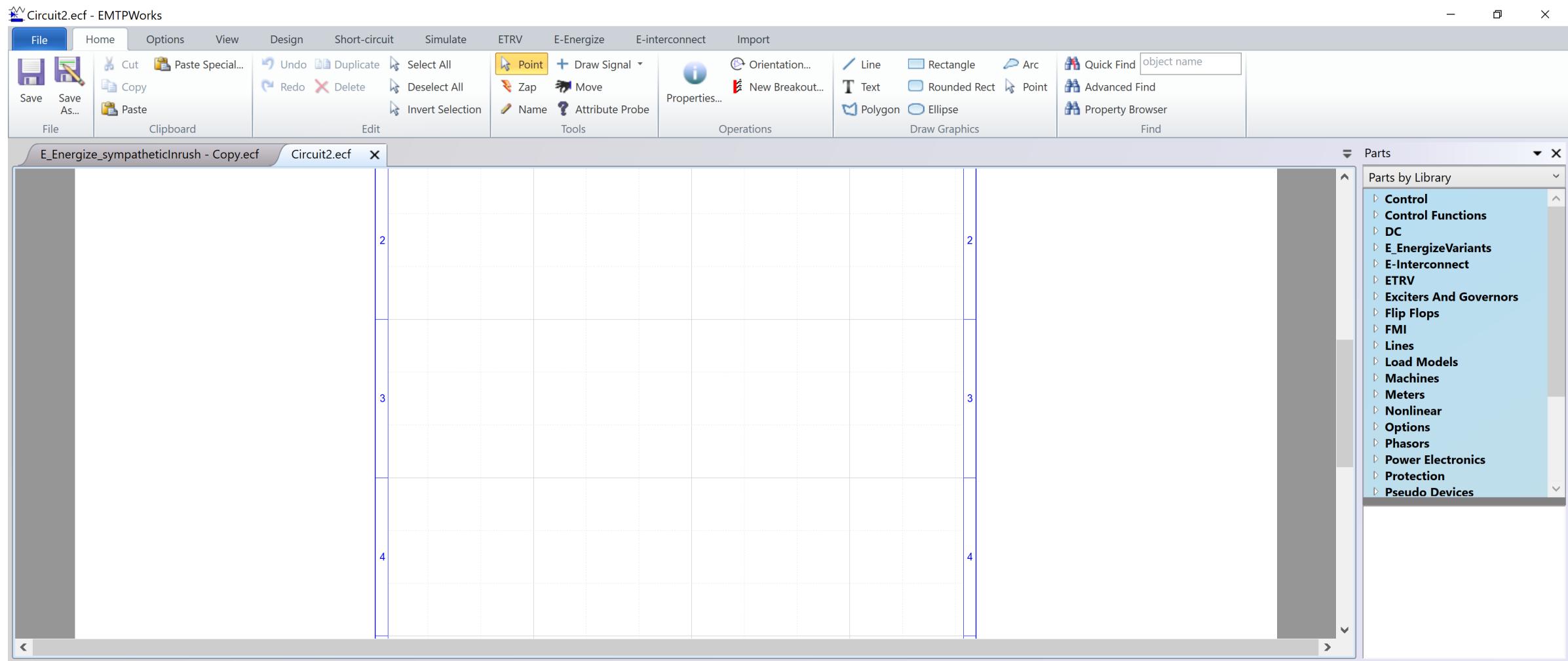
[info@emtp.com](mailto:info@emtp.com)

# Content

- I. Tool development capability of EMTP
- II. Automatic model parameter determination
- III. The IBR data fit tool
- IV. Case study
- V. Conclusions

## I. Tool development capability of EMTP

# I. Tool development capability of EMTP



# I. Tool development capability of EMTP

C:\Willy\_Nzale\PGSTech\Henry\Project\_IBR\_Data\_Fit\Mes\_tests\testcircuit9\L\_Benchmark\_PV\_example.ecf - EMTPWorks

File Home Options View Design Short-circuit Simulate ETRV E-Energize E-interconnect IBR\_data\_fit Parametric Import

Open IBR\_data\_fit interface  
Open IBR\_data\_fit documentation  
Open IBR\_data\_fit

IBR\_data\_fit panel: L\_Benchmark\_PV\_example.ecf

Optimisation definition

Select the optimisation method PSO

Set parameters for Particle Swarm Optimisation (PSO) method

Number of particles 5  
Inertia coefficient (w) 1  
Inertia Weight Damping Ratio (kappa) 0.99  
Personal acceleration coefficient (c1) 2.0  
Social acceleration coefficient (c2) 2.0  
Maximum number of iterations 10  
Cost function type Sum Absolute Error  
Convergence tolerance 1 %

Save data

IBR data fit version: 1.0

IBR\_data\_fit panel: L\_Benchmark\_PV\_example.ecf

A B C D

-control K<sub>p</sub> 1.2  
-control K<sub>i</sub> 7.4

PVPark1  
W/m<sup>2</sup>  
AVM  
75.015MVA  
120kV  
C-control

IBR Data Fit  
PVPark\_d.dwi  
MPLOT

V\_PPC VM ?v  
I\_park A  
PQ 3-phase PQm1  
60Hz 2s  
fault1 abcg 2|2.1|0  
BUS1 v1:1.00/  
v1

IBR\_data\_fit\_Meas\_Data  
Reference signal  
IBR\_data\_fit\_Meas\_data

Parts

Parts by Library

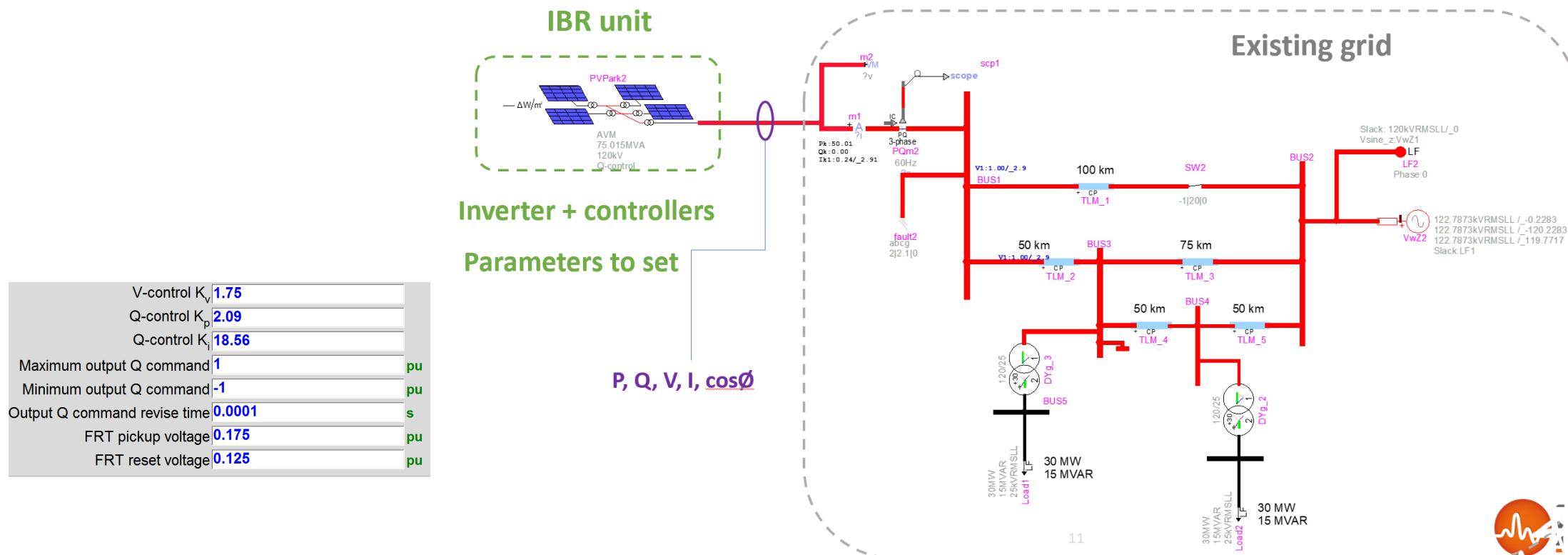
- Control
- Control Functions
- DC
- E-Interconnect
- ETRV
- Exciters And Governors
- Flip Flops
- FMI
- IBR\_data\_fit
- Lines
- Load Models
- Machines
- Meters
- Nonlinear
- Options
- Parameter Sweep
- Phasors
- Power Electronics
- Protection
- Pseudo Devices
- Renewables
- RLC Branches
- SimulinkDLL
- Sources
- Switches
- Symbols
- Transformations
- Transformers

## II. Automatic model parameter determination

## II. Automatic model parameter determination

Context:

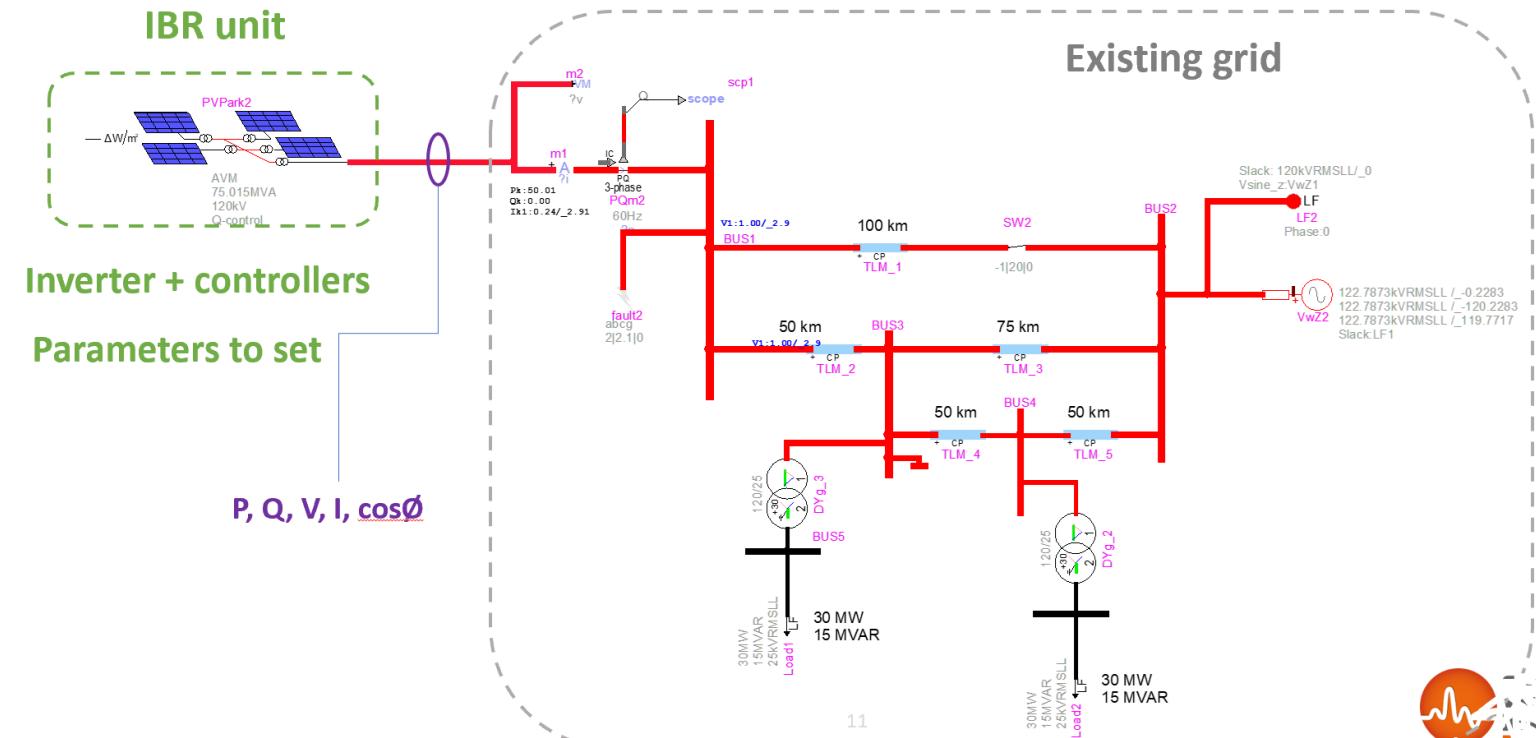
- Let's consider an IBR plant connected to a grid. The plant and the grid are modeled in EMTP, but we do not know the settings of some parameters in the IBR model (controller gains, time constants, etc...).
- On-site measurements of the physical plant response in some operating conditions are available.
- A tool has been created in EMTP to automatically determine some unknown IBR plant parameters values by fitting the IBR model response to on-site measurements.



## II. Automatic model parameter determination

Context:

- The tool performs the following tasks automatically:
  - Assign values to some parameters of the plant model,
  - simulate the model and get the plant response,
  - compare the simulation results with expected waveforms (experimental measurements),
  - update/modify parameters values and redo the process until the perfect match is found.
- Parameters update process is performed using an artificial intelligence method known as **the particle swarm optimization (PSO) algorithm**.



## II. Automatic model parameter determination

An overview of the particle swarm optimization (PSO) algorithm:

- Several entities (particles) are searching for the best solution.
- It is an iterative process (instant1, instant2, instant3, etc...).
- At each iteration (instant), each particle has a **position**, a **velocity** and a **cost**.
- At each iteration (instant), each particle position is updated based on past knowledge.

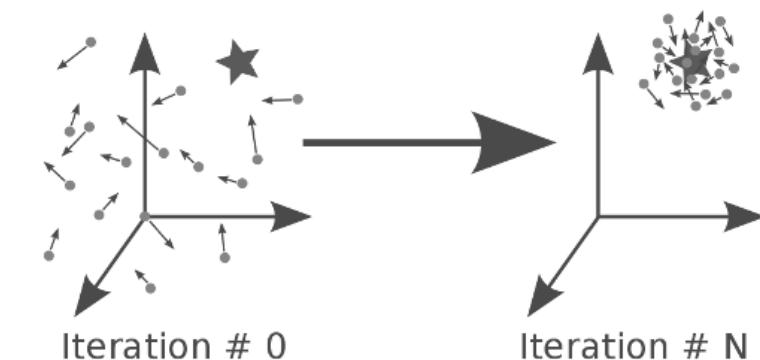
$$P_i^{t+1} = P_i^t + V_i^{t+1} \quad \text{Particle position update}$$

$$V_i^{t+1} = W \cdot V_i^t + c_1 U_1 (P_{b_1}^t - P_i^t) + c_2 U_2 (g_b^t - P_i^t) \quad \text{Velocity update}$$

**Inertia** : Makes the particle move in the same direction and with the same velocity.

**Personal Influence** : Improves the individual. Makes the particle return to a previous position, better than the current.

**Social Influence** : Makes the particle follow the best neighbors direction.



## II. Automatic model parameter determination

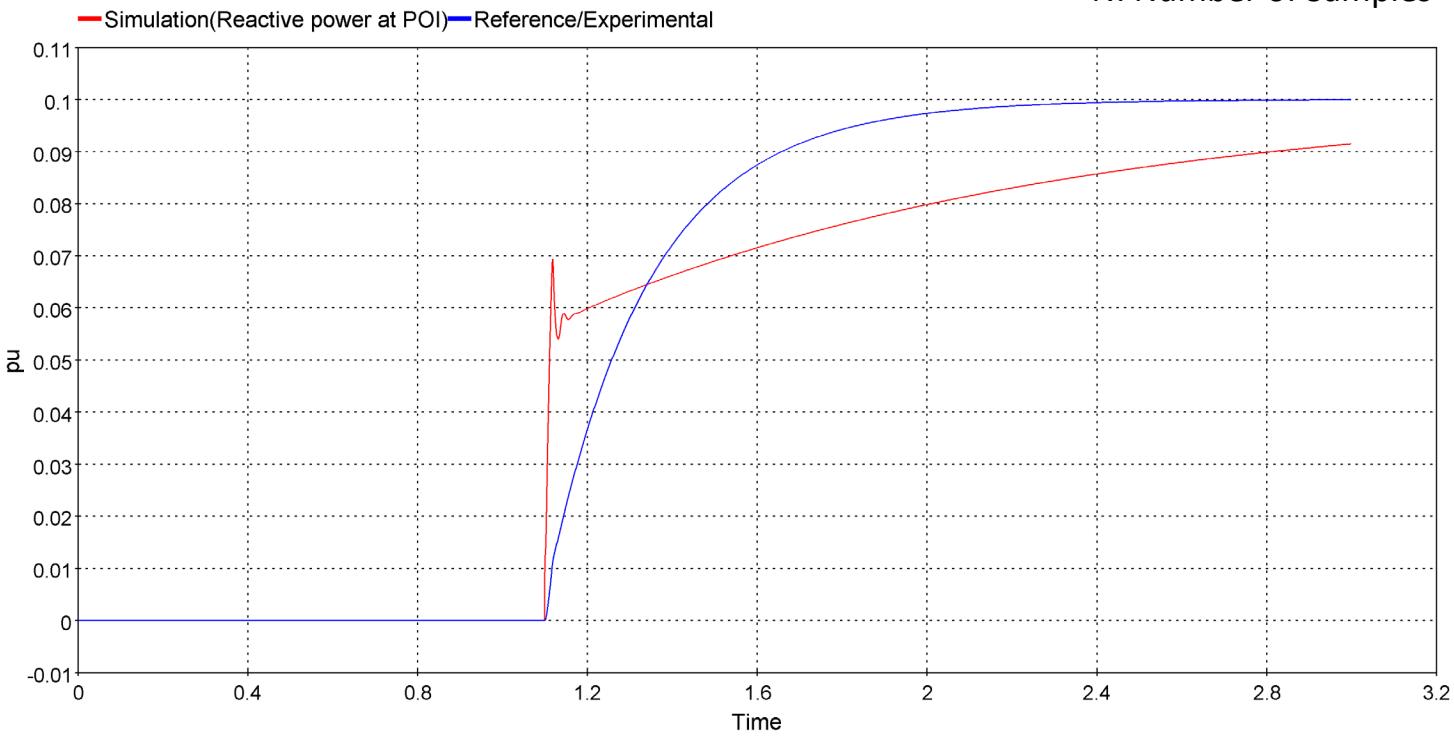
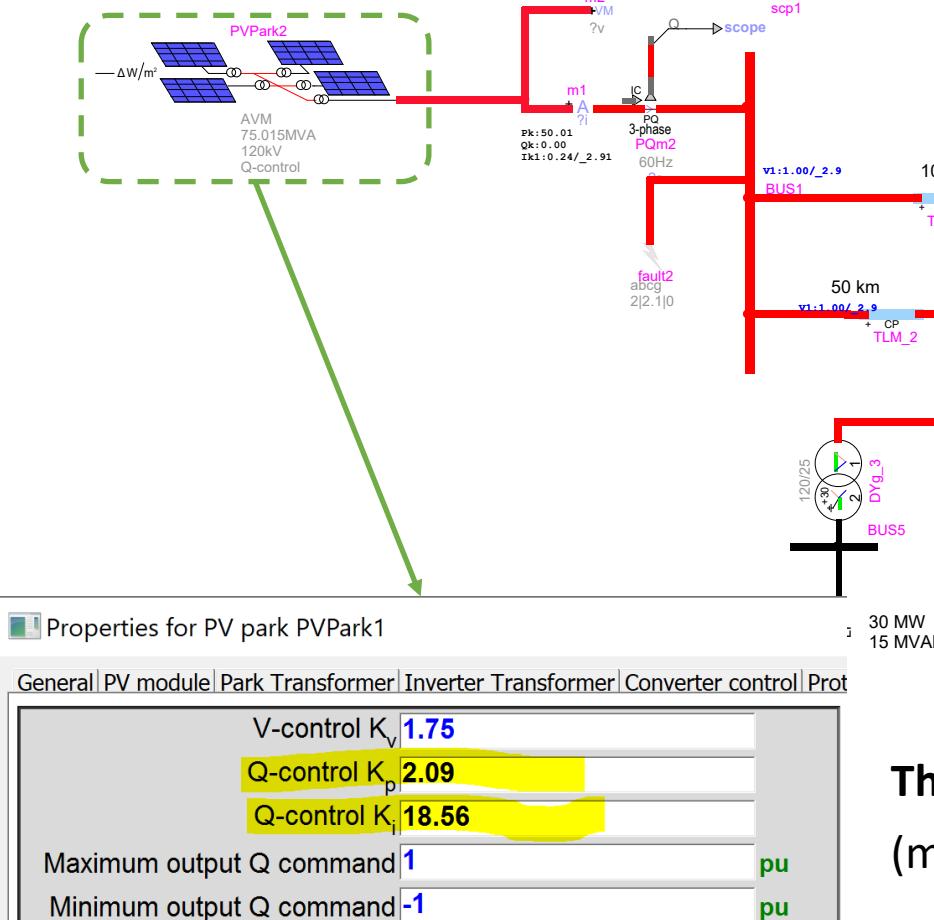
PSO applied to parameter determination Ex: estimating the PI controller gains of the reactive power control

The blue line represents the on-site measurement of the plant response to a Q-step change

Each value of  $K_p$  and  $K_i$  (a particle position) yields a specific plant response (red line)

$$Cost\ function: \sum_{i=0}^{i=N} |f_{Exp}(i) - f_{sim}(i)|$$

N: Number of Samples

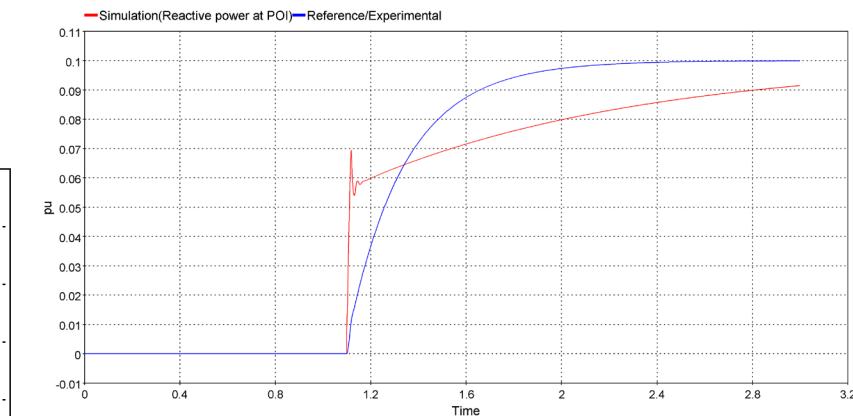
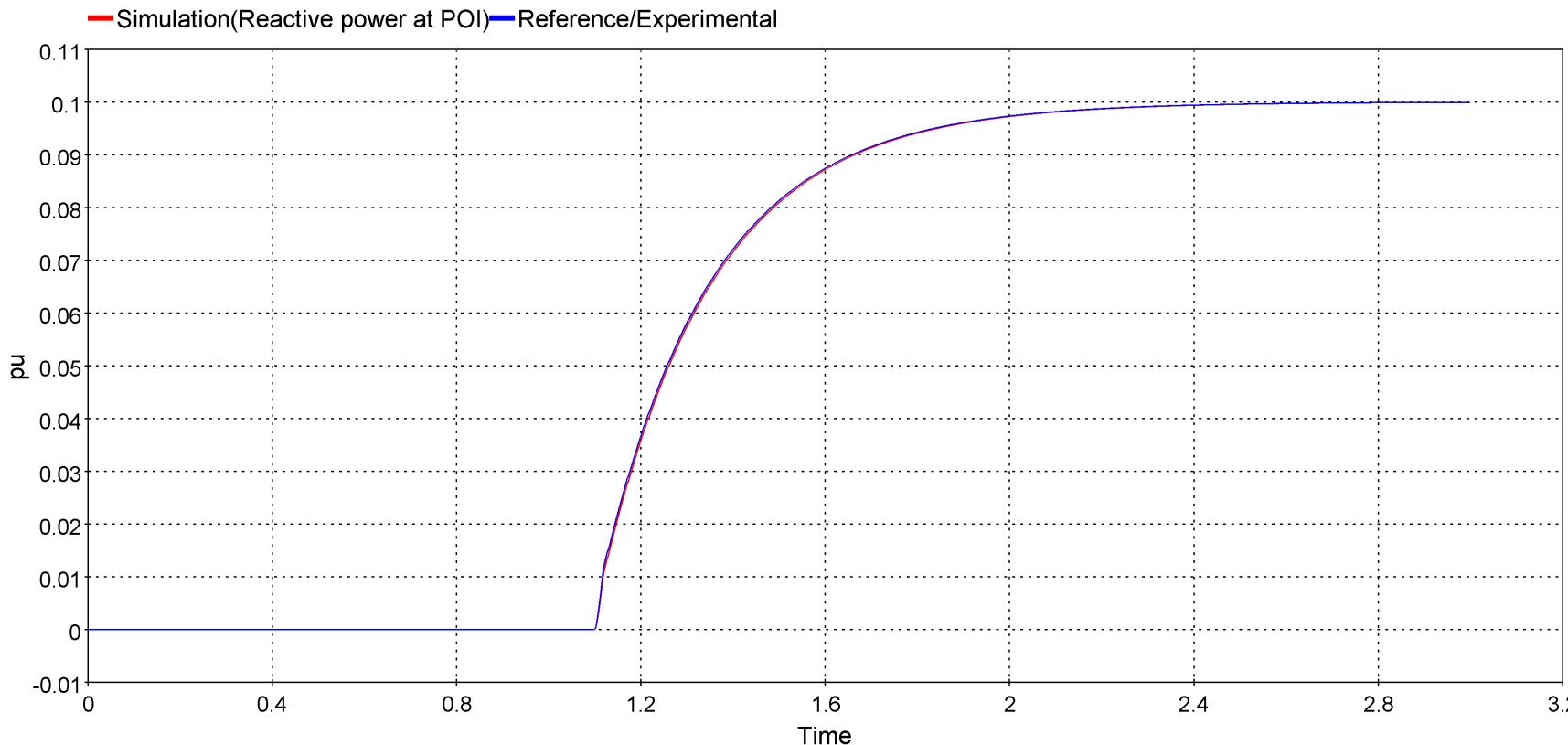


The goal is to find  $K_p$  and  $K_i$  so that the red line (simulated) fits the blue one (measurement).

## II. Automatic model parameter determination

PSO applied to parameter determination Ex: estimating the PI controller gains of the reactive power control

The PSO algorithm performs iterations and stops as soon as the cost associated to a particle gets lower than a predefined threshold. The corresponding particle value is taken as the final solution ( $K_p$  and  $K_i$ ).



	Final solution	Expected value
$K_p$	0.08	0.1
$K_I$	5.4	5.5

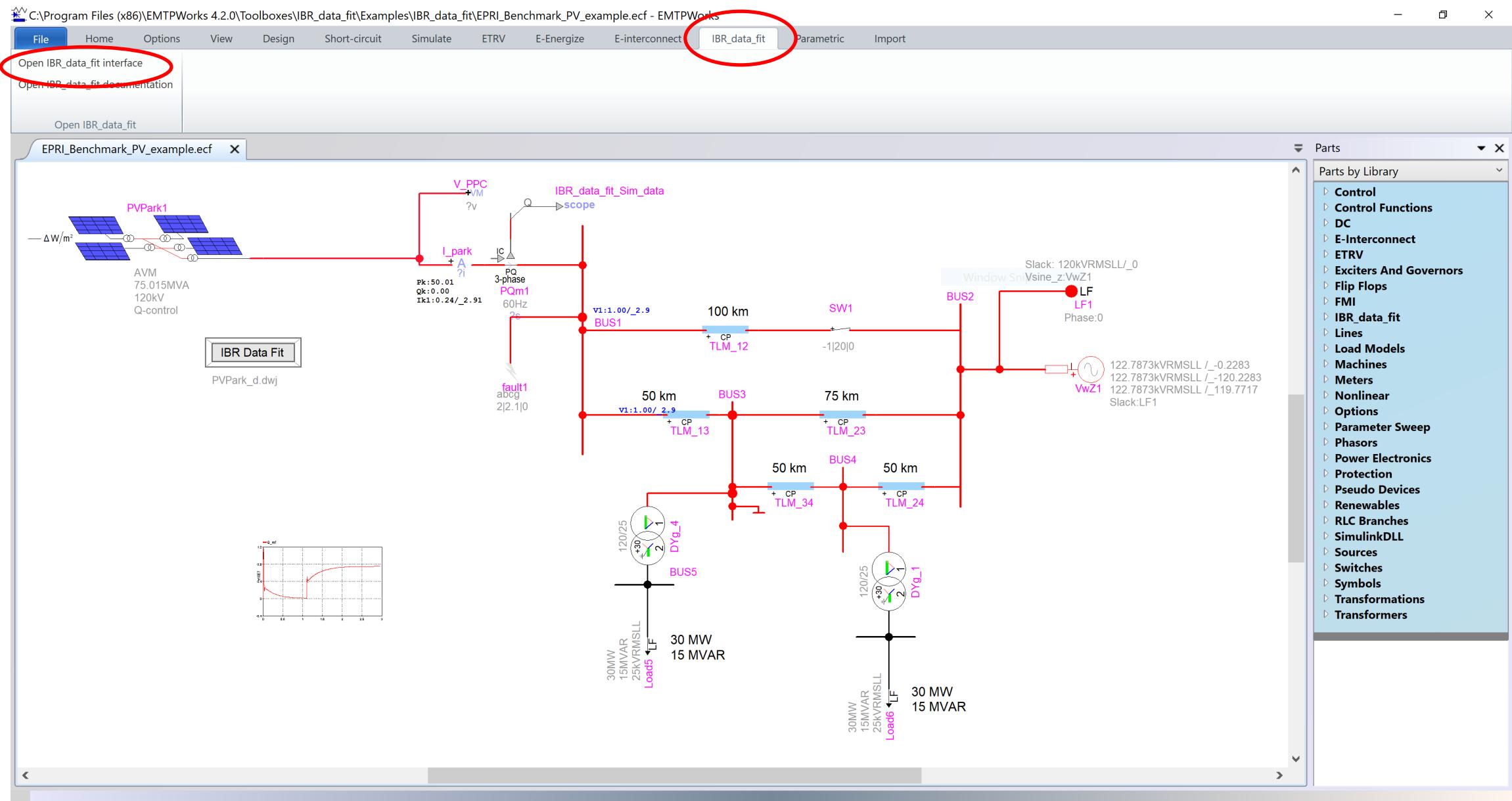
## II. Automatic model parameter determination

The automatic model parameter determination has been implemented in EMTP using the tool development capability presented at the beginning. The next section presents the tool itself.

### III. The IBR data fit tool

### III. The IBR data fit tool

#### Launch the tool



# III. The IBR data fit tool

## Step 1 : Select parameters to vary

C:\Program Files (x86)\EMTPWorks 4.2.0\Toolboxes\IBR\_data\_fit\Examples\IBR\_data\_fit\EPRI\_Benchmark\_PV\_example.ecf - EMTPWorks

File Home Options View Design Short-circuit Simulate ETRV E-Energize E-interconnect IBR\_data\_fit Parametric Import

Open IBR\_data\_fit interface  
Open IBR\_data\_fit documentation  
Open IBR\_data\_fit

IBR\_data\_fit panel: EPRI\_Benchmark\_PV\_example.ecf

Steps Step 1: Select parameters

IBR device parameters selection

Select the parameters to optimise and set the boundaries

Selected device name: PVPark1

Parameters	Default value	Min value	Max value
Ngen	45		
Freq	60		
Vgrid_kVRMSLL	34.5		
Vpoi_kVRMSLL	120		
Vgen_kVRMSLL	0.575		
Vdc_kv	1.264		
includeZigZagTransfo	1		
ZigZag_R0_ohm	0.1265		
ZigZag_L0_H	0.3831e-3		
Sgen	1.667		
Qfilt	75		
Rchoke	0.005		
Lchoke	0.15		
includeCollGrid	1		
R_Coll_Grid_Ohm	0.1265		
L_Coll_Grid_H	0.3831e-3		
C_Coll_Grid_F	7e-6		
Ngen_in_service	30		
QC_select	1		
Qpoi_pu	0		
pf_poi	1		
Vpoi_pu	1		
adjust_Qpoi_pu_withLF			
PC_select	1		
Pref_poi	1		

EPRI\_Benchmark\_PV\_example.ecf

The diagram illustrates a power system model. A PV park (labeled PVPark1) is connected to a bus. The bus is connected to a three-phase PQ source (PQm1). From the bus, a 100 km line leads to a load labeled TLM\_12. Another 50 km line leads to a bus labeled BUS3, which is connected to a 75 MVA load. From BUS3, a 50 km line leads to a bus labeled BUS5, which is connected to a 30 MW, 15 MVAR load. A fault point labeled 'fault1' is shown on the line between BUS1 and BUS3. A digital meter (Digi4) is connected to the line between BUS1 and BUS3. A current transformer (CT) is also present on the line. The system includes various protection and control components like a circuit breaker (CB), a voltage controller (V-PPC), and a scope. The entire model is contained within a frame labeled 'IBR Data Fit'.

Parts by Library

- Control
- Control Functions
- DC
- E-Interconnect
- ETRV
- Exciters And Governors
- Flip Flops
- FMI
- IBR\_data\_fit
- Lines
- Load Models
- Machines
- Meters
- Nonlinear
- Options
- Parameter Sweep
- Phasors
- Power Electronics
- Protection
- Pseudo Devices
- Renewables
- RLC Branches
- SimulinkDLL
- Sources
- Switches
- Symbols
- Transformations
- Transformers

ITP

### III. The IBR data fit tool

#### Step 2 : Simulation data definition

Steps
Step 1: Select parameters
<a href="#">Step 2: Simulation data definition</a>

**Simulation data definition**

**Reference and observed signals scopes names**  
Reference data signal scope: "IBR\_data\_fit\_Meas\_data"  
Observed data signal scope: "IBR\_data\_fit\_Sim\_data"

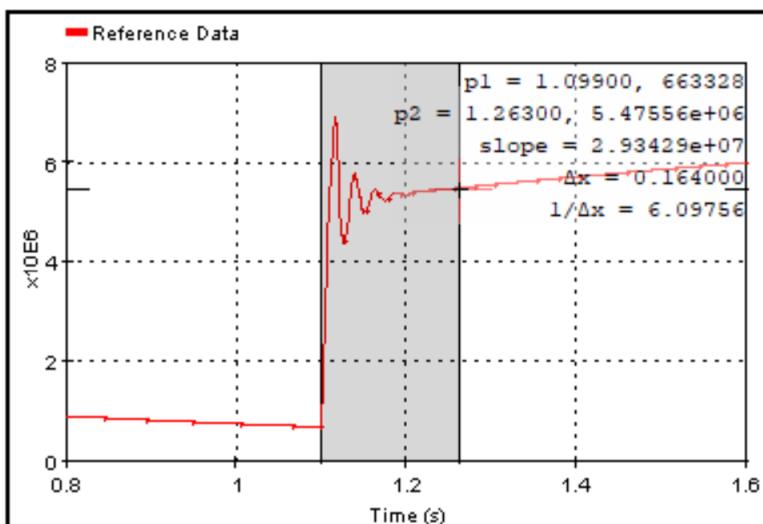
**Define observation interval**

Observation start time  s  
Observation stop time  s

**Save data**

On-site measurements data (reference) are saved in a .dat file.

A scope is added to the EMTP circuit with the name IBR\_data\_fit\_Sim\_data at the same location where on-site measurement had been performed.



### III. The IBR data fit tool

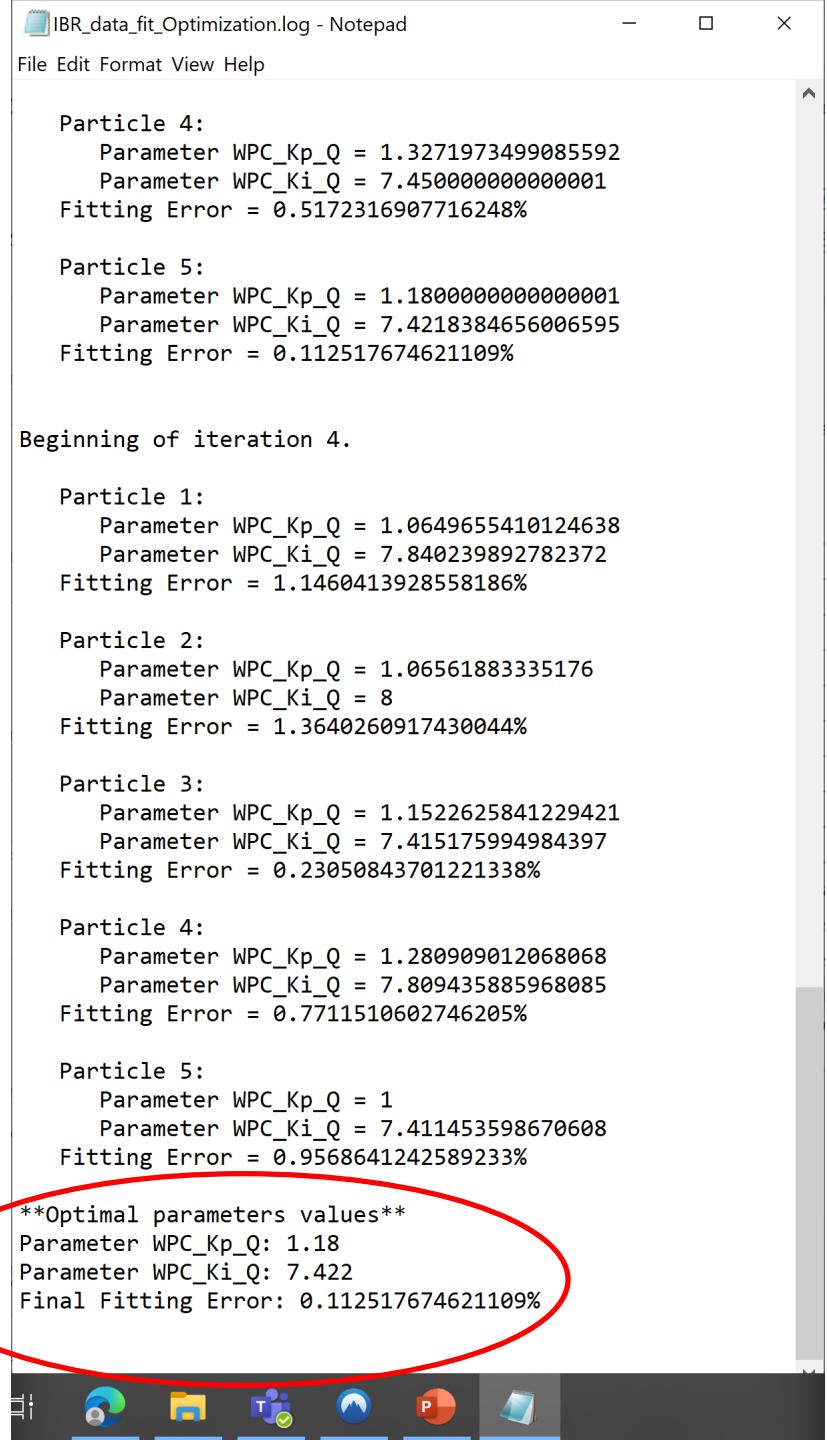
#### Step 3: Optimization definition

Steps	Optimisation definition
Step 1: Select parameters	Select the optimisation method <input type="text" value="PSO"/>
Step 2: Simulation data definition	
<a href="#">Step 3: Optimisation definition</a>	<b>Set parameters for Particle Swarm Optimisation (PSO) method</b> Number of particles <input type="text" value="5"/> Inertia coefficient (w) <input type="text" value="1"/> Inertia Weight Damping Ratio (kappa) <input type="text" value="0.99"/> Personal acceleration coefficient (c1) <input type="text" value="2.0"/> Social acceleration coefficient (c2) <input type="text" value="2.0"/> Maximum number of iterations <input type="text" value="10"/> Cost function type <input type="text" value="Sum Absolute Error"/> Convergence tolerance <input type="text" value="1"/> %  <input type="button" value="Save data"/>

### III. The IBR data fit tool

#### Final step: get optimization results

Optimization results are available in a text file located inside the \_pj folder with the name IBR\_data\_fit\_Optimization.log



The screenshot shows a Windows Notepad window titled "IBR\_data\_fit\_Optimization.log - Notepad". The window contains the following text:

```
File Edit Format View Help

Particle 4:
Parameter WPC_Kp_Q = 1.3271973499085592
Parameter WPC_Ki_Q = 7.4500000000000001
Fitting Error = 0.5172316907716248%

Particle 5:
Parameter WPC_Kp_Q = 1.1800000000000001
Parameter WPC_Ki_Q = 7.4218384656006595
Fitting Error = 0.112517674621109%

Beginning of iteration 4.

Particle 1:
Parameter WPC_Kp_Q = 1.0649655410124638
Parameter WPC_Ki_Q = 7.840239892782372
Fitting Error = 1.1460413928558186%

Particle 2:
Parameter WPC_Kp_Q = 1.06561883335176
Parameter WPC_Ki_Q = 8
Fitting Error = 1.3640260917430044%

Particle 3:
Parameter WPC_Kp_Q = 1.1522625841229421
Parameter WPC_Ki_Q = 7.415175994984397
Fitting Error = 0.23050843701221338%

Particle 4:
Parameter WPC_Kp_Q = 1.280909012068068
Parameter WPC_Ki_Q = 7.809435885968085
Fitting Error = 0.7711510602746205%

Particle 5:
Parameter WPC_Kp_Q = 1
Parameter WPC_Ki_Q = 7.411453598670608
Fitting Error = 0.9568641242589233%

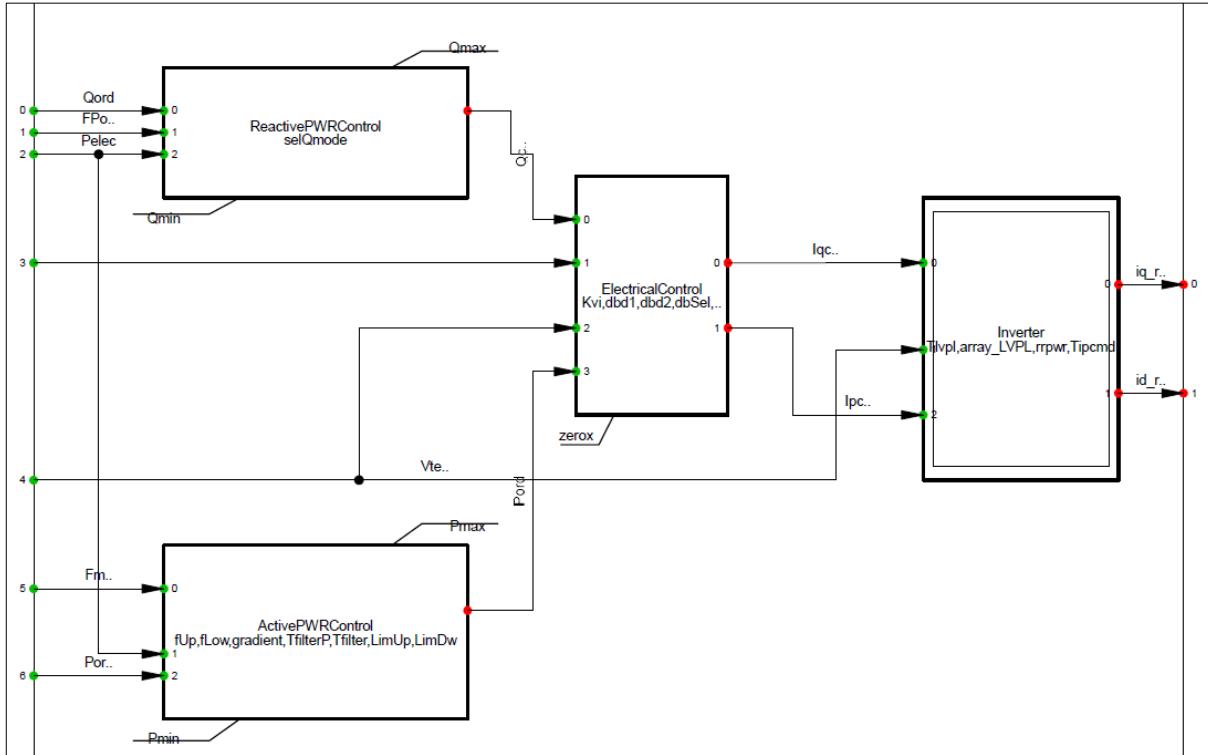
**Optimal parameters values**
Parameter WPC_Kp_Q: 1.18
Parameter WPC_Ki_Q: 7.422
Final Fitting Error: 0.112517674621109%
```

A red oval highlights the last three lines of the output, which represent the optimal parameter values and the final fitting error.

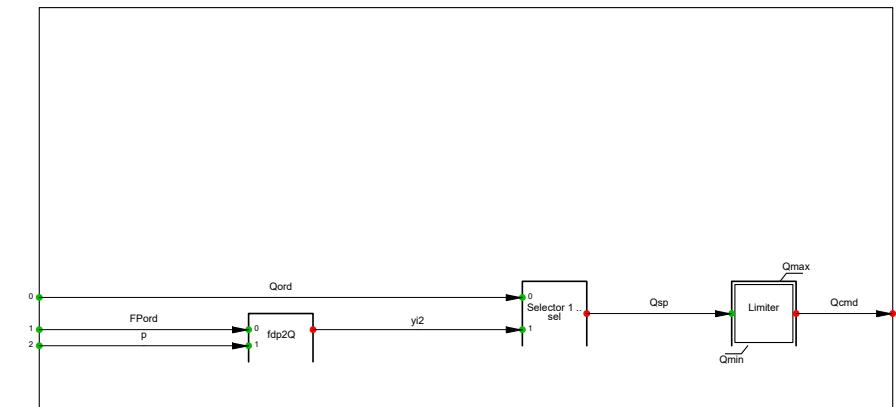
## IV. Case study

## IV. Case study

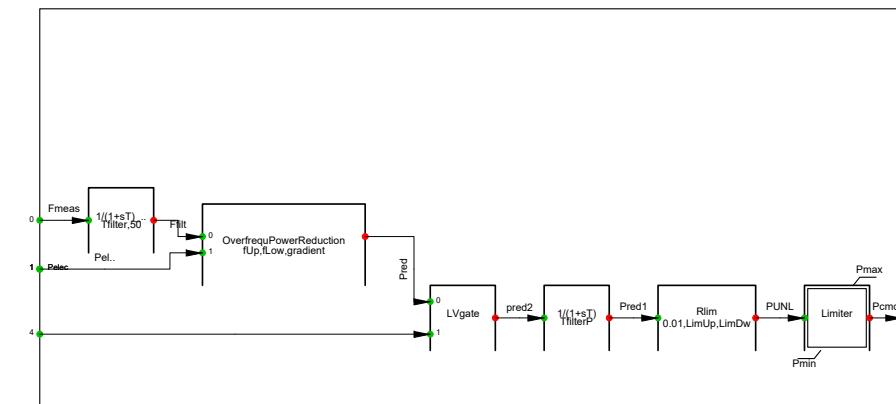
### Calibration of a NON-WECC Model (extremely simplified)



ReactivePWRControl:



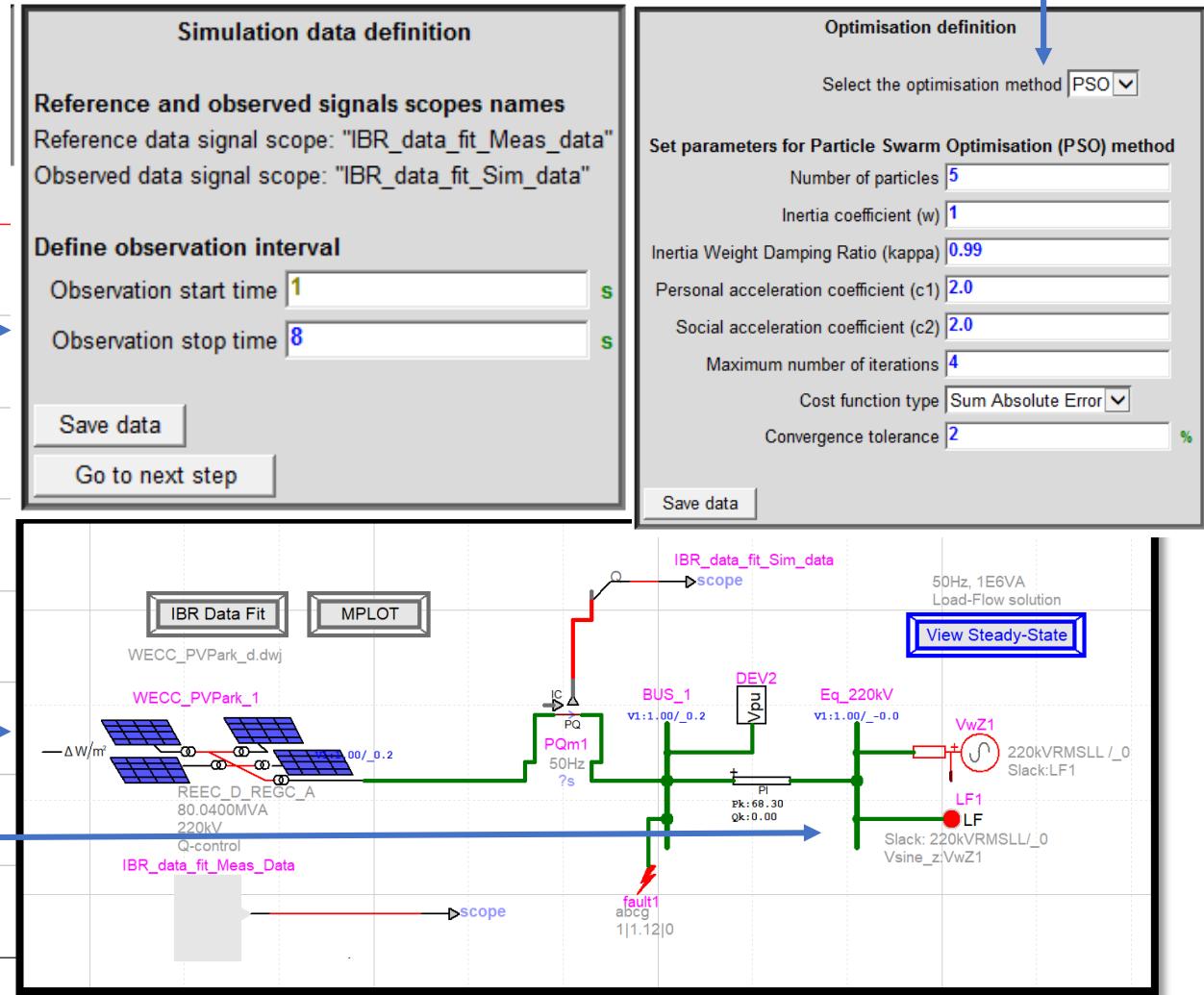
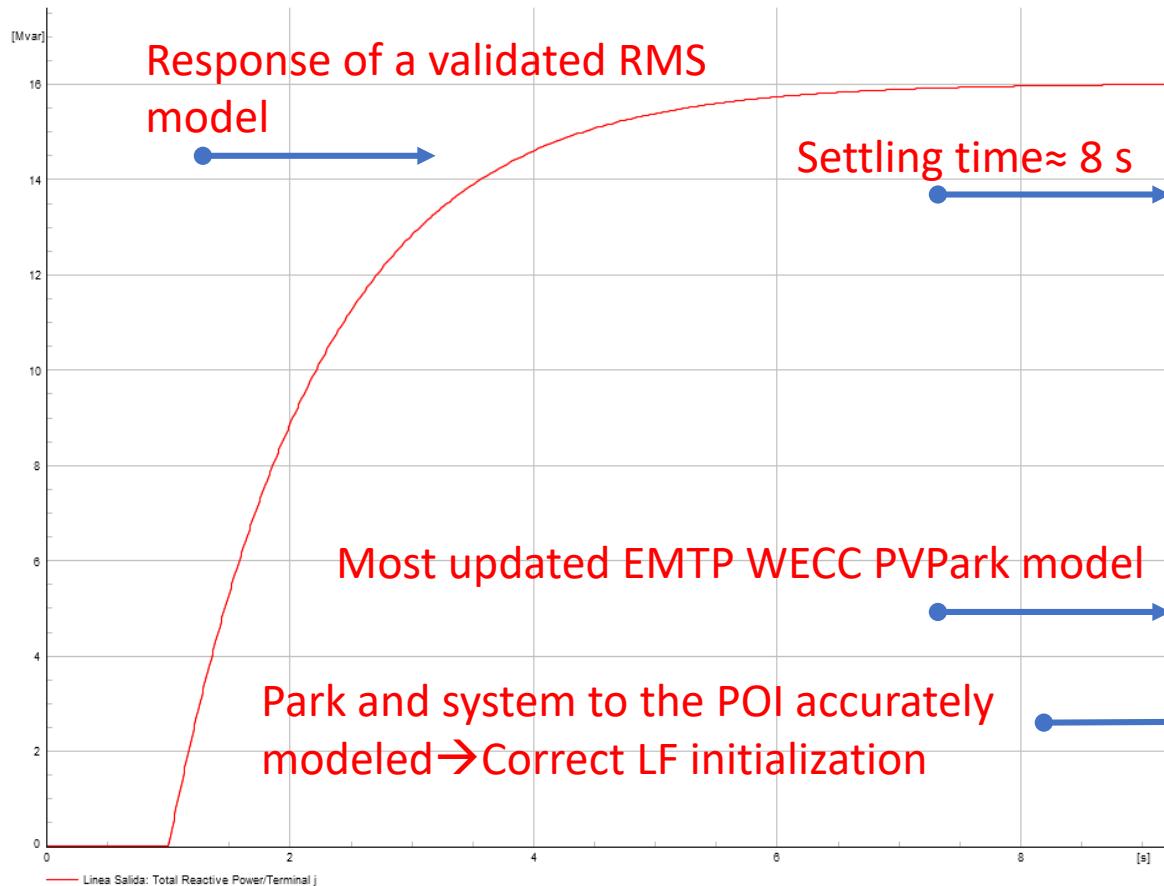
ActivePWRControl:



- RMS Model validated on site (Step changes, not faults)
- NON-WECC model → More challenging fitting
- Simple control blocks → difficult to translate to a more complex model
- NON WECC model → Almost equivalent to a Black Box model for practical effects

## IV. Case study

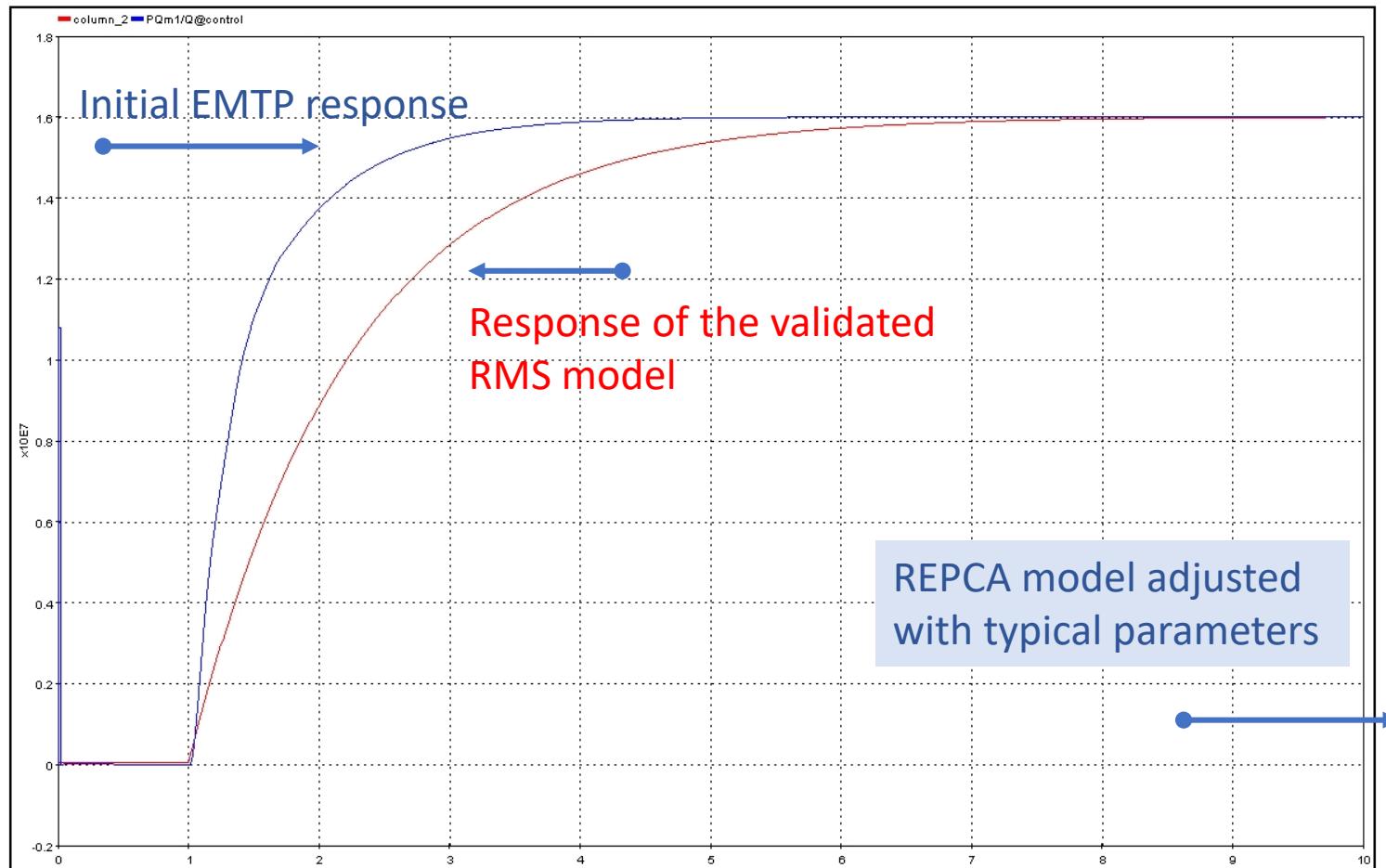
### PPC Parameter Calibration with a Q Step Change of 20% of the Nominal Power of the PV Park



## IV. Case study

### CASE 1: Default values for the REPC WECC\_PV\_Park

#### Model



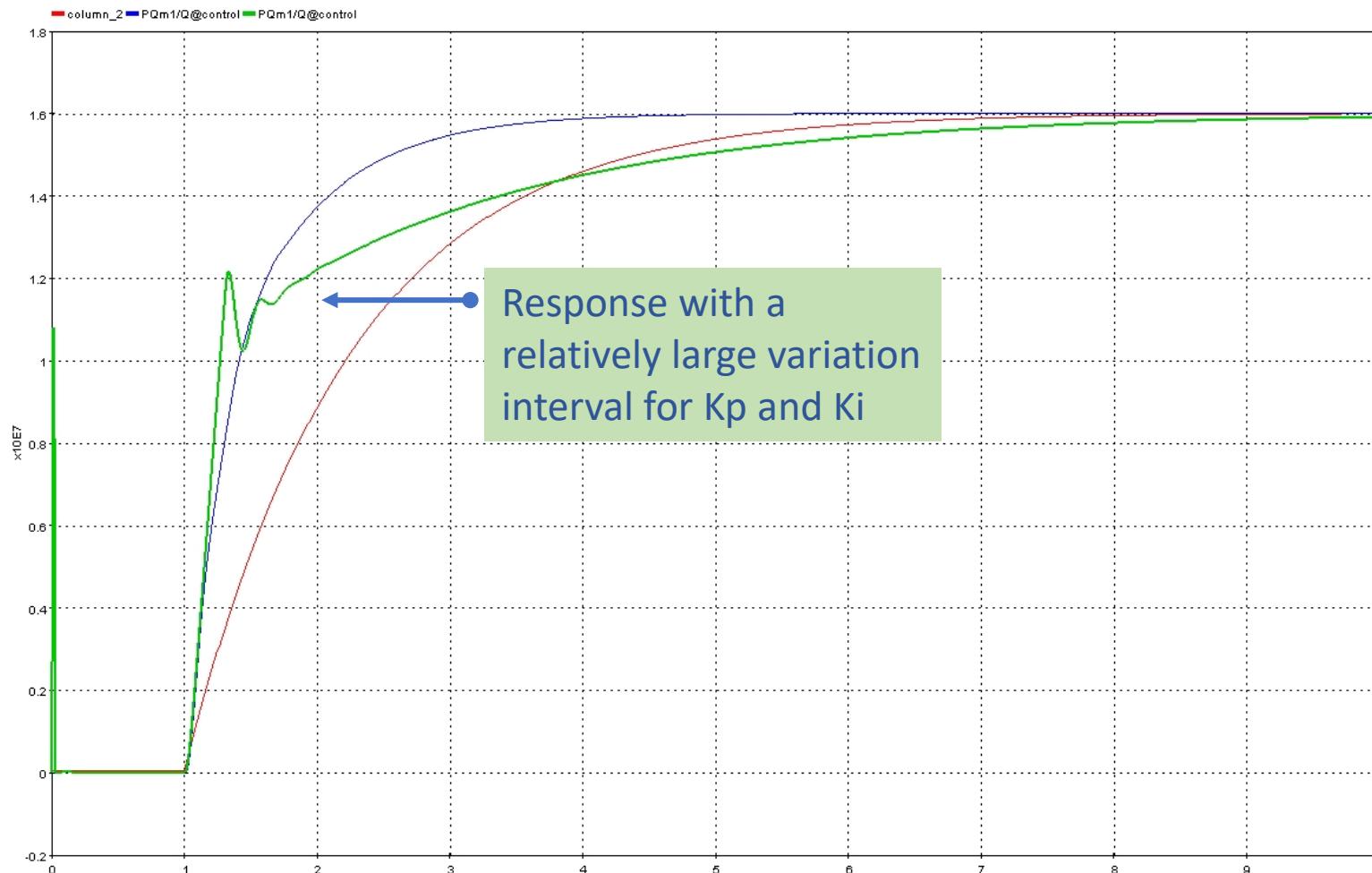
Q/V control - REPCA model

Voltage droop control	droop control
Time constant Tfltr	0.02
Proportional gain Kp	0.5
Integral gain Ki	2.5
Lead time constant Tf <sub>l</sub>	0
Lag time constant Tf <sub>v</sub>	0.05
Voltage Vfrz	0.7
Compensation resistance R <sub>c</sub>	0
Compensation reactance X <sub>c</sub>	0
Compensation gain K <sub>c</sub>	0.02
Upper limit on deadband emax	0.1
Lower limit on deadband emin	-0.1
Lower threshold for deadband dbd <sub>1</sub>	0
Upper threshold for deadband dbd <sub>2</sub>	0
Upper limit Qmax	0.436
Lower limit Qmax	-0.436
Output P and Q command revise time	0.001

Parameter Kp<sub>REPC</sub> = 0.5  
Parameter Ki<sub>REPC</sub> = 2.5  
Fitting Error = 10.75%

## IV. Case study

### CASE 2: PSO Optimisation of Parameter Kp\_REPC and Ki\_REPC with variation intervals set to [0-20]



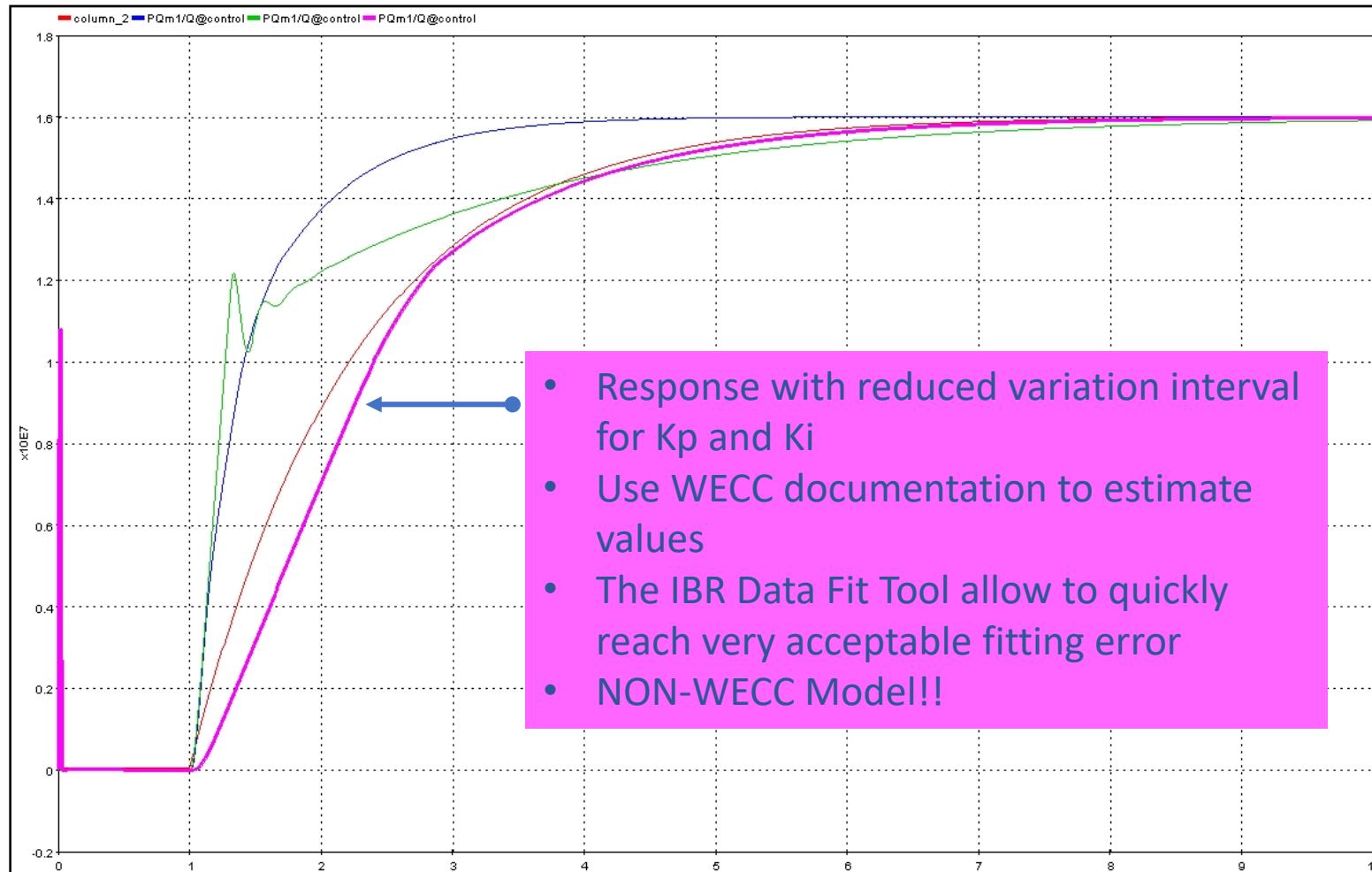
Q/V control - REPCA model	
Voltage droop control	droop control
Time constant Tf <sub>fltr</sub>	0.02
Proportional gain K <sub>p</sub>	1.40945
Integral gain K <sub>i</sub>	1.2812
Lead time constant T <sub>f1</sub>	0
Lag time constant T <sub>f2</sub>	0.05
Voltage V <sub>fref</sub>	0.7
Compensation resistance R <sub>c</sub>	0
Compensation reactance X <sub>c</sub>	0
Compensation gain K <sub>c</sub>	0.02
Upper limit on deadband e <sub>max</sub>	0.1
Lower limit on deadband e <sub>min</sub>	-0.1
Lower threshold for deadband dbd <sub>1</sub>	0
Upper threshold for deadband dbd <sub>2</sub>	0
Upper limit Q <sub>max</sub>	0.436
Lower limit Q <sub>max</sub>	-0.436
Output P and Q command revise time	0.001

\*\*Optimal parameters values\*\*  
 Parameter Kp\_REPC: 1.40945  
 Parameter Ki\_REPC: 1.2812  
 Final Fitting Error: 7.27%

## IV. Case study

### CASE 3: PSO Optimisation of Parameter Kp\_REPC and

### Ki\_REPC with variation intervals set to [0-5]



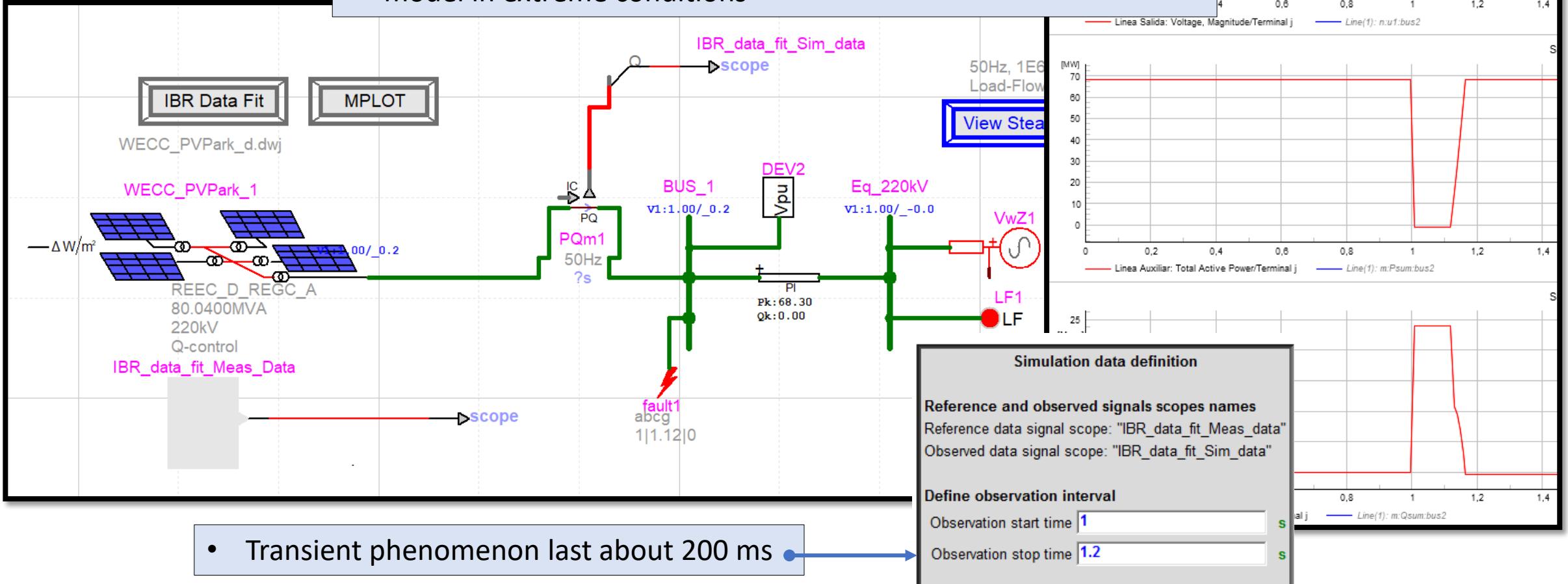
Q/V control - REPCA model	
Voltage droop control	droop control
Time constant Tfiltr	0.02
Proportional gain Kp	0
Integral gain Ki	0.9699999999999997
Lead time constant Tft	0
Lag time constant TfV	0.05
Voltage Vfrz	0.7
Compensation resistance Rc	0
Compensation reactance Xc	0
Compensation gain Kc	0.02
Upper limit on deadband emax	0.1
Lower limit on deadband emin	-0.1
Lower threshold for deadband dbd1	0
Upper threshold for deadband dbd2	0
Upper limit Qmax	0.436
Lower limit Qmax	-0.436
Output P and Q command revise time	0.001

\*\*Optimal parameters values\*\*  
Parameter Kp\_REPC: 0  
Parameter Ki\_REPC: 0.969  
Final Fitting Error: 1.5%

## IV. Case study

### Inverter and Electrical Control parameter calibration with a 3ph-g fault

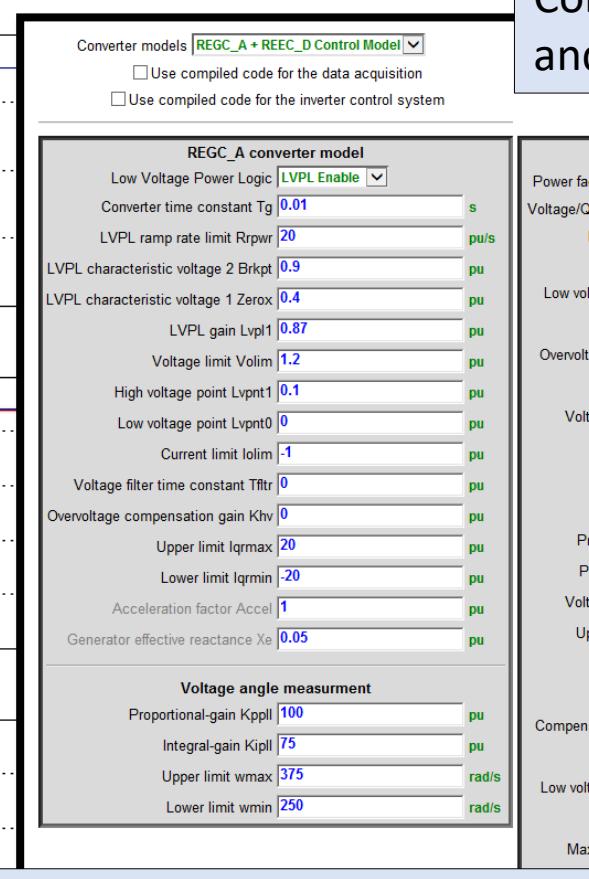
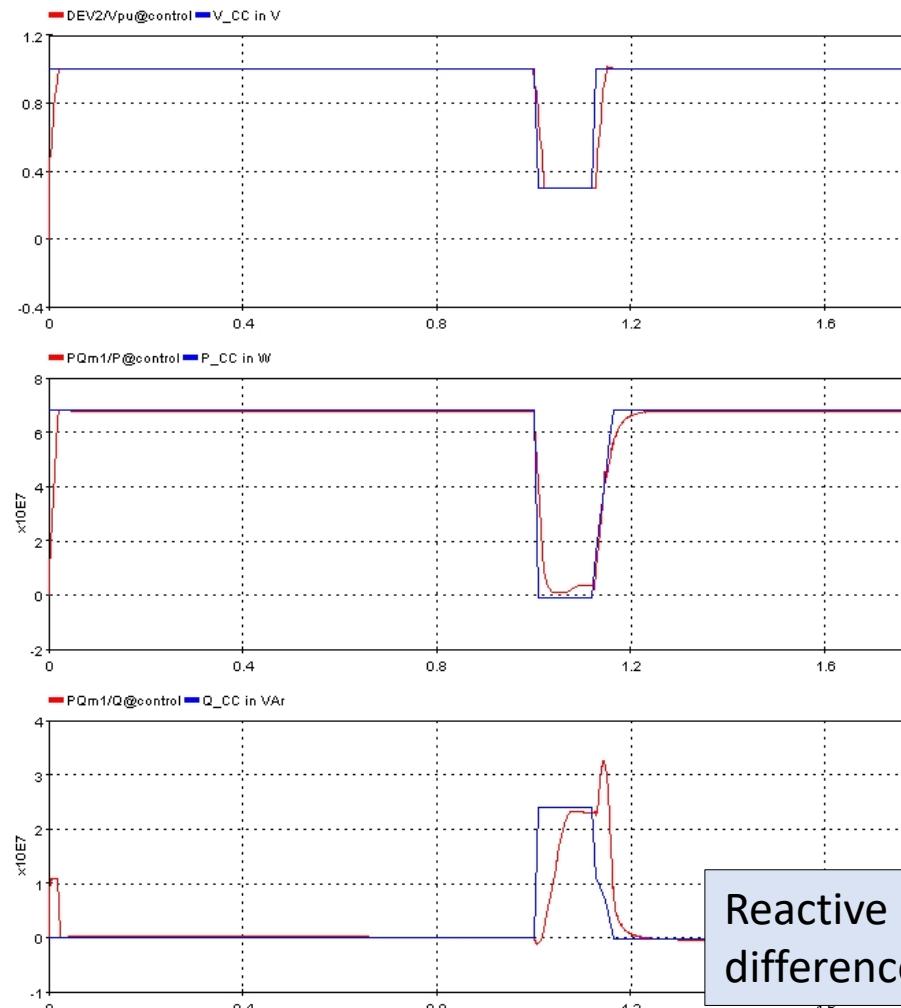
- 3Ph-g fault in the validated model at POI
- Fault impedance such to reach a deep voltage dip → Test the model in extreme conditions



## IV. Case study

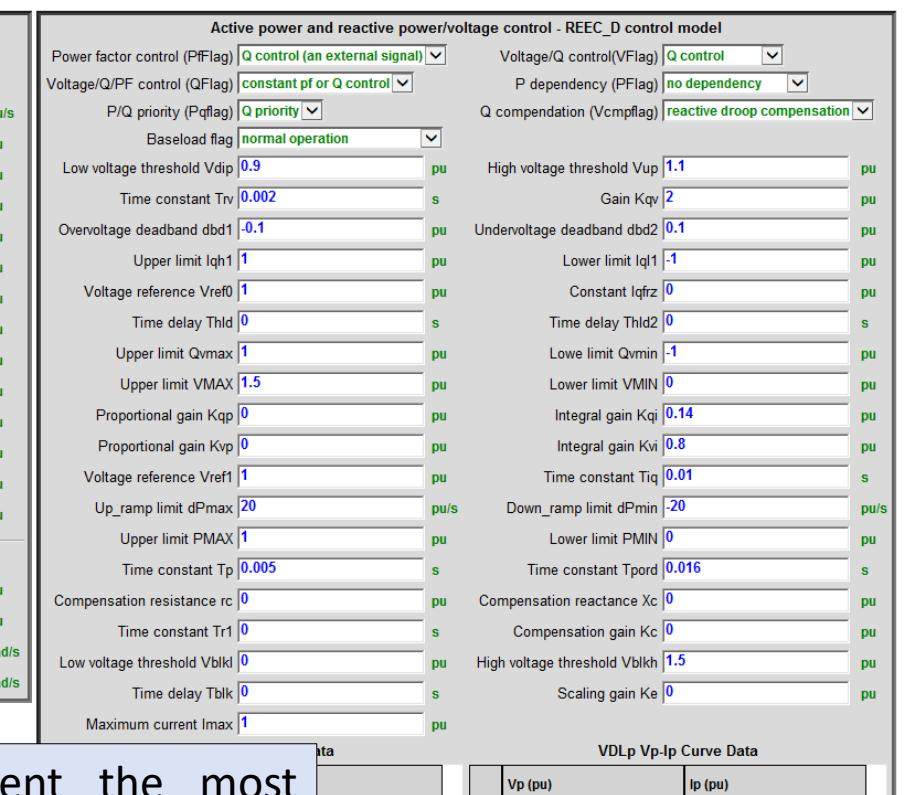
### CASE 1: Manually adjusted parameters of the WECC

#### REGC\_A+REEC\_D Control Model



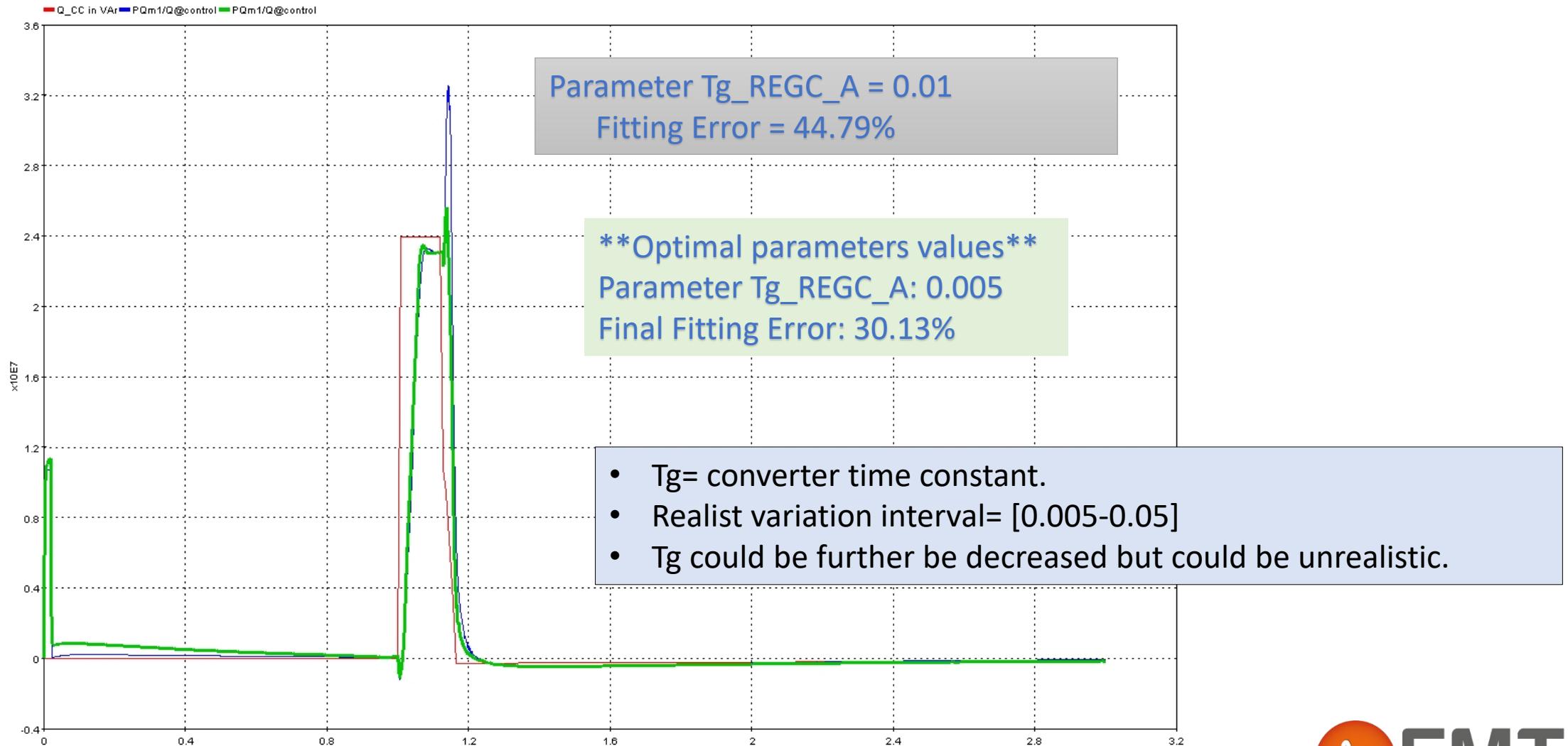
Reactive Power response present the most difference respect to the validated model

Converter model adjusted with typical values and manual simulations



## IV. Case study

CASE 2: PSO Optimisation of Parameter Tg\_REGC\_A with variation interval set to [0.005-0.05]

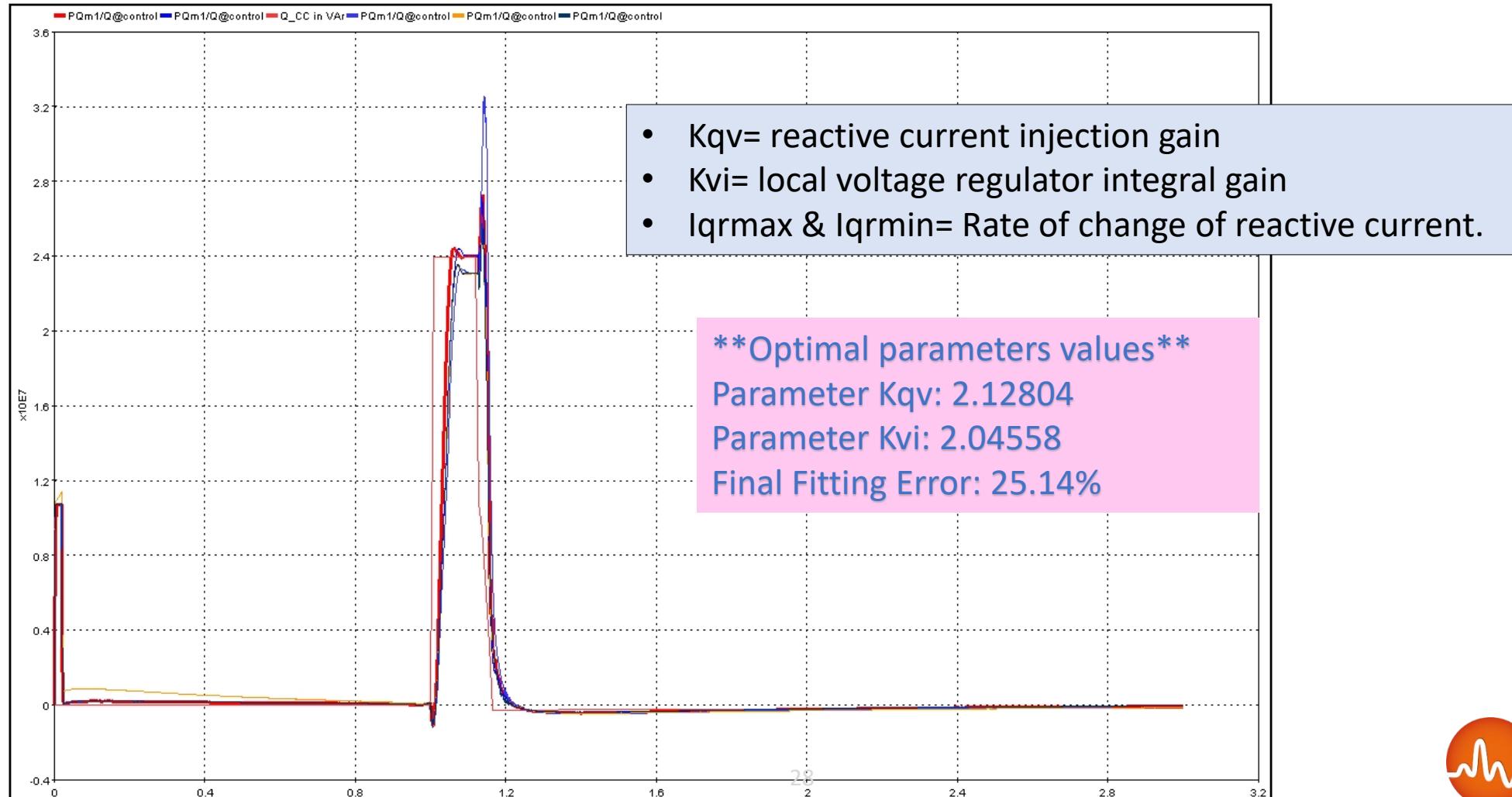


## IV. Case study

### CASE 3: PSO Optimisation of Parameters of the REEC\_D

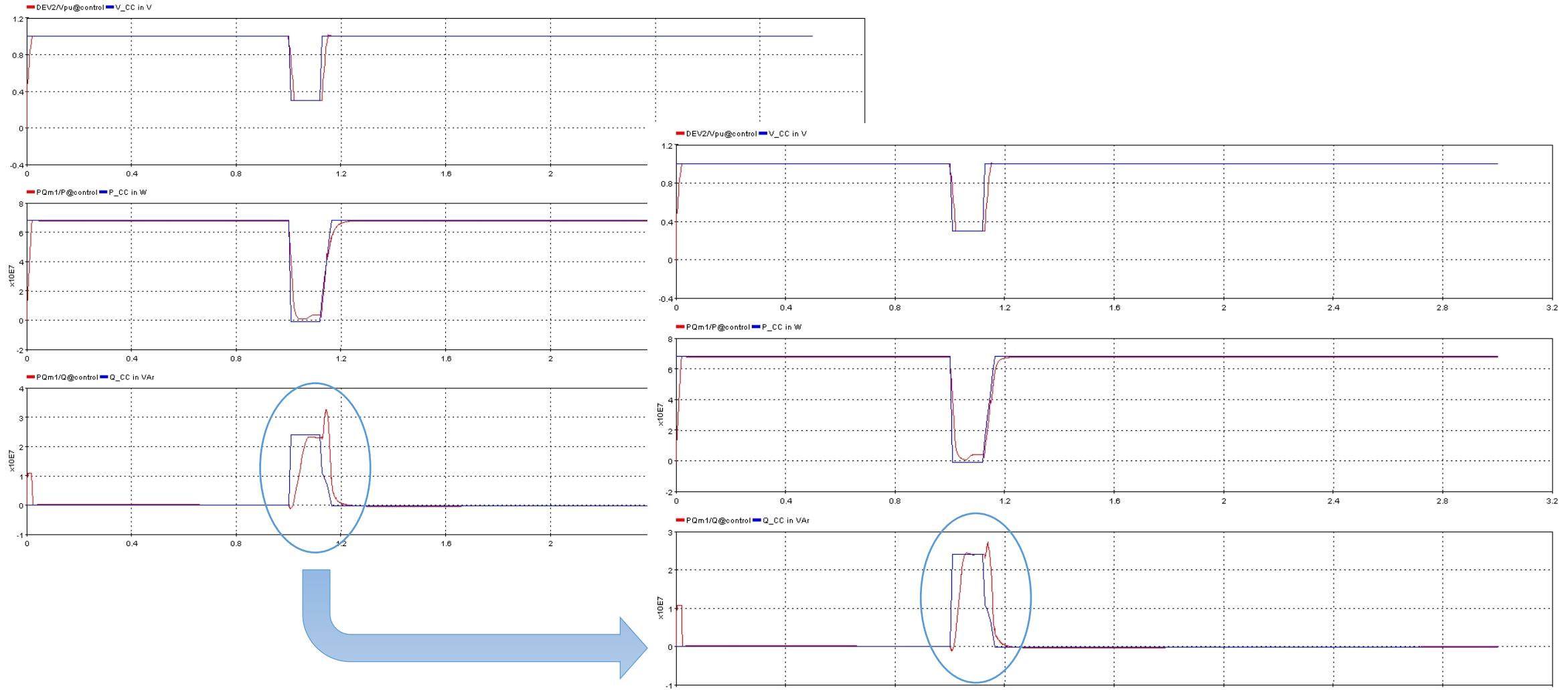
Kqv\_REEC\_D [0-5] (2); Kvi\_REEC\_D [0-5] (0.8)

(with Iqrmax and Iqrmin at maximum)



## IV. Case study

### Final Fitting for a 3ph-g fault



- IBR Data Fit Tool useful for the finer adjustment.

## V. Conclusions

## V. Conclusions

- The initial adjustment of the model parameters is extremely important.
- The IBR Data Fit Tool proved to be particularly effective when the ranges of variation of the parameters to be calibrated were shortened.
- The IBR Data Fit Tool is very effective for finer adjustments.
- The adjustment of non-WECC models (simplified) was more difficult and the IBR Data Fit Tool was particularly useful in this case.

## References

B. Poudel, B. Bhandari, E. Amiri, P. Rastgoufard, T. E. Field and R. A. McCanne, "*Interconnection Study and Optimization of Grid Connected Photovoltaic System Using Electromagnetic Transient Program*," 2021 IEEE Kansas Power and Energy Conference (KPEC), 2021, pp. 1-6, doi: 10.1109/KPEC51835.2021.9446233.