## APPENDIX: TRAINING AND APPLICATION OF MODULAR LINEAR POWER FLOW MODEL

The basic idea of the modular linear power flow model is first to split the wide operating conditions into several regions based on the k-means method, and then the best-fit linear power flow models with minimized errors are segmented trained for each region. A classifier is trained based on the divided sub-regions, which aims to select the applicable segmented linear model for each scenario regarding the scenario-based method. The applicable segmented linear models for different scenarios are dynamically selected and may be different. The individual sensitivity factors  $\Gamma$  for various scenarios are calculated based on the Jacobian matrix composed of different applicable segmented models. It suggests that the sensitivity factors  $\Gamma$  may be different for different scenarios. The process is illustrated in more detail below.

Since the power injections of numerous scenarios directly affect the operation boundary margins reserved for uncertainty and the chance constraint violations, operating conditions in this paper are considered to consist of power injections, which can be shown as follows:

$$h_{OC}\left(\mathbf{u} + \mathbf{P}^{G,fp} + \mathbf{P}^{W,fp} - \mathbf{P}^{D,fp}, \ \gamma \mathbf{u} + \mathbf{Q}^{G,fp} + \mathbf{Q}^{W,fp} - \mathbf{Q}^{D,fp}\right), (1)$$

where  $\gamma$  is the constant power ratio related to the power factor.

Various segmented linear models use different state variable functions for Taylor expansion. This paper uses the linear power flow model with individual independent variable function  $\varphi_i(v_i) = v_i^{k_i}$  for different buses and unified function  $\varphi(\theta) = \theta_{ij}$ , which is shown as follows:

$$PF_{ij}^{L} = g_{ij} \left( \frac{v_{i}^{k_{i}} - 1}{k_{i}} - \frac{v_{j}^{k_{j}} - 1}{k_{j}} \right) - b_{ij}\theta_{ij},$$
(2)

$$QF_{ij}^{L} = -b_{ij} \left( \frac{v_i^{k_i} - 1}{k_i} - \frac{v_j^{k_j} - 1}{k_j} \right) - g_{ij} \theta_{ij}.$$
 (3)

After dividing operating conditions into several regions, the segmented linear power flow models with minimized errors can be trained for each region. The segmented linear models are derived by optimizing the independent variables  $\varphi_i(v_i) = v_i^{k_i}$  and  $\varphi(\theta) = \theta_{ij}$  based on historical data. The following optimization model is used to select the independent variable with minimized linearization error of each region and then obtain specific linear power flow models.

$$\begin{split} \min_{k_{i}} & obj = \sum_{N_{c}} \sum_{N_{br}} \left( \left| PF_{ij}^{L} - PF_{ij}^{AC} \right| + \left| QF_{ij}^{L} - QF_{ij}^{AC} \right| \right) \\ & + \sum_{i} M_{1} a + \sum_{i} M_{2} t_{low} + \sum_{i} M_{3} t_{up} \end{split} \tag{4}$$

Decision variable: 
$$k_i$$
,  $i = 1, 2, ... N_{i}$  (5)

$$P_{ij,loss}^{L} + a \ge 0, \ Q_{ij,loss}^{L} + a \ge 0, \ a \ge 0$$
 (6)

$$VBRL_{ij} - e_{low} \leq v_{cp,L} \leq VBRU_{ij} + e_{up}, \ e \geq 0 \eqno(7)$$

$$v_{cp,L} \stackrel{define}{=} v_{i,0} v_{j,0} + \frac{v_{j,0} \left(v_i^{k_i} - v_{i,0}^{k_i}\right)}{k_i v_{i,0}^{k_i-1}} + \frac{v_{i,0} \left(v_j^{k_j} - v_{j,0}^{k_j}\right)}{k_j v_{j,0}^{k_j-1}}, \quad (8)$$

Linear Power Flow: 
$$(2) - (3)$$
,  $(9)$ 

where  $PF_{ij}^L/QF_{ij}^L$  and  $PF_{ij}^{AC}/QF_{ij}^{AC}$  are the linear and AC branch power flow, respectively,  $v_{cp,L}$  is the linearization of  $v_iv_j$ ,  $VBRU_{ij}$  and  $VBRL_{ij}$  are the upper and lower bounds of  $v_{cp,L}$  after tightening. a and t are the nonnegative slack variables, and M is a large positive number.  $v_i$  and  $v_j$  represent the voltage magnitudes from historical data.  $v_0$  is the initial point of Taylor expansion. The exponent  $k_i$  is the decision variable. The segmented linear models for given regions with minimized errors are obtained by solving the optimization problem (4)-(9).

Support vector machine (SVM) as a common classifier is used to divide the operating conditions and determine the applicable segmented linear model according to the output class of the classifier. The historical operating conditions of the power grid are regarded as the input training samples of the classifier. The results of divided sub-regions are regarded as training labels. The trained classifier can automatically place a new operating condition (scenario) into the proper class, and it indicates the applicable segmented linear model.

Finally, the individual sensitivity factors can be calculated according to different segmented linear power flow models for CCOPF modeling.