

Designing Medium Voltage Distribution Networks Based on Probabilistic Risk Criteria to Improve Network Reliability – Part II: A Study on Load Uncertainty and Distributed Generation

S.Najafi Ravadanegh, H.Mohammadian-Alirezachaei

Abstract–This study presents the optimal placement of automatic power switches in the distribution network of Tabriz city, considering distributed generation sources and uncertainties arising from load forecasting and renewable energy generation. Given the urban context of the design scope, photovoltaic source penetration is prioritized. The design focuses on optimally positioning remote control switches (RCS) to enhance network reliability and meet automation objectives. For this purpose, an interface has been created to use the optimization force of MATLAB software and the technical computing force of DIGSILENT software. In this case, system index's are calculated in DIGSILENT software for each scenario in MATLAB, and by calculating the cost probability distribution function, the level of riskiness of the system is evaluated according to the number of installed switches. Due to the large volume of study results, a part of the studies and research results will be presented in this article.

Keywords: Reliability; design; automatic switches; automation; renewable resources; Risk.

I. INTRODUCTION

A medium-voltage distribution network is considered the backbone of modern distribution systems. Principled, optimal design based on electrical, geographical and financial limitations can play a significant role in reducing the costs of distribution companies and prevent unscientific assumptions, and on the other hand, respond to the needs of consumers in an efficient method [1-3].

The purpose of this research is to provide a methodology for the optimal design of medium voltage distribution networks to increase the reliability of the network in the presence of important uncertainties and to evaluate the risk values of

important network variables in the presence of these uncertainties.

The basic issue raised in this research is a new approach to the medium voltage distribution network from the point of view of design. The progress of new technologies in the field of operation equipment, measurement, as well as the rules governing distribution companies, new topics have also entered the studies of distributed networks [4-6]. It is also important to note that current design studies target future networks, and over the coming years, numerous emerging technologies will integrate into distribution systems. Paying attention to the electricity market and smart distribution grids has made it necessary to create new attitudes in the future distribution networks. As examples of research in this area, reference [7] proposes a hybrid AC/DC distribution network in which the use of existing elements in the network structure is considered to improve the system reliability. In reference [8], an operational planning framework is proposed to evaluate the optimal design for multiple microgrids with the aim of increasing the security and reliability of power supply. In a different direction, research conducted in [9] developed a cyber-physical laboratory platform for distribution networks in which remote access tools were used to model and analyze security and reliability issues in power systems. Also, in another study presented in reference [10], an integrated approach is proposed to evaluate the performance of the distribution network from the perspective of self-resilience and identifying vulnerable points. The above-mentioned cases show that presenting new methods in the design and operation of future distribution networks will not only help distribution companies in advancing their current goals, but will also prepare them in planning to face new future events. In the conducted studies, many of the existing uncertainties have not been taken into consideration, but this issue is important in the mid-term and long-term design. Most studies have been conducted in an algebraic and non-probabilistic environment. In short, it can be said that the aim of this project is to provide

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a suitable framework to solve the problem of optimal design of medium voltage distribution by considering the uncertainty of the important input parameters and assessing the risk caused by them in a probabilistic environment [11-14].

II. STUDY SYSTEM AND REGIONAL AUTOMATION STATIONS

Fig. 1 illustrates the locations of automation stations within the study area. With this assumption, out of the total number of 1288 switches, 58 switches are RCS type and the rest are manual switches. Additionally, since some parts of the network include overhead feeders, switches are included along these feeders; in this study, there are 114 such switches.

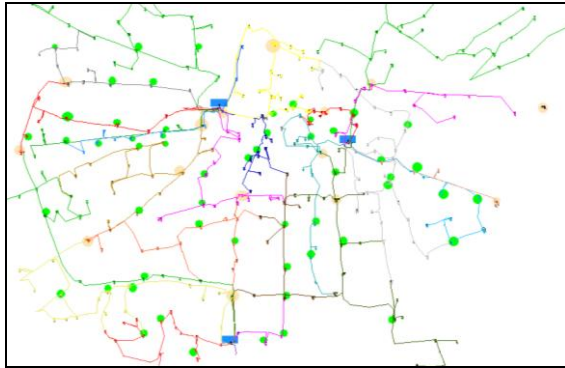


Fig. 1. The location of sub-distribution, automated substations and medium voltage feeders in the study area

For the design of the network in the year of horizon, it has been assumed that there is no remote-control switches in the network, and with this assumption, the type and number of switches necessary to improve the reliability of the network and increase its level of automation have been studied. In this study, according to the investigations, the level of network reliability for different scenarios after installing a certain number of remote-control switches has not improved much and remains almost constant. In the current study and for the studied network, the maximum number of switches that can be selected for each feeder is 8 switches. According to the conditions of each feeder, the remote-control switches may be selected between 0 and 8. The selection conditions depend on the study conditions, including the amount of load, the amount of penetration of PVs, the arrangement of the medium voltage network in each scenario.

III. INCREASING THE BURDEN OF THE PILOT AREA

According to the growth of the load of the region, the amount of load of each substation has been calculated and calculated as the last load point for the year of horizon 1400. It should be noted that a simultaneous increase in substation loads does not affect switch placement; therefore, attention should be focused on potential variations in load forecasting.

A. PROBABLE MODELING OF DISTRIBUTED GENERATION SOURCES IN THE PILOT AREA [3].

Since the rate of growth of distributed generation sources (including photovoltaic cells) in the area under study is not known as a point, therefore, the penetration rate of distributed generation sources should be studied in the form of surface growth within the study area and up to 20% maximum. Based on this, the studied area was divided into regular blocks with dimensions of 100 meters by 100 meters using GIS software. For each block, the probability of the presence of photovoltaic distributed generation sources was calculated using the normal distribution function between zero and 20% of the block's consumption load and entered into the calculations. After this step, by using the spatial attachment of photovoltaic cell blocks to each substation, the amount of power of the photovoltaic cell connected to each substation is calculated and entered into the optimization software. In this project, the studied area is divided into 4366 square blocks. Figure (2) shows the overview of this zoning in the GIS software environment.

As an example, table (1) shows the coordinates and the amount of distributed generation resources of the blocks attached to substation No. SB020034 on feeder No. 5 of Golestan. Also, in figure (3), the supply range of this substation is calculated and shown by using spatial attachment of the closest distance of each block to a certain substation.

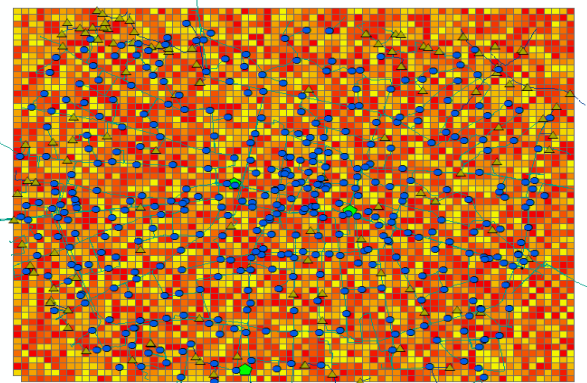


Fig. 2. Possible penetration values of distributed generation sources between 0 and 20% load

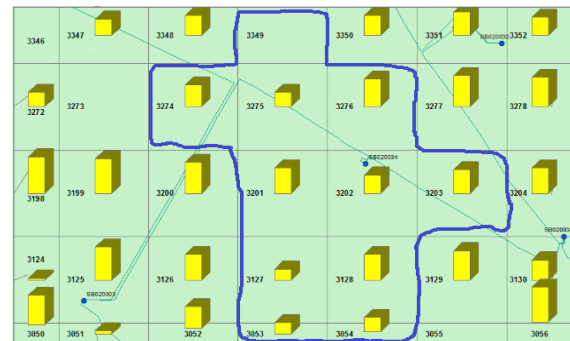


Fig. 3. The supply range of substation SB020034 by blocks of distributed generation resources on feeder No. 5 of Golestan

TABLE I
BLOCKS ATTACHED TO SUBSTATION NO. SB020034 ON GOLESTAN FEEDER NO. 5

Substation Code	Substation Capacity	Feeder	Substation X	Substation Y	Block Num	Block X	Block Y
SB020034	1000	F5 Golestan	611517.69	4216518	3127	611423.8	4216383
					3128	611523.8	4216383
					3201	611423.8	4216483
					3202	611523.8	4216483
					3203	611623.8	4216483
					3274	611323.8	4216583
					3275	611423.8	4216583
					3276	611523.8	4216583
					3349	611423.8	4216683

IV. RISK MODELING IN THE PROBLEM OF LOCATING AUTOMATIC SWITCHES

The final goal of this project is the optimal design of medium voltage distribution networks based on possible risk criteria to increase reliability by finding the right place to install remote control switches (RCS) with the aim of improving the reliability index at the lowest possible cost. For this purpose, in the above problem, risk modeling has been done by taking into account the uncertainties caused by load demand and penetration of Photovoltaic renewable energy sources.

As previously stated, energy not supplied (ENS) is critical for distribution network operations, though its level of importance varies among different operators. Risk-taking companies give less importance to lost energy, but more risk-averse companies are trying to reduce this risk with higher investments. Hence, accurate risk modeling is essential. In order to control the related risk in an environment with load uncertainty and penetration of Photovoltaic resources, as detailed in the previous chapters, a function to measure the risk related to the reliability gain is included in the objective function. CVaR (Conditional Value at Risk) is one of the most effective risk measures used in statistical planning. CVaR evaluation is able to quantify the risk potential beyond VaR (value at risk) and represents a suitable approach to manage the inherent problem of risk. Based on the above definition and considering the level of certain probability α , VaR is a sample of the lowest cost that, with the probability α , the cost does not exceed a certain value. Considering that the conditional expected cost CVaR is higher than that value, this parameter as a function of the decision variables is convex CVaR, so it can be effectively controlled or optimized using linear or convex programming. For a given α ($\alpha \in (0,1)$), VaR represents the lowest cost, with the assurance that the probability of obtaining an overall cost above this value is less than $(1-\alpha)$.

CVaR can be modeled as the following linear equation:

$$CVaR(\alpha, x) = \max\left\{\eta - \frac{1}{1-\alpha} E_s\{\max\{\eta - f(x, s), 0\}\}\right\}, \forall \alpha \in (0,1) \quad (1)$$

$$s(s) \geq (\eta - f(x, s)), s(s) \geq 0 \quad (2)$$

$f(x, s)$ is the profit function in scenario s .

There is always a possibility of error and interruption of energy supply in power networks. When there is a power outage in the distribution system based on the occurrence of a fault, the supply of a number of loads is interrupted. On the other hand, profitability increases by feeding loads with higher reliability. Therefore, in order to show the importance of the energy not supplied of the system, $C_{ENS}(s)$ is assumed as a function of $f(x, s)$ in CVaR calculation. Therefore, the value of CVaR is calculated based on relations (1) and (2) and can finally be entered into the objective function as follows:

$$\begin{aligned} \text{Risk based Cost Function} \approx \\ \text{minimize } ((1-\beta) * \text{cost}_{RCS} + \beta * CVaR) \end{aligned} \quad (3)$$

It should be noted that the cost of installing RCS switches is cost_{RCS} .

The purpose of the above function is to minimize the system costs for risk aversion through the parameter β . Entering the weighted correction factor β indicates the balance between risk and expected costs. β values vary from 0 to 1. Therefore, considering zero value for the β parameter, the problem becomes a risk-free problem. As long as the value of β increases, the expected profit becomes less important and becomes a risk aversion problem.

It is clear that for a risk-averse company, cost_{RCS} increases in proportion to its risk aversion. Because by adjusting the level of risk aversion, the optimal result will change and due to the higher importance of energy not supplied and the decrease of CVaR, the number of installed Switches will increase and, as a result, the installation cost will increase.

According to the explanations provided, the mentioned problem should be solved for several β s. In the following, the results obtained from the optimization of the above problem will be presented by setting β to different numbers.

Figure (4) shows the relationship between CVaR and Cost in several examples of different BBBs. It should be noted that by analyzing the obtained results, a part of the results that were of great importance has been presented. As it can be seen from the figure, the cost of the system has increased with the

increase of β value. In this regard, the value of CVaR, which is considered here in line with the amount of energy not supplied for the sake of simplification, has been further reduced for risk-averse companies and according to the level of risk aversion of the company.

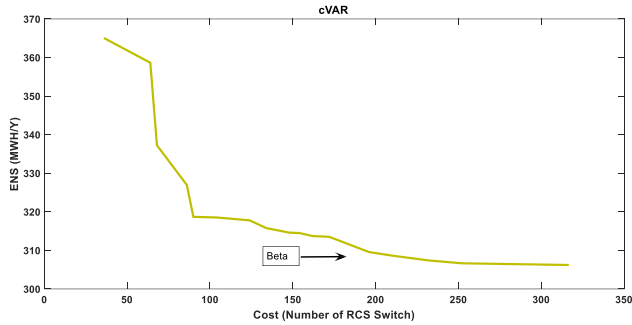


Fig. 4. CVaR diagram in terms of Cost in different β s

Here, the cost of the company is the same as the cost of purchasing RCS switches, and the benefit of reducing the energy not supplied of the system is expressed by installing these Switches. Table (2) shows the numerical representation of the relationship between the amount of energy not supplied and the cost of the system for different β s. In simpler terms, the table below shows that assuming the company's risk aversion is equal to β_i , this company needs capital equal to Total Cost_i so that the amount of energy not supplied is equal to Total ENS_i.

TABLE II
COMMUNICATION BETWEEN TOTAL ENS AND COST IN DIFFERENT β s

Total Cost (Number of RCS)	Total ENS	β	Risk No
36	365.0549	1	1
64	358.6485	2	2
68	337.2431	3	3
86	326.9911	6	4
90	318.6936	7	5
104	318.5283	8	6
124	317.7828	10	7
134	315.7829	12	8
148	314.5882	15	9
154	314.4962	16	10
162	313.7086	18	11
172	313.5171	20	12
196	309.5673	21	13
210	308.6393	22	14
232	307.39	23	15
252	306.6529	24	16
316	306.2133	25	17

In order to show the details of the obtained results, the information about the RCS Switches installed in the network for some β values is presented in Table (3).

If we assume that the network is assumed to be a completely risk-averse company with a risk aversion index of BBB, and if you do not have an RCS type Switches, it is assumed to be a completely risk-averse company with a BBB index. In this case, according to the obtained results and the statistics of the total number of Switches in the network, the company's risk-taking index for this number and location of RCS Switches will be as shown in Table (3). Also, in table (4), the calculation of the company's risk tolerance for the best acceptable scenario based on profit (reduction of energy not supplied) is shown. The results of the calculations according to the two tables below show that increasing the number of Switches and decreasing the amount of energy not supplied present almost equal risk indicators. Finally, according to the calculations, the probability distribution function of the values exposed to risk is presented in Figure (5). In this figure, the probability density function of company costs is presented and can be used in CVaR calculation.

TABLE III
CALCULATION OF THE COMPANY'S RISK TOLERANCE FOR THE BEST ACCEPTABLE
SCENARIO BASED ON COST (NUMBER OF SWITCHES)

available RCS Switches	Chosen RCS Switches	Total	
58	186	Statistics of Switches contributing to the design	Statistics of Switches contributing to the design
		1288	Total number of candidate Switches
		0.19	Maximum risk tolerance

TABLE IV
CALCULATION OF THE COMPANY'S RISK TOLERANCE FOR THE BEST ACCEPTABLE
SCENARIO BASED ON PROFIT (REDUCTION OF ENERGY NOT SUPPLIED)

Network RCS Switches	Amount of lost energy (MWh/a)	Reduction of lost energy (MWh/a)	risk index (β)
0	365	---	$\beta = 0$
244	306	59	$\beta = 0.224$
1288	105	260	$\beta = 1$

Table (5) presents all the selected switches. Also, in this table, the substations in which automatic switches are placed are shown. According to this table, it can be seen that in some substations several switches have been selected simultaneously for network automation.

TABLE V
THE FINAL LOCATION OF THE AUTOMATIC SWITCHES IN THE BEST ANSWER AND THE SUBSTATION CORRESPONDING TO IT

RCS Switch	Substation	RCS Switch	Substation	RCS Switch	Substation	RCS Switch	Substation
1-CBK1131835-RCS	SB010053	7-DCS1592047	SB020094	21-CBK1133215-	SB020056	25-DCS1590401	SB010125
1-CBK1132327-RCS	SB010053	7-DCS1592153	SB020108	21-DCS1589685	SB020049	25-DCS1593321	SB020113
1-CBK1132887-RCS	SB010036	8-CBK1131745-RCS	SB020068	21-DCS1590269	SB020053	25-DCS1593453	SB020090
1-CBK1134169-RCS	SB010036	8-DCS1591791	SB020096	21-DCS1590329	SB020069	25-DCS1593525	SB010122
1-DCS1589583	SB010048	8-DCS1592095	SB020099	21-DCS1591617	SB020131	25-DCS1593665	SB020158
1-DCS1590053	SB010070	8-DCS1598185	SB020117	21-DCS1592949	SB020066	25-DCS1593943	SB020071
1-DCS1590709	SB010076	8-DCS1598461	SB020107	21-DCS1593309	SB020064	25-DCS1594083	SB010076
1-DCS1591641	SB020133	8-DCS1598565	SB020113	21-DCS1594003	SB020086	25-DCS1594679	SB020048
1-DCS1592375	SB020088	8-DCS1601315	SB020098	21-DCS1598979	SB020126	25-DCS1595423	SB020094
1-DCS1594187	SB020106	9-DCS1593641	SB020049	21-DCS1599739	SB020092	25-DCS1596005	SB010129
1-DCS1597671	SB020095	9-DCS1599309	SB020118	21-DCS1601137	SB020153	25-DCS1596155	SB020070
1-DCS1600107	SB020038	9-DCS1600639	SB010130	21-DCS1610480	SB020012	25-DCS1596293	SB010136
1-DCS1600247	SB020060	9-DCS1608322	SB010009	22-DCS1594315	SB020127	25-DCS1596679	SB020140
1-DCS1600363	SB020112	9-DCS1608358	SB010008	22-DCS1594585	SB020066	25-DCS1597749	SB010119
1-DCS1600985	SB010038	10-CBK1133393-RCS	SB020040	22-DCS1594703	SB020107	25-DCS1597933	SB020050
1-DCS1602173	SB010044	10-DCS1596143	SB020120	22-DCS1598321	SB020067	25-DCS1598333	SB020158
1-DCS1604707	SB010046	10-DCS1600177	SB010044	22-DCS1598659	SB020086	25-DCS1598613	SB020072
2-DCS1589997	SB010074	10-DCS1600411	SB020064	22-DCS1599027	SB020093	25-DCS1598637	SB020096
2-DCS1590245	SB020105	10-DCS1609488	SB010006	22-DCS1602101	SB020130	25-DCS1599085	SB020120
2-DCS1590481	SB020072	11-DCS1591363	SB020143	23-DCS1589709	SB020109	25-DCS1599935	SB020105
2-DCS1591071	SB010033	11-DCS1595579	SB020122	23-DCS1590305	SB010148	25-DCS1600165	SB020057
2-DCS1592913	SB010054	11-DCS1600447	SB010020	23-DCS1592881	SB020119	25-DCS1600223	SB010071
2-DCS1593369	SB020106	12-DCS1594071	SB020135	23-DCS1593779	SB020110	25-DCS1600495	SB010147
2-DCS1595493	SB010068	12-DCS1599773	SB010136	23-DCS1596281	SB020159	25-DCS1600591	SB020106
2-DCS1597007	SB010075	13-CBK1131439-RCS	SB020154	23-DCS1597231	SB020088	25-DCS1600827	SB020046
2-DCS1597993	SB020071	13-CBK1133813-RCS	SB020111	23-DCS1600949	SB020126	25-DCS1601363	SB010008
2-DCS1598991	SB020059	13-DCS1594843	SB020073	23-DCS1601009	---	25-DCS1601653	SB010054
2-DCS1599715	SB020046	13-DCS1599061	SB020092	23-DCS1601279	SB020112	25-DCS1601905	SB010017
2-CBK1134089	SB010127	14-DCS1600973	SB010129	23-DCS1607834	SB020123	25-DCS1601985	SB020093
2-DCS1599819	SB020096	15-DCS1591665	SB020135	24-51737AB100001	SB020178	51336AB100043	
2-DCS1605668	SB020010	15-DCS1595075	SB020134	24-CBK1135279-	---	51336AB100044	
3-DCS1598729	SB020163	16-DCS1591115	SB020141	24-DCS1591849	SB010120	51336CB100019	
3-DCS1600131	SB020074	16-DCS1596937	SB010147	24-DCS1594175	SB020114	51347AB100004	
4-DCS1591433	SB020110	16-DCS1601961	SB020148	24-DCS1596553	SB010020	51737AB100002	
4-DCS1594443	SB020078	17-DCS1590733	SB020139	24-DCS1598505	SB020163	51737CB100003	
4-DCS1596865	SB020108	17-DCS1595507	SB010119	24-DCS1599415	SB020119	51738AB100005	
5-DCS1593837	SB020118	17-DCS1597365	SB020091	24-DCS1601291	SB020124	51738AB100006	
5-DCS1599109	SB020127	18-DCS1595543	SB020109	24-DCS1602041	SB020130	51738CB100017	
5-DCS1599553	SB020050	19-DCS1593967	SB020098	24-DCS1608370	SB010122	51747CF100006	
6-DCS1591351	SB020159	19-DCS1600687	SB020099	24-DCS1609464	SB010006	51747CF100007	
6-DCS1593093	SB020131	20-DCS1595637	SB020065	24-DCS1609476	SB010006	51749AB100001	
6-DCS1601399	SB010063	20-DCS1596901	SB010118	24-DCS1629130	SB020013		
		20-DCS1608346	SB010008				

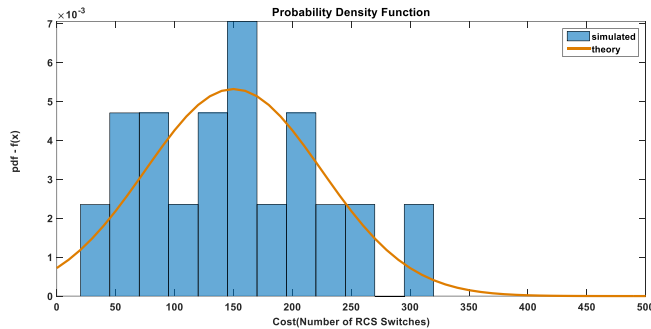


Fig. 5. Probability density function of company costs

V. CONCLUSION

This research has investigated the design of a medium-voltage distribution network in terms of reliability and based on probabilistic risk criteria. The studied network is a large part of the medium voltage network of Tabriz city, including the affairs of Golestan, Roshanayi and Emamiye. In this study, the maximum reliability values that can be obtained are calculated and corresponding to those optimal values that can be obtained at an acceptable cost. As an important part of the research, the impact of the influence of distributed generation resources with a probable nature and also some distributed generation resources with a deterministic and focused nature on the design of the network has been evaluated in index of reliability and the numerical values of the reliability index have been calculated and compared in the definite and probable space. The location of selected automatic Switches is also provided. Based on the results of the research, the risk aversion and risk tolerance of the company in terms of the cost of installing the switch and reducing the amount of energy not supplied has been calculated and compared with the ideal state.

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