



Solar microgrids fast and accurate fault detection, location and classification strategy using on-line phaselet, current injection kits', traveling-waves, and mathematical morphology

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A B S T R A C T

In this paper, a new fast and accurate method for fault detection, location and classification on multi-terminal direct current (MTDC) distribution networks connected to solar distributed generation and loads presented. Some issues such as DC resources and loads expanding, and try to the power quality increasing have led to MTDC networks' development. It is important to recognize the fault type in order to continue service and prevent further damages. In this method, a circuit kit is connected to the network. Fault detection is performed with the measurement of the current of the connected kits and the traveling-waves of the fault current and applying it to a mathematical morphology filter, in the Fault time. Determine the type and location of faults using a mathematical morphology filter, circuit equations and current calculations. DC series and ground arc faults are considered as DC distribution network disturbances. The presented method was tested in a solar DC network connected to energy storages and solar resources with many faults. The results illustrate the validity of the proposed method. The main advantages of the proposed fault location and classification strategy are higher speed and accuracy than conventional approaches. The fault location error of the presented algorithm is less than 6.5 percent in the worst case. This method robustly operates to changing in sampling frequency [0.5-50 KHz], fault resistance [0.005-120 Ohm], and works very well in high impedance fault.

Keywords:

MTDC solar network, fault detection and classification, mathematical morphology, current injection kit, on-line phaselet

1. Introduction

The rapid and accurate detection, location and classification of the fault are highly effective in developing the reliability indexes, diminishing expected energy not supply, and increasing the speed of network recovery and reconstruction. Growing DC consumption in distribution networks,

connecting photovoltaic resources to the network, trying to enhance the power quality and responding to further loads has expanded MTDC solar networks. And the other hand, power electronic converters usage improves the capacity, quality, size, weight, and equipment consumption and the load flow on the MTDC networks.

The implementation of conventional schemes for fault detection and location on MTDC networks has some problems. The accuracy of impedance-based methods is not sufficient at power frequency for distribution networks. Traditional protection methods that are based on undervoltage/overcurrent [1, 2], rate of change in current/voltage [3], or either lack the required sensitivity for detecting high-resistance faults, or are Unreliable to communication delay and failure. Passive-Overcurrent protection based on discrete wavelet transform is proposed to different types of faults detection and classification [1, 4]. Yet, none of the above schemes meet the speed required for an MTDC solar network [2], [4]. In [5], a directional current-based protection scheme is proposed for low voltage dc (LVDC) grids. However, apart from these schemes' unreliability to communication fails, the operation of high bandwidth fiber optic communication is expensive and not excusable for LVDC grids.

The protection challenges of the solar and wind networks analyzed in [6]. A novel fault detection, characterization, and fault current control algorithm for a standalone solar-photovoltaic (PV) based DC microgrids presented in [7]. A distributed incremental adaptive filter (DIAF) controlled utility interfaced photovoltaic (PV) - battery microgrid system is presented with power quality features [8]. In [8-16], Solar DC microgrids control and management methods and voltage stability are presented.

Traveling waves based methods have been used widely to detect transmission line faults [17, 18]. Methods based on traveling waves and time-domain analysis are used in HVDC networks [19]. In some methods, Wavelet has been mixed with other techniques such as artificial neural networks, fuzzy logic increases the performance of the proposed protection algorithm [17-24]. Some of these methods depend on the location of the fault and the arrival time of the traveling waves [25]. Determination of the fault location in the MTDC distribution grids should be done with high accuracy and sureness due to multiple and longitudinal splits and branches. In [26, 27], a DC bus protection scheme in DC distribution grids have been suggested. The actual impedance approximation method is used in a network. Detection of voltage drop in DC microgrid is provided using the power probe unit proposed [1,

28, 29]. Few power electronics converter-based power probe units for injecting AC signals are in [26, 30-33]. The simple solar network or single-bus solar microgrids investigated in most of the mentioned researches. In this paper, a high-precision and high-speed method is proposed to fault determination, location and classification in multi terminal DC solar distribution networks using current injection, mathematical morphology, and online phaselet. Using the traveling waves derived from the online phaselet transform and mathematical morphology, the fault occurrence and its type are identified. The attenuation factor (γ) is calculated based on the sampling of the fault current in the interconnected multi-bus solar system. This method was tested on an MTDC solar network with many faults. The results illustrate the validation of the proposed method.

2. Method

2.1 Current injection technique

By adjoining the C_q branch to the current injection kit provided in [34], part of the switching current passing through this branch and reduces the switching losses and reduces the error of fault locating in the line close in fault. Figure 1 illustrates the proposed kit. The optimum value of C_q is selected by considering the total loss of switching, the cost of capacitors installing, and reactive losses. The kit that be shown in figure 1 connected to the MTDC network buses. Figure 2 illustrate the equivalent circuit of the faulted network and the kit in the fault time. If the fault does not occur, the current does not pass through the connected kits.

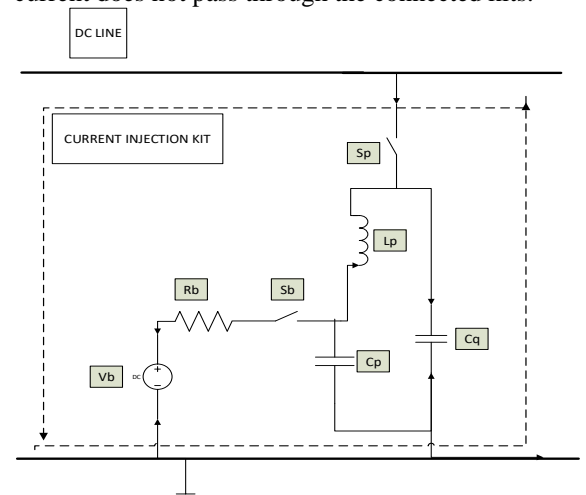


Figure 1 Proposed current injection kit

The equation of the current of the faulted part of the grid is 1 and 2.

$$\frac{d^2 i_p(t)}{dt^2} + \frac{R}{L} \frac{di_p(t)}{dt} + \frac{1}{LC} i_p(t) = 0 \tag{1}$$

$$i_p(t) = K_1 e^{-\gamma t} \sin(\omega_d t) + K_2 e^{-\gamma t} \cos(\omega_d t) \tag{2}$$

$$\omega_d^2 = \omega_n^2 - \gamma^2 \tag{3}$$

$$\gamma = \frac{R}{2L}$$

$$\omega_n = \sqrt{\frac{L_d + C_q}{C_q C_p (L_d + L_p) + C_p L_d L_p}}$$

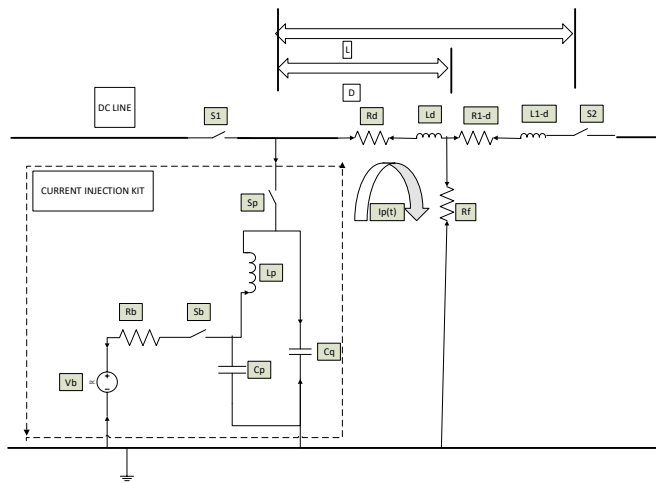


Figure 2 Equivalent circuit of the faulted network and the current injection kit

Due to the line impedance-to-length dependence, the natural frequency (ω_n) of fault current calculates by 3.

$$\omega_n = \sqrt{\frac{L_u d + C_q}{C_q C_p (L_u d + L_p) + C_p L_u d L_p}} \tag{3}$$

Thus, the location of the occurred fault is calculated using 4 and the error percent of the estimation of fault's location is calculated by 5.

$$d = \frac{C_q - C_p L_p C_q \omega_n^2}{L_u (C_p (L_p + C_q) \omega_n^2 - 1)} \tag{4}$$

$$= \frac{C_q - C_p L_p C_q (\omega_d^2 + \gamma^2)}{L_u (C_p (L_p + C_q) (\omega_d^2 + \gamma^2) - 1)}$$

$$\varepsilon = \left| \frac{d_{cal} - d_{act}}{d_{act}} \right| \times 100 \tag{5}$$

The equation of the curve of the fault current is 6, and by two samples of the fault current, the attenuation factor γ is calculated by (7).

$$i_{p'}(t) = K e^{-\gamma t} \tag{6}$$

$$\gamma_n = \frac{\ln(i_n) - \ln(i_{n+1})}{t_{n+1} - t_n} \tag{7}$$

2.2 Online phaselet transform

The wavelet transform holds the correlation between the frequency and time of the signal and is a very good ability at the time-frequency resolution. This feature of the wavelet transform can be used to determine the arrival time of the traveling wave and its following reflections. It is well known that discrete wavelet transform (DWT) is operated to different sampled (digitized) signals, to show their time-scale representation. The phaselet transform is a shift invariant wavelet transform. The mother phaselet is the mother wavelet multiply to a phase shifter functions. To execute this transform, the primary signal is passed into a band-pass filter (G is named mother phaselet) to give a detail part, for the first level. At the same level, and by convolving the signal with a low-pass filter (H), results in an approximate component. G and H are orthogonal vectors with $N \times 1$ elements [26]. For the second level, the approximate component is down-sampled by two, i.e. its samples are halved, and then are passed into G and H to give the next level approximate and detail components, respectively at this level. Continuing this algorithm to the i th level, makes the original signal to be decomposed to i detail components and an approximate one. This scheme is presented in Figure 3 up to four decomposition levels.

To present the above description in a mathematical form, for the i th level of decomposition, the detail, and approximate signals may be obtained as equations 8 and 9.

$$A_i = A_{i-1} \times G \tag{8}$$

$$D_i = A_{i-1} \times H \tag{9}$$

According to the aforementioned descriptions, in order to find the i th level detail component, a serial process must be done through successive

convolutions of approximate components with band-pass and low-pass filters up to the wanted decomposition level.

In on-line usages of DWT/DPT, this successive manner, which is time-delay, is not justifiable. In these applications, some detail or approximate components at some predefined frequency bands (various levels of decomposition) must be inspected. But, according to equations 8 and 9, the elements of the components in the i th level cannot be computed unless the approximate component of the $(i-1)$ th level is fully finished, and the latter cannot be defined unless the computations of the previous level are finalized. Consequently, the monitoring or inspecting mission cannot be produced, unless through a successive method. In this section, a digital filter for on-line applications of DPT is formed. The first feature of this filter is that the k th element of $D1$ or $A1$ comes out along with the k th element of all of the upper-level detail or approximate components. Let equation (1) and (2) with $i=2$ to be rewritten as equations 10 and 11.

$$G_j = H_{ii} G_{i-1} = \left(\prod_{m=2}^i H_{mm} \right) \times G \tag{12}$$

$$A_j = X \times G^i \tag{13}$$

$$D_j = X \times G_i \tag{14}$$

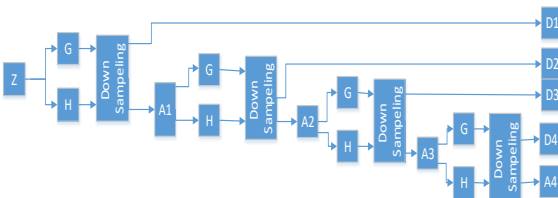


Figure 3 Four decomposition levels of main signal Z by DPT

$$A_2 = A_1 \times G = X \times G \times G \tag{10}$$

$$D_2 = D_1 \times G = X \times G \times H \tag{11}$$

Therefore

In the same way, the matrix for generating the elements of online approximate components can be introduced[35].

2.3 Mathematical morphology filter

Mathematical morphology in the field of time and with brief information windows examines the appearance of high-frequency signals. The erosion(f_e) and dilation(f_d) of the F input signal with the domain D_f and the structural element S with the domain D_s are defined as 15 and 16. And using the dilation and erosion the close(f_c) and open(f_o) relations are defined as 17 and 18, and the morphological filter is defined according to the four relations that have been introduced in the form of equation 19.

The equation 19 filter operates at high accuracy in detecting the range of disturbances but is incapable of detecting the direction of the disturbance signal. This problem is resolved by using equation 20.

$$f_e(k) = (f \ominus s)(k) = \max \{f(k+l) - s(l)\} \tag{15}$$

$$f_d(k) = (f \oplus s)(k) = \max \{f(k-l) + s(l)\} \tag{16}$$

$$f_c(k) = (f \cdot s)(k) = ((f \ominus s) \oplus s)(k) \tag{17}$$

$$f_o(k) = (f \circ s)(k) = ((f \oplus s) \ominus s)(k) \tag{18}$$

$$MF(k) = \frac{(f(k) \circ (f \cdot s)(k))}{2} + \frac{(f(k) \cdot (f \circ s)(k))}{2} \tag{19}$$

$k \in D_f$ and $l \in D_s$ Such that

$$D_f = \{0, 1, 2, \dots, K-1\}$$

$$D_s = \{0, 1, 2, \dots, L-1\}$$

Based on [32-34], when the erosion signal lags the dilation signal, the MMF output has a positive pole; when the erosion signal leads the dilation signal, the MMF output has a negative pole. Also, when there is no sudden change in the initial signal, there isn't the phase difference between the erosion and dilation signals and the MMF output is zero.

With applying the online phaselet to the disturbance signal in MTDC distribution networks, traveling waves is obtained.

By applying the traveling wave of the fault current signal to the morphological filter, for various types of fault, faults type is detected. Many types of faults are simulated in the Cigre benchmark MTDC solar network and the accuracy of the proposed method is

validated in part 3. The proposed method's Flowchart is shown in figure 4.

3. Simulation results

In order to examine the proposed method, a multi-terminal DC solar distribution network was simulated in PSCAD/EMTDC software. This network includes cable and overhead lines and connected to the power grid and several solar distributed generation and loads. Loads include

electrical vehicle stations, agriculture electrical vehicle and other public and commercials loads. Distributed generation include photovoltaic farm, solar houses and parking. Figure 5 illustrates this solar network. Different types of faults were simulated with different conditions in terms of the location of faults, arc's resistance, and type of faults.

$$MMF(k) = \begin{cases} MF_d(k) - MF_e(k) \Leftrightarrow MF_d(k-1) < MF_d(k) \\ MF_e(k) - MF_d(k) \Leftrightarrow MF_e(k-1) < MF_e(k) \end{cases}$$

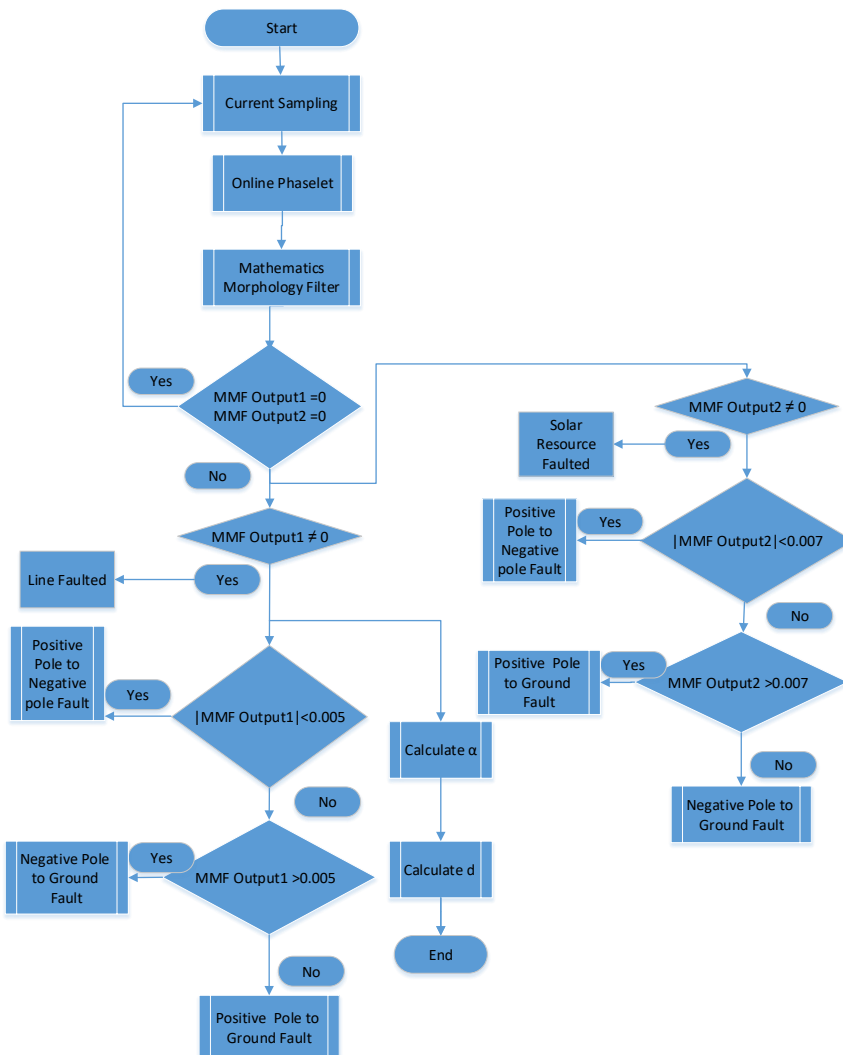


Figure 4 Proposed method flowchart

Fault current signals were measured in all mode. In each case, two samples used to calculate the fault current attenuation factor. Using the phaslet transform, traveling waves was calculated from fault current. The resulted waves applied to the mathematical morphology filter and the type of faults was determined for different situations with

high precision. Determine the fault location in lines faulted cases using equation 4 and the fault location error percent calculated in lines faulted cases by formula 5. Results of some of the lines faults simulations are presented in Table 1. Results of some of the solar resources faults simulations are presented in Table 2.

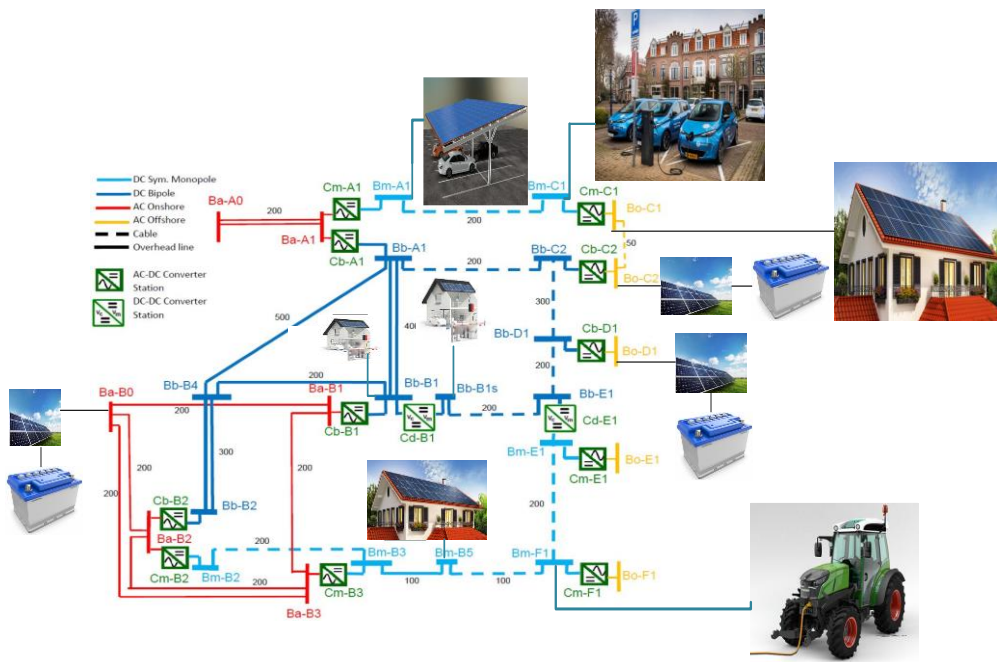


Figure 5 Multi terminal direct current solar Test network

Table1 Fault location and classification method's results in lines faulted cases

No.	Fault		Actual fault distance(%)	Fault resistance	PN fault			PG fault			NG fault		
	Beginning of Line	End of line			Calculated fault distance(%)	Fault location error (%)	Fault type	Calculated fault distance(%)	Fault location error (%)	Fault type	Calculated fault distance(%)	Fault location error (%)	Fault type
1	Bb-D1	Bb-E1	10	5	10.09	0.9	PN	10	0	PG	9.92	0.8	NG
2	Bm-B5	Bm-B3	95	50	95.5	0.526	PN	95.64	0.6736	PG	95.95	1	NG
3	Bm-B2	Bm-B3	65	50	65.31	0.4769	PN	65.47	0.7231	PG	65.27	0.4154	NG
4	Bb-B4	Bb-A1	30	0.05	30.22	0.7333	PN	30	0	PG	30.19	0.633	NG
5	Bb-B1	Bb-B4	5	0.05	4.98	0.4	PN	4.977	0.46	PG	4.984	0.32	NG
6	Bb-C2	Bb-A1	5	0.05	4.97	0.6	PN	4.981	0.38	PG	4.978	0.44	NG

7	Bb-A1	Bb-B4	15	0.05	14.98	0.133 3	PN	15	0	PG	14.97	0.153 3	NG
8	Bm-A1	Bm-C1	5	50	5.025	0.5	PN	5.043	0.86	PG	5.0099	0.198	NG
9	Bb-D1	Bb-C2	10	50	10.071	0.71	PN	10.056	0.56	PG	10.049	0.49	NG
10	Bm-F1	Bm-E1	35	50	35	0	PN	35.18	0.5143	PG	35.2	0.571 4	NG
11	Bm-F1	Bm-B5	60	5	60	0	PN	60.18	0.30	PG	60.08	0.133	NG
12	Bb-B1	Bb-A1	30	50	30.04	0.133 3	PN	30.042	0.14	PG	30.026	0.866 6	NG
13	Bm-B3	Bm-B2	99	0.05	99.2	0.020 2	PN	98.9	0.0101	PG	99.8	0.808	NG
14	Bb-B1	Bb-A1	45	0.05	45.04	0.088 89	PN	44.989	0.02444	PG	45	0	NG
15	Bb-E1	Bb-D1	40	0.05	40	0	PN	39.89	0.275	PG	40	0	NG
16	Bm-C1	Bm-A1	80	0.05	79.96	0.05	PN	80	0	PG	80	0	NG
17	Bb-C2	Bb-D1	50	5	50.11	0.22	PN	50.3	0.6	PG	50.24	0.48	NG
18	Bb-B4	Bb-B1	5	50	4.98	0.4	PN	5.3	0.6	PG	5.015	0.3	NG
19	Bm-E1	Bm-F1	15	50	14.95	0.333 3	PN	15.011	0.07333	PG	15.031	0.206 7	NG
20	Bm-B3	Bm-B5	55	0.05	55.11	0.2	PN	54.907	0.1691	PG	55.031	0.056 36	NG
21	Bm-F1	Bm-E1	5	5	70	0	PN	70.322	0.46	PG	69.989	0.015 7	NG
22	Bb-B1s	Bb-E1	30	0.05	30.22	0.733 3	PN	30	0	PG	30.19	0.633	NG
23	Bb-B4	Bb-B1	25	0.05	25.11	0.44	PN	24.81	0.76	PG	24.991	0.036	NG
24	Bb-B4	Bb-A1	45	0.05	44.86	0. 3111	PN	45.21	0.4667	PG	45	0	NG
25	Bb-B2	Bb-B4	40	5	40.01	0.775	PN	40.07	0.175	PG	39.86	0.35	NG
26	Bb-A1	Bb-B1	85	5	85.29	0.341 2	PN	85.2	0.2353	PG	85.18	0.211 8	NG
27	Bb-B1s	Bb-E1	5	50	5.025	0.5	PN	5.023	0.46	PG	5.016	0.32	NG
23	Bb-B4	Bb-B1	85	25	85.29	0.341 2	PN	85.2	0.2353	PG	85.18	0.211 8	NG
24	Bb-D1	Bb-C2	35	25	35.18	0.514 3	PN	35.27	0.771	PG	35.2	0.571 4	NG
25	Bb-D1	Bb-C2	5	25	70	0	PN	70.322	0.46	PG	69.989	0.015 7	NG
26	Bm-B2	Bm-B3	55	25	55.11	0.2	PN	54.907	0.1691	PG	55.031	0.056 36	NG
27	Bm-B3	Bm-B5	55	25	55.22	0.4	PN	54.031	0.05636	PG	55.907	0.169 1	NG

28	Bm-B2	Bm-B3	65	30	65.27	0.415 4	PN	65.31	0.4769	PG	65.47	0. 7231	NG
29	Bb-B1	Bb-B4	30	25	30.22	0.733 3	PN	30	0	PG	30.19	0.633	NG
30	Bm-B2	Bm-B3	5	30	70	0	PN	70.322	0.46	PG	69.989	0.015 7	NG
31	Bb-B4	Bb-A1	99	30	98.9	0.010 1	PN	99.8	0.808	PG	99.2	0.020 2	NG
32	Bb-E1	Bb-D1	40	50	39.89	0.275	PN	40	0	PG	10.11	0.275	NG
33	Bm-B3	Bm-B2	30	30	30.042	0.14	PN	30.026	0.8666	PG	30.04	0.133 3	NG
34	Bb-B1	Bb-A1	10	5	10.09	0.9	PN	10	0	PG	9.92	0.8	NG
35	Bb-E1	Bb-D1	45	30	44.96	0.088 89	PN	44.989	0.02444	PG	45.04	0.088 89	NG

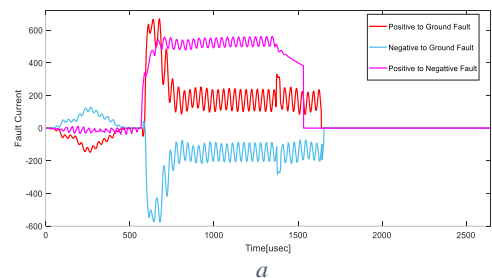
Table2 Fault classification method results in solar resources faulted cases

No.	Bus	Applied Fault		PN	PG	NG
		Bus type	Resistance			
1	Bo-D1	Source	0.005 & 5	PN	PG	NG
2	Ba-B0	Source	0.005 & 25	PN	PG	NG
3	Bm-B5	Source	10 & 25	PN	PG	NG
4	Bo-C2	Source	0.005 & 25	PN	PG	NG
5	Bm-A1	Source	10 & 25	PN	PG	NG
6	Bo-D1	Source	10 & 25	PN	PG	NG
7	Ba-B0	Source	5 & 15	PN	PG	NG
8	Bm-B5	Source	0.005 & 5	PN	PG	NG
9	Bo-C2	Source	5 & 15	PN	PG	NG
10	Bm-A1	Source	0.005 & 5	PN	PG	NG
11	Bo-D1	Source	8 & 20	PN	PG	NG
12	Ba-B0	Source	8 & 20	PN	PG	NG
13	Bm-B5	Source	8 & 20	PN	PG	NG
14	Bo-C2	Source	8 & 20	PN	PG	NG
15	Bm-A1	Source	8 & 20	PN	PG	NG

Three types of faults were applied, with the distance of 60% of the length of the line between the bus Bb-E1 and the bus Bb-B4 and among the fault's resistance equal to 15 ohms, including positive pole

to the ground, negative pole to the ground and positive pole to negative pole. Fault currents were measured in three modes. Figure 6-a and Figure 7-a shows the fault currents in each type of operated fault.

The three-mode phaselets of fault currents were calculated, using online phaselet and db4 as mother wavelet with shift invariant phase zero degree for lines faults, 30 degrees for solar resources faults. Application of online phaselet has increased the speed of fault detection and fault classification. Figure 6-b and Figure 7-b illustrates its diagrams. Figure 6-c and Figure 7-c shows the output of the mathematical morphology filter for three types of faults. Depending on the diagrams of Figure 6-c, the proposed method operates with high accuracy and speed to detect types of faults.



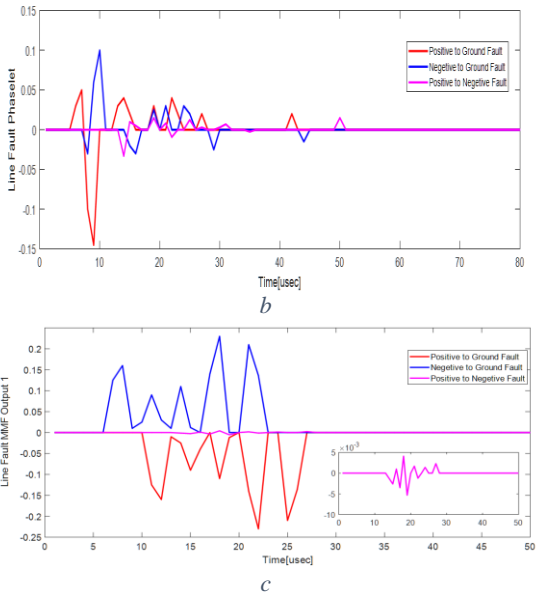


Figure 6 Line faults signals

6-a: Fault current in three modes 6-b :online phaselet output of fault current, 6-c :mathematical morphology filters out put

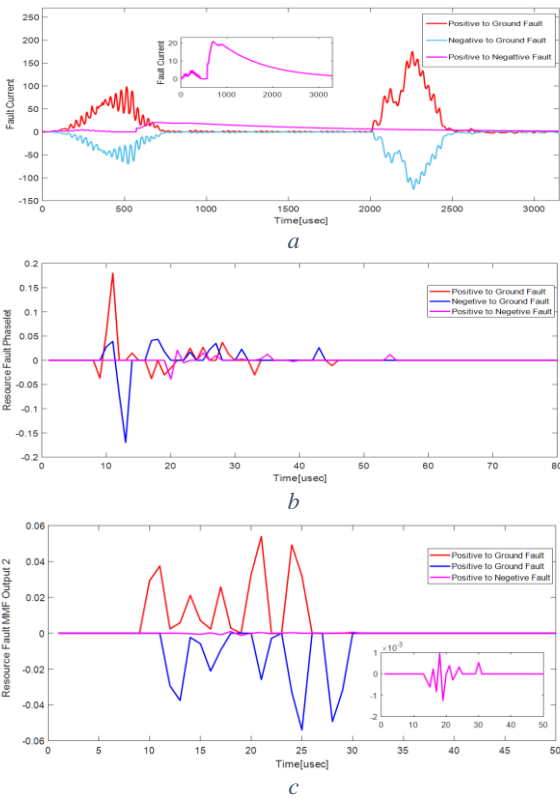


Figure 7 Source faults signals

7-a: Fault current in three modes, 7-b :online phaselet output of fault current, 7-c :mathematical morphology filters out put

4. Classification criteria Selection

2500 various faults evaluated on the MTDC solar network with different conditions of arc resistance, sampling frequency, location, and type of faults. Some of which are presented in tables 1, 2, and 3. The value of the mathematical morphology filter output of all line faulted cases given in Figure 8 and solar resource faulted cases is shown in Figure 9.

According to various line faults conditions simulations, the fault classification criterion was considered equal to ± 0.005 for line faults MMF output 1.

The various sources faults applied to microgrid and by attention to results, ± 0.007 for solar generation faults MMF output 2. Faults classify properly with selected thresholds according to Figure 8 and Figure 9.

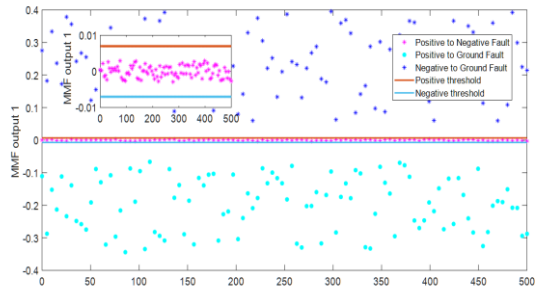


Figure 8 Value of the mathematical morphology filter output of various line faults

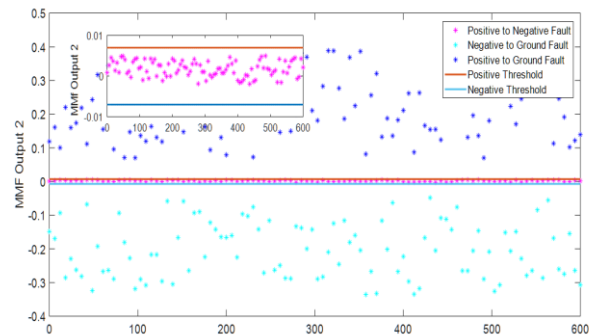


Figure 9 Value of the mathematical morphology filter output of various resources faults

5. The fault resistance effect on the proposed fault location method

By adding the Cq branch to the flow injection kit presented in [34] and installing the kits, the proposed fault location method was used to determine the fault location. Also, the error of the fault location was significantly reduced compared to the presented

and with variable values of arc resistance from zero to 120 ohms using the proposed method, the location

method in [34] for faults near the beginning of the line (close in fault) with high fault resistance. The error of the method presented in [34] is shown in Figure 10. By applying different faults along the line

of the occurrence for all types of fault is calculated. Figure 11 illustrate the average error of the three types of fault by the proposed fault location method.

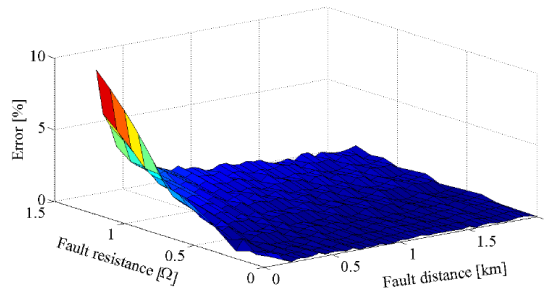


Figure 10 error of the method presented in [34]

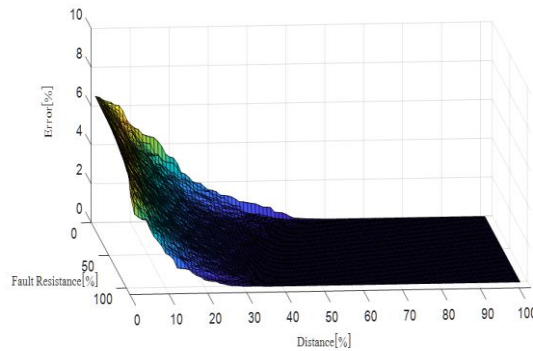


Figure 11 error of the presented method

According to Figure 11 and the results of Table 1, the proposed method has a good response to the impedance fault. Faults with 120-ohm resistance applied at various points along the line length and

6. Sampling Frequency Effect

The proposed method was used by different sampling frequency from 500Hz to 50KHz, and the results of 0.5, 1, 2, 10, 20, 30, 40, and 50KHz sampling frequencies given in Table 3. Furthermore, the arc resistance and applied fault location changed to determine the effect of sampling frequency variation in the different fault cases and conditions.

the fault location calculated using the proposed method.

The maximum of the fault location error is less than 6.5%. According to the above mentioned, this method is robust to high impedance fault.

The results of Table 3 clearly show that changing the sampling frequency in a wide range does not change the efficiency and accuracy of the proposed method in the fault locating. For each fault types using the proposed method, the fault location was computed with very lowly error.

Table 3 Results of fault location method with various sampling frequency

Sampling frequency	Actual fault distance (%)	5	5	25	30	45	60	85	95
	Fault resistance	5	0.05	25	25	5	25	0.05	5
500 Hz	Estimated fault distance (%)	4.992	4.99	$\frac{25.123}{6}$	30.03	44.986	60	85.58	94.71
	Fault location error (%)	0.16	0.2	0.4944	0.1	0.0311	0	0.683	0.3053
1 kHz	Estimated fault distance (%)	5.09	4.99	25.19	30.15	45	60.7	84.6	95.8
	Fault location error (%)	0.18	0.2	0.76	0.5	0	1.167	0.47	0.842
2 kHz	Estimated fault distance (%)	5	5.037	25.15	30	44.972	60.156	85.9	95
	Fault location error (%)	0	0.74	0.6	0	0.06223	0.26	0.9412	0
5 kHz	Estimated fault distance (%)	5.017	4.995	24.4	29.99	45.2	59.69	85.2	95
	Fault location error (%)	0.34	0.1	0.24	$\frac{0.0333}{4}$	0.44445	0.5167	0.2352	0
10 kHz	Estimated fault distance (%)	4.991	5.019	25.09	29.89	45	60.8	85.09	94.62
	Fault location error (%)	0.18	0.38	0.36	0.3667	0	1.334	0.10588	0.3984
20 kHz	Estimated fault distance (%)	5.012	4.994	25.145	29.93	44.79	60	85.1	94.54
	Fault location error (%)	0.24	0.12	0.592	0.2334	0.4667	0	0.1176	0.4842
30 kHz	Estimated fault distance (%)	4.99	4.96	25.2	30.09	44.7	60.1	84.4	95.54
	Fault location error (%)	0.2	0.8	0.8	0.3	0.667	0.1667	0.706	0.4632
40 kHz	Estimated fault distance (%)	4.991	5	24.94	30.1	45	59.8	84.6	95.19
	Fault location error (%)	0.18	0	0.24	0.334	0	0.3334	0.47	0.2
50 kHz	Estimated fault distance (%)	5.001	5	25.1	30.057	45.04	59.7	85.18	95.7
	Fault location error (%)	0.02	0	0.4	0.1899	0.0889	0.5	0.2118	0.737

7. Conclusion

In this paper, a new method of fault detection and fault location in MTDC solar networks is proposed which can be used to detect and classify line faults and solar resources faults fast, accurate, and efficiently in MTDC solar microgrids. The occurred fault location is determinate exactly by connecting a kit to the solar microgrid. Using the proposed kit diminishes the fault location error, especially the faults that occurred at the line beginning and near the kit. In fault occurrence, fault detection is done with measurement of traveling waves of fault current by online phaselet and applying to a mathematical morphology filter. The fault type classification is done according to the mathematical morphology

filter amount. The line faults distance determines using the circuit equations and current calculations. The accuracy of the presented method in an MTDC solar system was tested with various faults in terms of type, location, resistance, and sampling frequencies. The proposed approach is robust to sampling frequency change and Arc resistance fluctuations. The proposed method works with excellent performance in high impedance faults. More importantly, the proposed approach works properly for any MTDC solar network, irrespective of the grid's topology, that is, the number of network's radial lines, meshes, nodes, and rings.

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Appendix

Nomenclature			
R	Equivalent resistance of the KIT and faulted part of the grid	S_b , S_p	switches of the KIT
L	Equivalent inductance of the KIT and faulted part of the grid	L_p	Inductance of the KIT
C	Equivalent capacitance of the KIT and faulted part of the grid	V_b , R_b	Battery voltage and resistance of the Kit
i_p	Current of the faulted part of the grid	ω_n	The natural frequency
d	Distance of fault location	γ	The attenuation factor
C_p , C_q	Capacitors of the KIT	NG	Negative pole to ground fault
		PG	Positive pole to ground fault
		PN	Positive pole to negative pole fault
		f_d	Dilation of signals
		f_e	Erosion of signals
		f_c	The closing Function
		f_o	The opening Function

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