

# DYNAMIC PROGRAMMING FOR OPTIMAL ALLOCATION OF MAINTENANCE RESOURCES ON POWER DISTRIBUTION NETWORKS

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# Summary

- 1 Introduction**
- 2 Motivation
- 3 Optimisation Model
- 4 Heuristic Method
- 5 Dynamic Programming
- 6 Experiments
- 7 Conclusions

## *Labore*

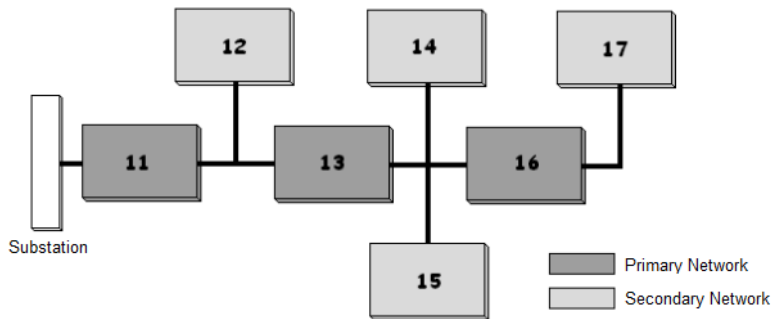
### Laboratory of Power Networks Optimisation

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# Radial Network



**Figure:** Reference Radial Network

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## Failure Rates

$$\lambda_s^t = \lambda_s + \sum_{e \in E_s} \lambda_e^t$$
$$\lambda_e^t = \lambda_e^{(t-1)} \sum_{n \in N_{k_e}} \delta_{k_e n} x_{en}^t$$
$$\sum_{n \in N_{k_e}} x_{en}^t = 1,$$

where:

- $\lambda_e^t$  is the failure rate for equipment  $e$  in the period  $t$ ;
- $\lambda_e^{(t-1)}$  is the failure rate for equipment  $e$  in the last year or the initial failure rate for equipment  $e$  ( $t = 1$ );
- $N_{k_e}$  is a set of all preventive maintenance actions;
- $\delta_{k_e n}$  is the failure rate multiplier for equipment  $k_e$  for action level  $n$ ;
- $x_{en}^t$  is a boolean decision variable denoting whether the equipment  $e$  received ( $x_{en}^t = 1$ ) or not ( $x_{en}^t = 0$ ) maintenance level  $n$  in the period  $t$ .

## Reliability Constraints

Thus, the SAIFI (The System Average Interruption Frequency Index) of system can be calculated:

$$SAIFI^t = \frac{1}{N_T} \sum_{s \in S} \lambda_s^t N_s,$$

where:

- $S$  is the set of all sections;
- $\lambda_s^t$  is failure rate of section  $s$  in the period  $t$ ;
- $N_s$  is the number of customers into section  $s$ ;
- $N_T$  is the number of all customers into the Network.



## Optimisation Problem

$$\min_{x_{en}^t} \sum_{t=1}^{HP} \left\{ \sum_{e \in E} \left[ \sum_{n \in N_{k_e}} (p_{k_e n} x_{en}^t) + \lambda_e^t c_{k_e} \right] \times \alpha_t \right\}$$

$$s.t : \quad SAIFI^t \leq SAIFI_{perm} \quad \forall t = 1, \dots, HP,$$

where:

- $E$  is a set that contains all the equipment which can receive preventive maintenance;
- $SAIFI_{perm}$  is the maximum permitted for SAIFI;
- $p_{k_e n}$  is the cost for action preventive level  $n$  for equipment  $k_e$ ;
- $c_{k_e}$  is the cost for action corrective level for equipment  $k_e$ ;
- $\alpha_t$  is a parameter which is related to each period.

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# State Space Search

## Example 30 equipments and HP=3

$$S_0 = \begin{bmatrix} 1 & 0 & 0 & 1 & 0 & 1 & 0 & 1 & 0 & 1 & 0 & 1 & 0 & 1 & 1 & 0 & 1 & 1 & 0 & 1 & 0 & 1 & 0 & 1 & 0 & 1 & 0 & 1 & 0 & 0 \\ 1 & 0 & 1 & 1 & 0 & 0 & 1 & 0 & 1 & 0 & 1 & 0 & 1 & 0 & 1 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 1 & 1 & 0 & 0 & 0 & 0 & 0 & 0 \\ 1 & 1 & 1 & 0 & 1 & 0 & 1 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 1 & 1 & 0 & 0 & 1 & 0 & 1 & 0 & 1 & 0 & 1 & 0 & 1 & 0 & 1 & 1 \end{bmatrix}$$

*Chosen\_action(S, S')*

- Depth Search with Simulated Annealing
  - Constructive Heuristic
  - Depth Search
  - Simulated Annealing

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## Knapsack Problem on Dynamic Programming

$$F_n(b) = \max \sum_{j=1}^n c_j x_j$$
$$\text{s.t. } \sum_{j=1}^n a_j x_j \leq b$$
$$x_j \in \{0, 1\}, j = 1, \dots, n$$

Where:

$\sum_{j=1}^n c_j x_j$ : total value of the selected elements to the Knapsack;

$\sum_{j=1}^n a_j x_j$ : total volume of the selected elements to the Knapsack.

## The aim is get $n$ -th value as from basic cases

Get  $F_n(a_0)$

Where  $F_k(a) = \max \{F_{k-1}(a), F_{k-1}(a - a_k) + c_k\}$

With  $F_0(a) = 0 \forall a$

To determine the optimal solution:

- Create an indicator  $p_k$  that is equal 0 if  $F_n(b) = F_{n-1}(b)$ , and 1 otherwise.
- Analyzes all indicators from  $p_n$  up to  $p_1$ . If the indicator  $p_k = 0$  then  $x_k^* = 0$ , else  $x_k^* = 1$ .

## Problem Adapted

$$\begin{aligned}
 &\text{Get } F_n(M_0) \\
 &\text{Where } F_k(M) = \min \left\{ F_{k-1}(M) + Cp_k + (\lambda_k^{cm} \times Cc_k), \right. \\
 &\quad \left. F_{k-1}(M - v_k) + (\lambda_k^{sm} \times Cc_k) \right\} \\
 &\text{With } F_0(M) = 0 \quad \forall M
 \end{aligned}$$

Where:

$Cp_k$  is the maintenance preventive cost for equipment  $k$ ;

$Cc_k$  is the maintenance corrective cost for equipment  $k$ ;

$\lambda_k^{cm}$  is the failure rate for equipment  $k$  which was received preventive maintenance;

$\lambda_k^{sm}$  is the failure rate for equipment  $k$  which was not received preventive maintenance;

$v_k$  is the volume of the equipment  $k$  which was selected to the knapsack  $M$ .

## Important Steps of the Algorithm Developed

- 1 Calculating the boundary given by the reliability constraints;
- 2 Calculating the volume of equipments:

$$v_k = \delta \times (\lambda_k^{sm} - \lambda_k^{cm})$$

- 3 Calculating knapsack size:

$$M = \delta \times (SAIFI_{perm} - SAIFI_{min})$$



# Pseudocode

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## Pseudocode Knapsack Problem (n,M)

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- 1: **Calculate**  $SAIFI_{min}$   $SAIFI_{max}$
  - 2: **Calculate**  $v_k \forall k = 1..n$
  - 3: **knapsack(n,M)**  $\rightarrow p$
  - 4: **Calculate**  $F_0(M) = 0 \forall M$
  - 5: **For**  $k = 1..n$  **do**
  - 6:     **For**  $m = 1..M$  **do**
  - 7:          $F_k(M) = \min \{ F_{k-1}(M) + Cp_k + (\lambda_k^{cm} \times Cc_k),$   
 $F_{k-1}(M - v_k) + (\lambda_k^{sm} \times Cc_k) \}$
  - 8:     **End For**
  - 9: **End For**
  - 10: **OptimalSolution(p,M)**  $\rightarrow S$
- Return:** S
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# Studied Cases

## Instances

- Six instances were created for the problem:
  - 1 Instance with 30 equipments;
  - 2 Instance with 50 equipments;
  - 3 Instance with 100 equipments;
  - 4 Instance with 150 equipments;
  - 5 Instance with 300 equipments;
  - 6 Instance with 400 equipments;
- All instances were executed for only one period.

## SAIFI permitted

- For each instance five values of constraints were chosen;
- Calculating via Equation:

$$SAIFI_{\alpha} = SAIFI_{min} + (SAIFI_{max} - SAIFI_{min}) \times \alpha$$

where  $\alpha$  is 0.2, 0.4, 0.6, 0.8 and 1.0.

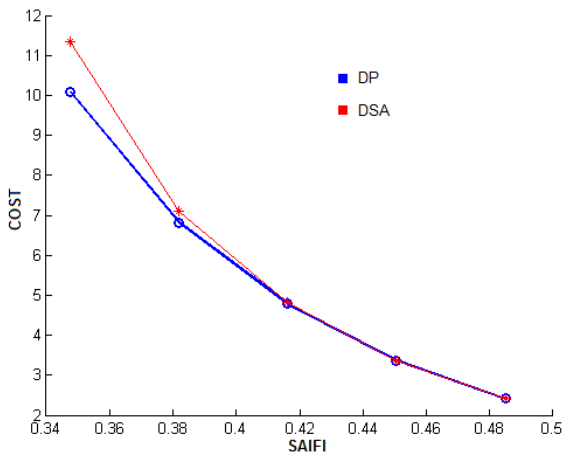
## Equipments Values

<b>Tipo</b>	<b>CMC</b>	<b>CMP</b>	<b>Mtx MP</b>	<b>CSM</b>	<b>Mtx SM</b>	<b>TF Initial</b>
Cable	0.06	0.03	0.92	0	1.08	0.02
infrastructure 1	0.94	0.47	0.79	0	1.26	0.05
infrastructure 2	0.94	0.47	0.79	0	1.26	0.05
Post 1	14.5	7.25	0.69	0	1.2	0.001
Post 2	14.5	7.25	0.69	0	1.2	0.001
Regulator	16	8	0.89	0	1.12	0.029
Recloser	1.2	0.6	0.91	0	1.28	0.015
Primary Pruning	2.05	1.025	0.95	0	1.51	0.05
Secondary Pruning	1.05	0.525	0.95	0	1.51	0.05
Transformer	1.692	0.846	0.95	0	1.51	0.01

# Results

Instance with 30 equipments:

	DP		DSA		
<b>SAIFI</b>	<b>Cost</b> (x 1000)	<b>Time</b> (s)	<b>Cost</b> (x 1000)	<b>Time</b> (s)	<b>Profit</b> (%)
0.3476	<b>10.0766</b>	<b>0.1560</b>	11.3455	1.2012	11.14
0.3819	<b>6.8217</b>	0.7020	7.1146	<b>0.2964</b>	4.11
0.4163	<b>4.7854</b>	1.6224	4.8152	<b>0.2184</b>	0.61
0.4506	<b>3.3699</b>	2.8548	<b>3.3699</b>	<b>0.1404</b>	0
0.4849	<b>2.4145</b>	4.6332	<b>2.4145</b>	<b>0.1716</b>	0

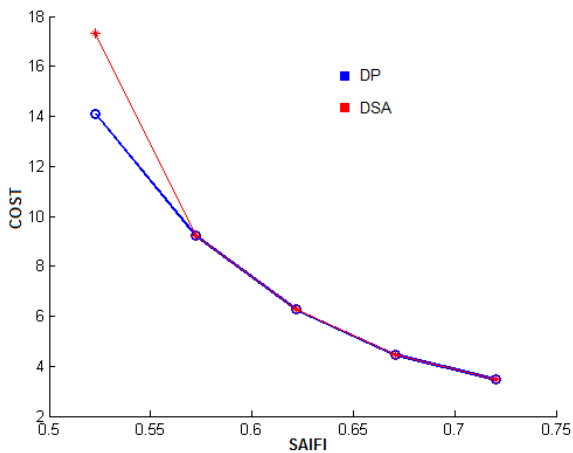


# Results

Instance with 50 equipments:

	DP		DSA		
<b>SAIFI</b>	<b>Cost</b> (x 1000)	<b>Time</b> (s)	<b>Cost</b> ( x 1000)	<b>Time</b> (s)	<b>Profit</b> (%)
0.5227	<b>14.0922</b>	<b>0.3276</b>	17.3370	1.2480	18.71
0.5722	<b>9.2147</b>	1.4196	9.2326	<b>0.3120</b>	0.19
0.6216	<b>6.2706</b>	3.5256	6.3004	<b>0.2964</b>	0.47
0.6710	<b>4.4370</b>	6.3804	<b>4.4370</b>	<b>0.4060</b>	0
0.7204	<b>3.4518</b>	10.1245	<b>3.4518</b>	<b>0.3276</b>	0

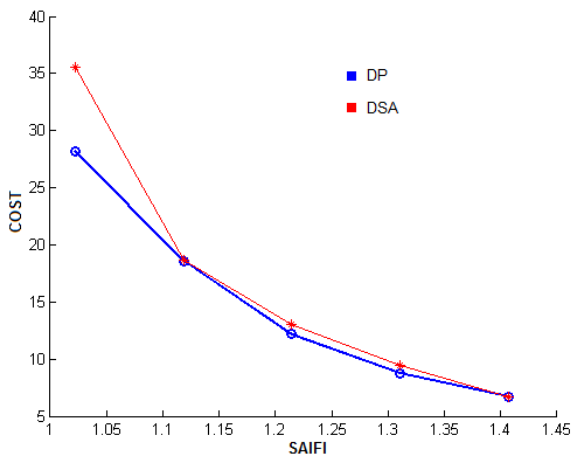




# Results

Instance with 100 equipments:

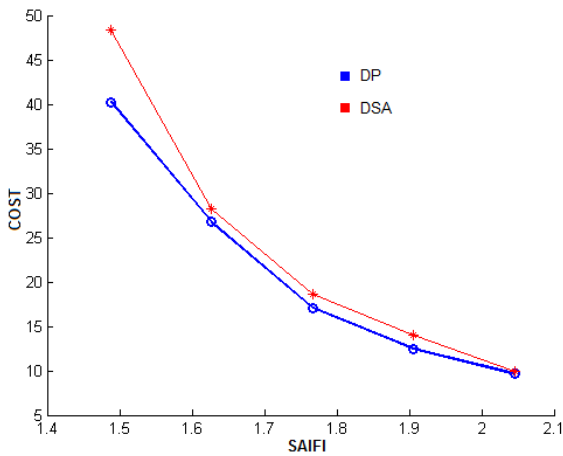
	DP		DSA		
<b>SAIFI</b>	<b>Cost</b> (x 1000)	<b>Time</b> (s)	<b>Cost</b> (x 1000)	<b>Time</b> (s)	<b>Profit</b> (%)
1.0221	<b>28.1844</b>	<b>0.9672</b>	35.5105	2.0124	20.63
1.1183	<b>18.5903</b>	4.9140	18.6201	<b>3.2136</b>	0.16
1.2144	<b>12.1651</b>	12.7141	12.9892	<b>1.0452</b>	6.34
1.3106	<b>8.7846</b>	22.1521	9.4412	<b>0.7020</b>	6.95
1.4068	<b>6.6941</b>	35.0690	6.7239	<b>0.9204</b>	0.44



# Results

Instance with 150 equipments:

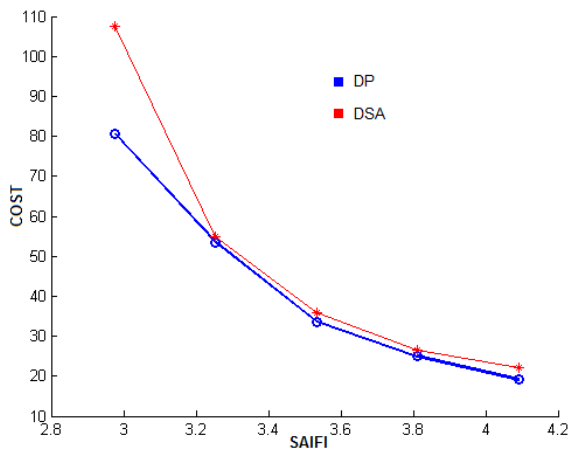
	DP		DSA		
<b>SAIFI</b>	<b>Cost</b> (x 1000)	<b>Time</b> (s)	<b>Cost</b> (x 1000)	<b>Time</b> (s)	<b>Profit</b> (%)
1.4878	<b>40.2492</b>	<b>2.1060</b>	48.3821	11.7157	16.80
1.6272	<b>26.7334</b>	10.7017	28.2550	<b>2.1216</b>	5.38
1.7665	<b>17.1013</b>	25.6564	18.5819	<b>6.6456</b>	7.96
1.9059	<b>12.4963</b>	47.1747	14.0188	<b>8.3461</b>	10.86
2.0452	<b>9.6598</b>	75.2237	9.9876	<b>13.3849</b>	3.28



# Results

Instance with 300 equipments:

	DP		DSA		
<b>SAIFI</b>	<b>Cost</b> (x 1000)	<b>Time</b> (s)	<b>Cost</b> (x 1000)	<b>Time</b> (s)	<b>Profit</b> (%)
2.9757	<b>80.4985</b>	<b>11.5285</b>	107.6273	290.8254	25.20
3.2543	<b>53.4667</b>	<b>52.2447</b>	54.9088	250.8466	2.61
3.5330	<b>33.5628</b>	<b>124.7384</b>	35.8796	1295.7478	6.45
3.8117	<b>24.9627</b>	<b>242.0356</b>	26.6641	1681.9546	6.38
4.4068	<b>19.1697</b>	398.0834	22.2148	<b>231.5548</b>	13.70

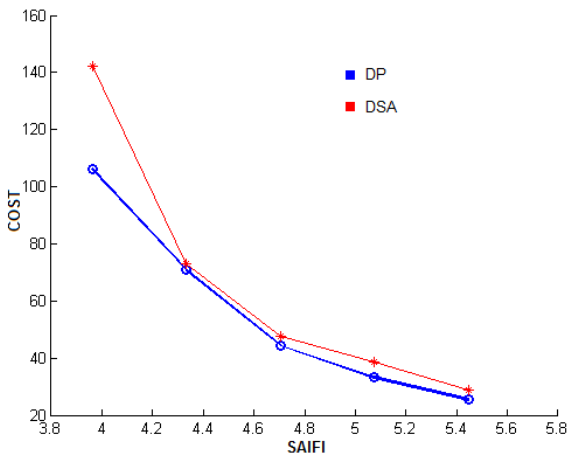


# Results

Instance with 400 equipments:

	DP		DSA		
<b>SAIFI</b>	<b>Cost</b> (x 1000)	<b>Time</b> (s)	<b>Cost</b> (x 1000)	<b>Time</b> (s)	<b>Profit</b> (%)
3.9625	<b>106.2709</b>	<b>25.7558</b>	142.0483	7258.3457	25.18
4.3336	<b>70.8304</b>	<b>111.8215</b>	72.9871	705.4124	2.95
4.7046	<b>44.3515</b>	<b>255.7480</b>	47.6116	2240.4517	6.84
5.0757	<b>33.1114</b>	<b>446.8805</b>	38.4557	1290.7457	13.89
5.4468	<b>24.4373</b>	<b>712.9090</b>	28.6017	5070.5072	14.55





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## Conclusions

- In all instances the dynamic programming algorithm achieved the best results;
- In small instances, the state space search algorithm achieved good results when the constraints were looser, but the results deteriorates when the number of equipment grows;
- Dynamic programming has maintained a standard result on the all values of time achieved, increasing as the number of equipments grows.
- The same not happened with the state space search algorithm, increasing a lot of the computational time.

## Future Works

As Future Works:

- Increase the period of optimisation;
- The development of equations dividing the problem in  $2^{HP}$  subproblems.

## Acknowledgements

- *CNPq*
- *CAPES*