

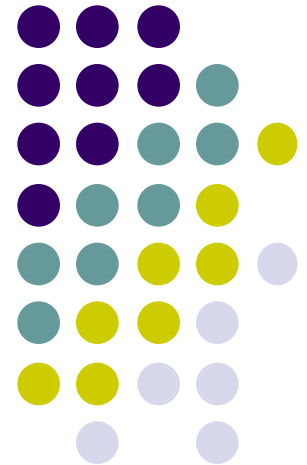
# Module 1

# Quantitative Reliability

# Analysis

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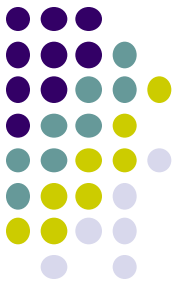
Chanan Singh  
Texas A&M University





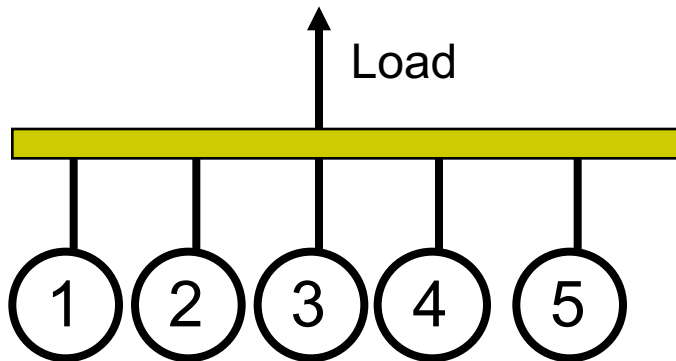
# Introduction to Quantitative Reliability

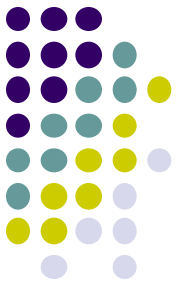
- Reliability relates to the ability of a system to perform its intended function
- Qualitative vs. quantitative concept of reliability.
- When quantitatively defined, reliability becomes a parameter that can be traded off with other parameters like cost
- Necessity of quantitative reliability : **decision making**
  - Ever increasing complexity of system design and operation
  - Evaluation of alternate design proposals
  - Cost competitiveness and cost-benefit trade off



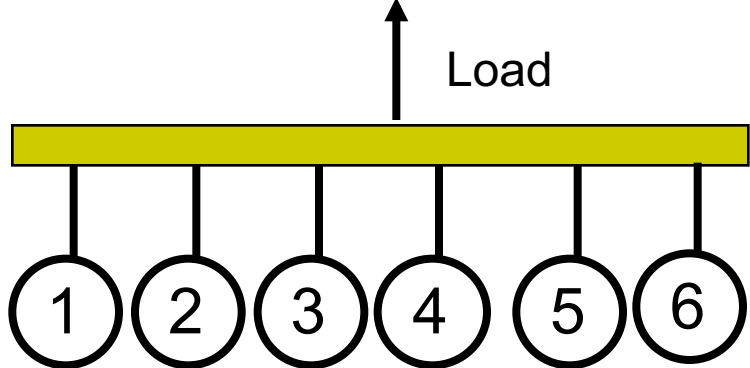
# Discussion Questions

- WE HAVE A LOAD OF 500 MW
- WHICH OF THE FOLLOWING ALTERNATIVES IS THE BEST CONSIDERING COST AND RELIABILITY?
  - 5 GENERATORS 100 MW EACH

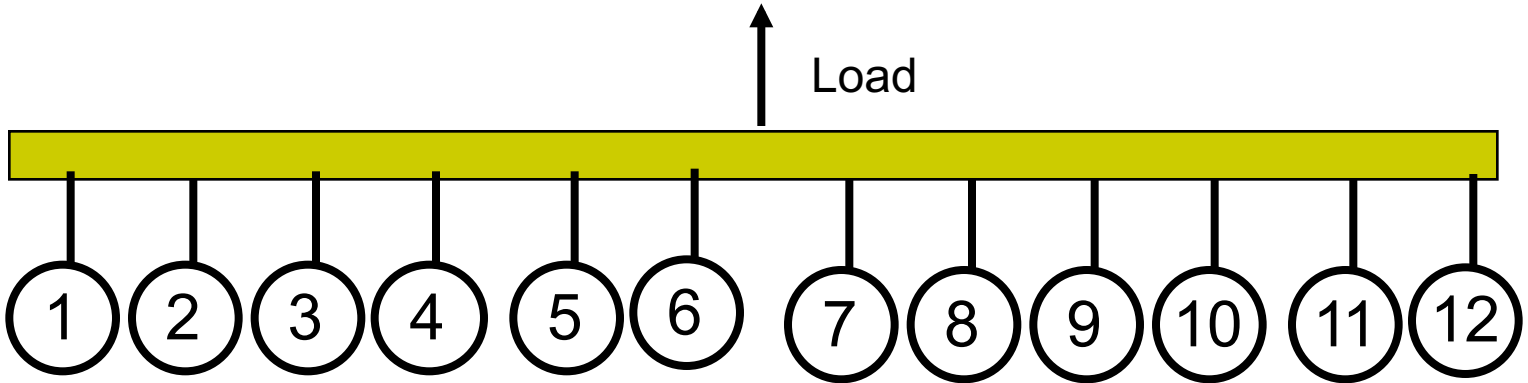




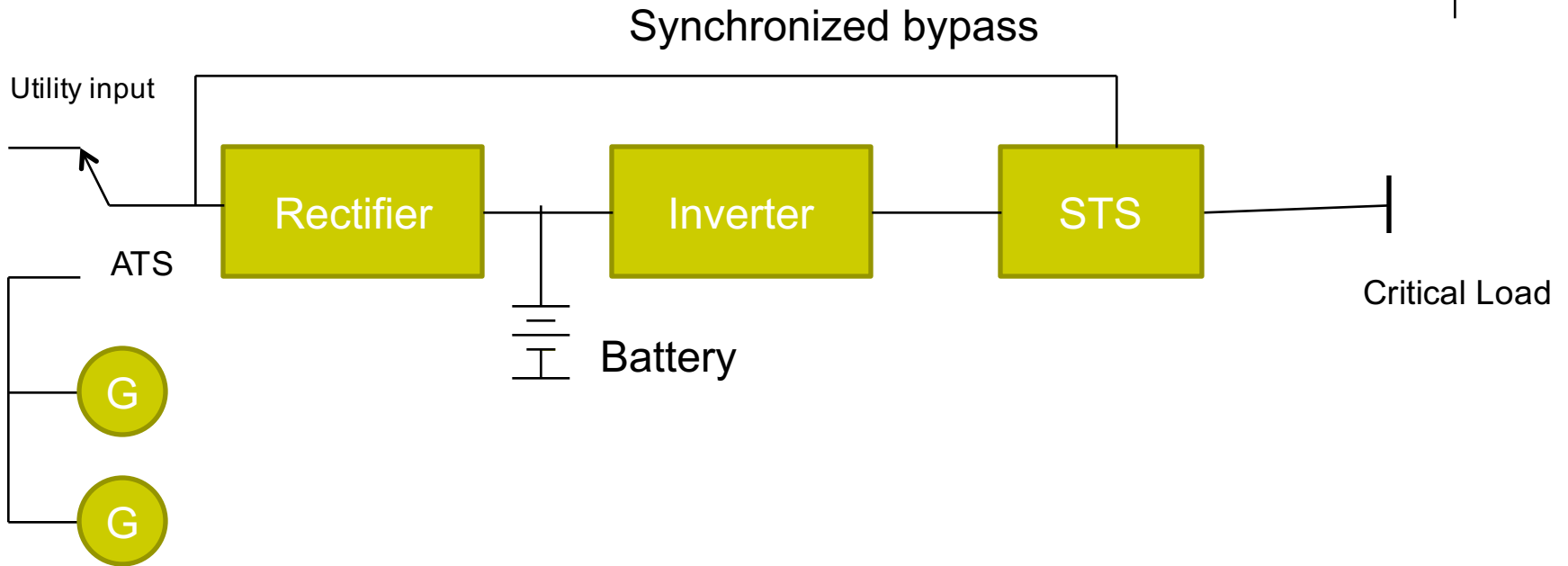
- 6 GENERATORS OF 100 MW EACH



-12 GENERATORS OF 50 MW EACH



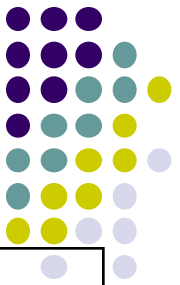
# Standby Power



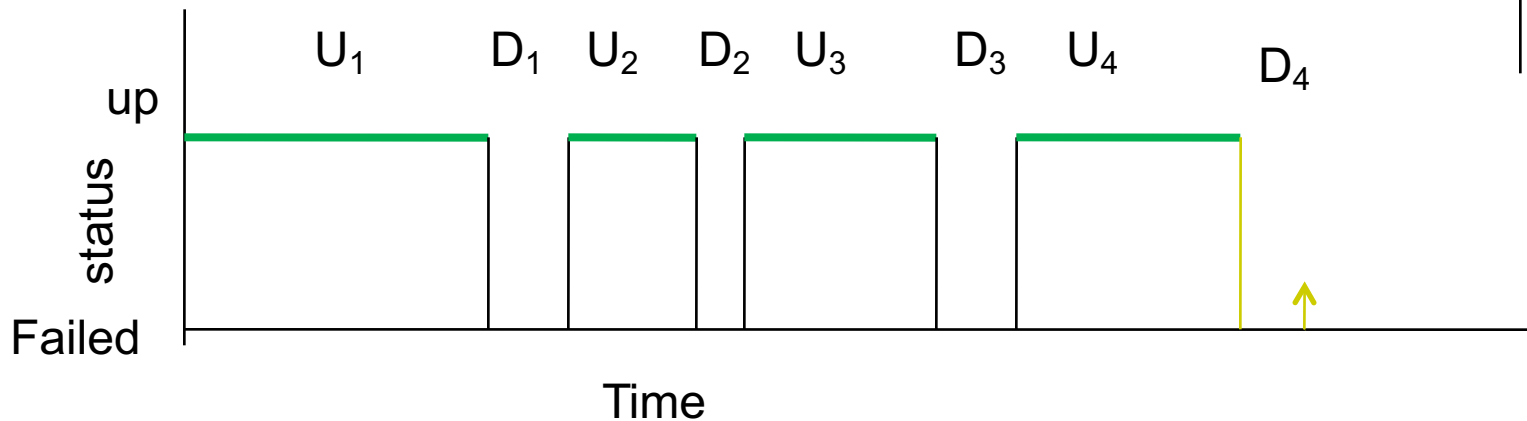


# Measures of Reliability Quantification

- Basic indexes
  - Probability of failure  
Long run fraction of time system is failed
  - Frequency of failure  
Expected or average number of failures per unit time
  - Mean duration of failure  
Mean duration of a single failure
- Other indexes can be generally obtained as a function of the above.



# Example of Indices



Up Times	Down Times	Cycle Time
$U_1 = 100 \text{ h}$	$D_1 = 10 \text{ h}$	110
$U_2 = 50 \text{ h}$	$D_2 = 5 \text{ h}$	55
$U_3 = 80 \text{ h}$	$D_3 = 6 \text{ h}$	86
$U_4 = 90 \text{ h}$	$D_4 = 4 \text{ h}$	94
<b>Total= 320 h</b>	<b>25h</b>	<b>320+25=345</b>

Estimate of probability of failure =  $25/(320+25) = .072$

Estimate of frequency of failure (FF) =  $4/345 = .011594 \text{ f/h} = 101.56 \text{ f/y}$

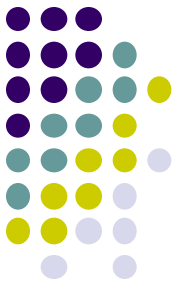
Estimate of mean down time(MDT) =  $25/4 = 6.25 \text{ h}$

Estimate of mean up time(MUT) =  $320/4=80 \text{ h}$

Estimate of mean cycle time=  $86.25 \text{ h} = \text{MUT} + \text{MDT} = 1/\text{FF}$



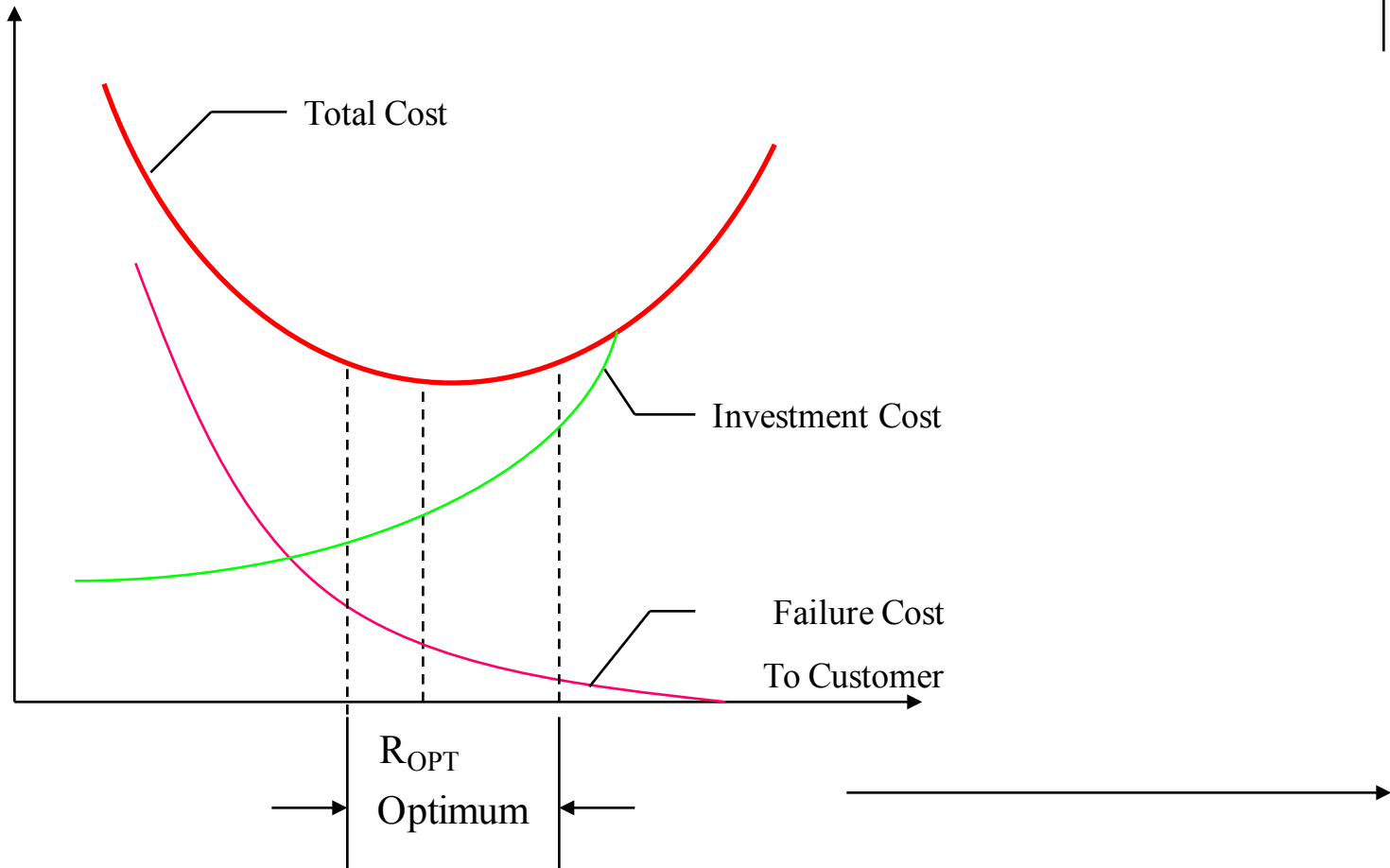
# Basic Approaches to Reliability Considerations

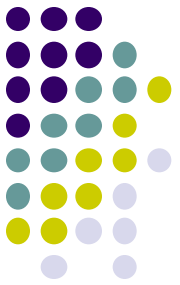


- Reliability implemented as a constraint
- Reliability worth included in overall cost optimization
- Multi-objective optimization with reliability as one of the objectives



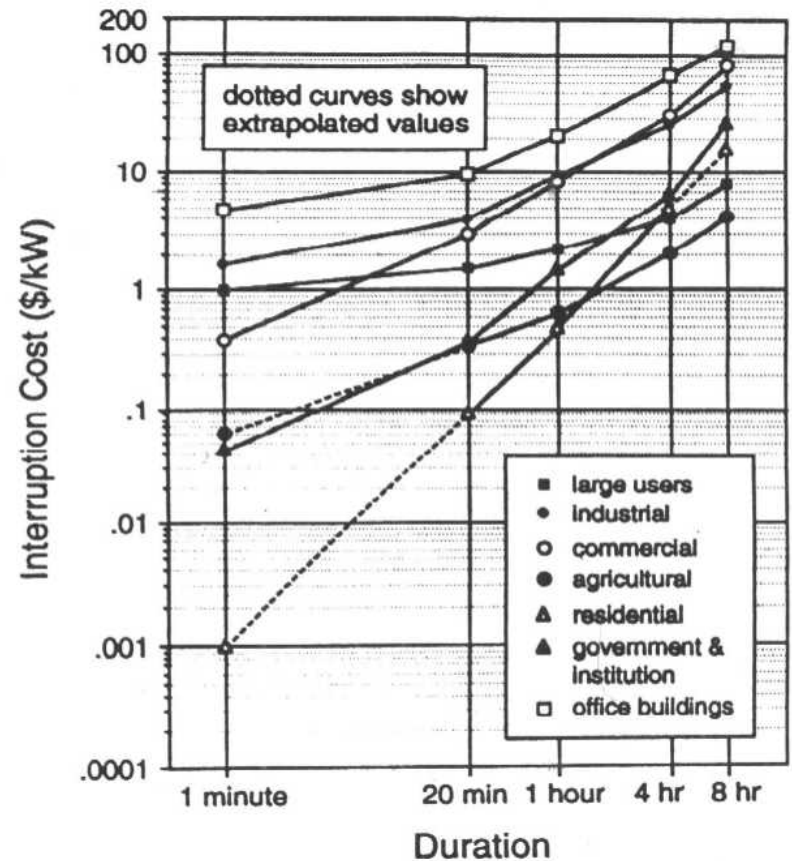
# Cost VS Reliability





# Reliability Cost/Worth Indices

- Reliability worth assessment provides the opportunity to incorporate cost analysis in system design.
- Two approaches:
  - As an objective: Achieve minimum total cost, which is the sum of investment, operating and customer interruption/failure costs.
  - As constraint: Expected power interruption or cost associated with failure is less than a pre-selected value.



# Interruption / Failure Cost



$$IC = \sum_{i=1}^n L_i \cdot \lambda_i \cdot c_i(d_i)$$

- $n$  : the number of load points
- $L_i$  : load requirement [kW]
- $\lambda_i$  : failure frequency [f/year]
- $c_i(d_i)$  : customer damage function [\$/kW]
- $d_i$  : outage duration [hours].

# Multi-objective Optimization & Pareto-optimality



- Most problems in nature have several (very often conflicting) objectives to be satisfied or optimized.
- **Multi-objective optimization** (or programming) also known as **multi-criteria** or **multi-attribute** optimization, is the process of simultaneously optimizing two or more conflicting objectives subject to certain constraints.

- Standard definition

$$\text{Min } \mathbf{f} = [f_1(x), f_2(x), \dots, f_n(x)]$$

subject to  $x \in S$  (constraints)

where each  $f_1(x)$  is an objective function

- One solution often employed is to optimize a weighted objective function

$$\text{Min } \mathbf{f} = [w_1 f_1(x) + w_2 f_2(x), \dots, w_n f_n(x)]$$

subject to  $x \in S$  (constraints)

where each  $f_1(x)$  is an objective function

- An other technique is to pick one objective as the primary objective to be optimized and transform the remaining objectives into constraints.

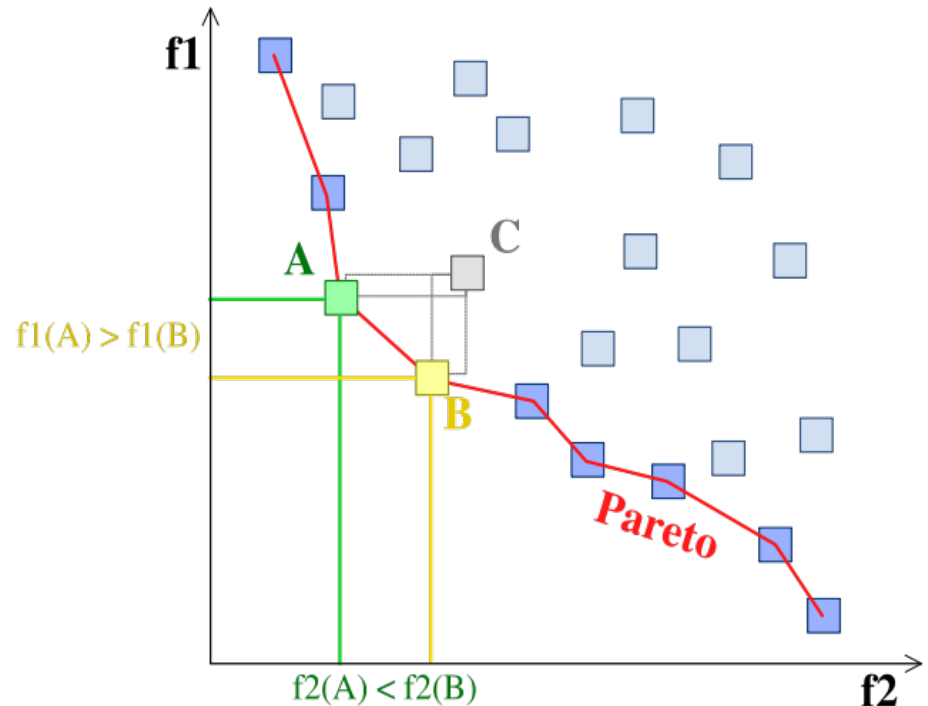


- In multi-objective optimization an other approach is to find Pareto-optimal solutions.
- No part of Pareto optimal solution can be improved without making some other part worse.
- This approach is useful when it is difficult to formulate a global objective function.

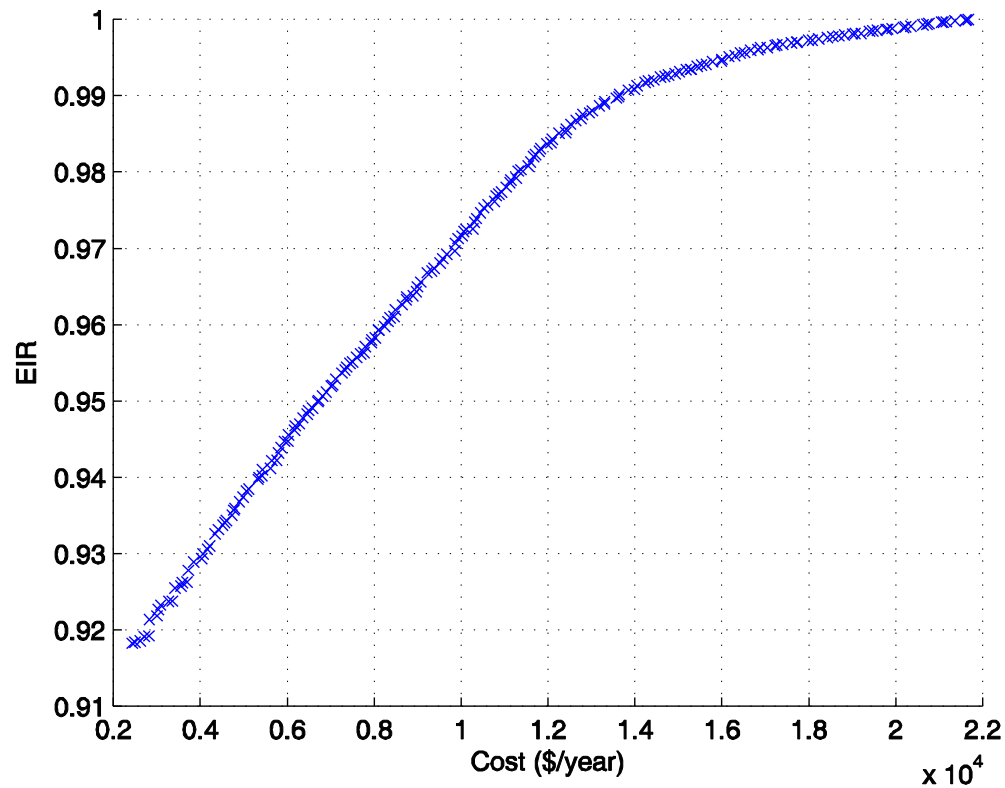


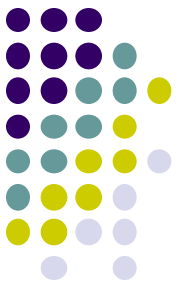
# Pareto Frontier

- See opposite the objective space of  $f_1$  and  $f_2$
- Given that lower values are preferred to higher values, point **C** is not on the Pareto Frontier because it is dominated by both point **A** and point **B**; and Points **A** and **B** are non-inferior.
- A vector is said to be dominated if other vectors of system variables can be found that have improved values of any function  $f_i$  without creating a lower value in any objectives in **f**.



# An Example: Tradeoff Curve Between Reliability & Cost





# Question Reformulated

- WE HAVE A LOAD OF 500 MW
- WHAT IS THE PROBABILITY NOT SUPPLYING THE LOAD(loss of load) IN THE FOLLOWING SCENARIOS ?
  - 5 GENERATORS 100 MW EACH
  - 6 GENERATORS OF 100 MW EACH
  - 12 GENERATORS OF 50 MW EACH
- ASSUME PROBABILITY OF FAILURE OF EACH GENERATOR IS 0.1 IN ALL CASES.



# References:



1. **C. Singh & R. Billinton, System Reliability Modelling and Evaluation, Hutchinson, London, 1977. (you can download from my website)**
2. **J. Endrenyi, Reliability Modeling in Electric Power Systems, John Wiley, 1978**
3. **R. Billinton & R. Allan, Reliability Evaluation of Power Systems, Plenum Press, 1984**
4. **Course Notes by C. Singh**
5. **Selected papers**