



Next Generation Modeling and Optimization of Integrated Site Designs

March 2009

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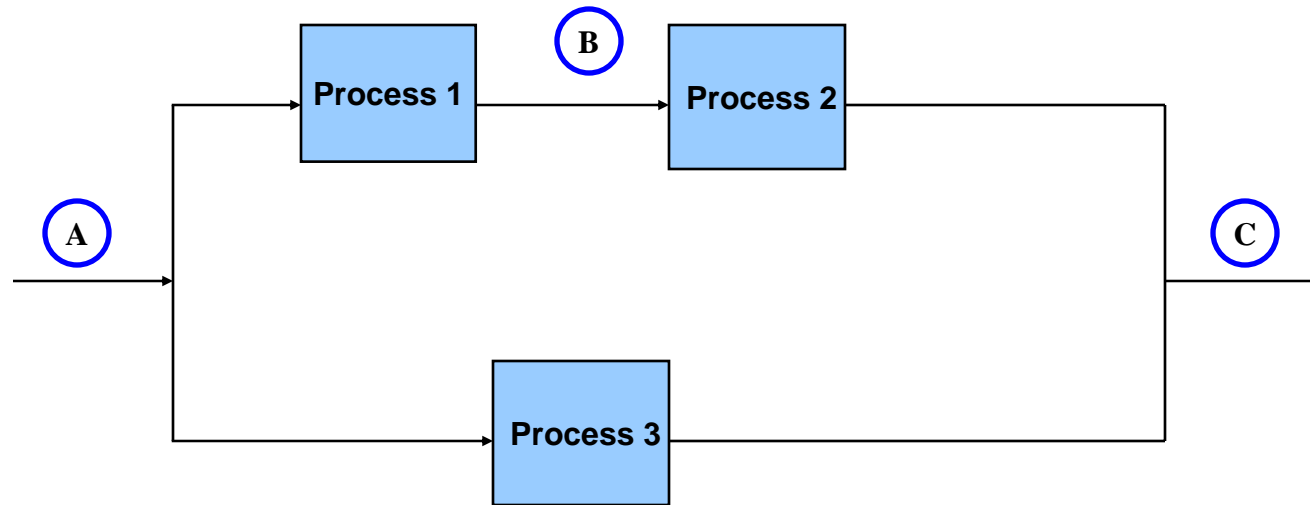
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Objective:

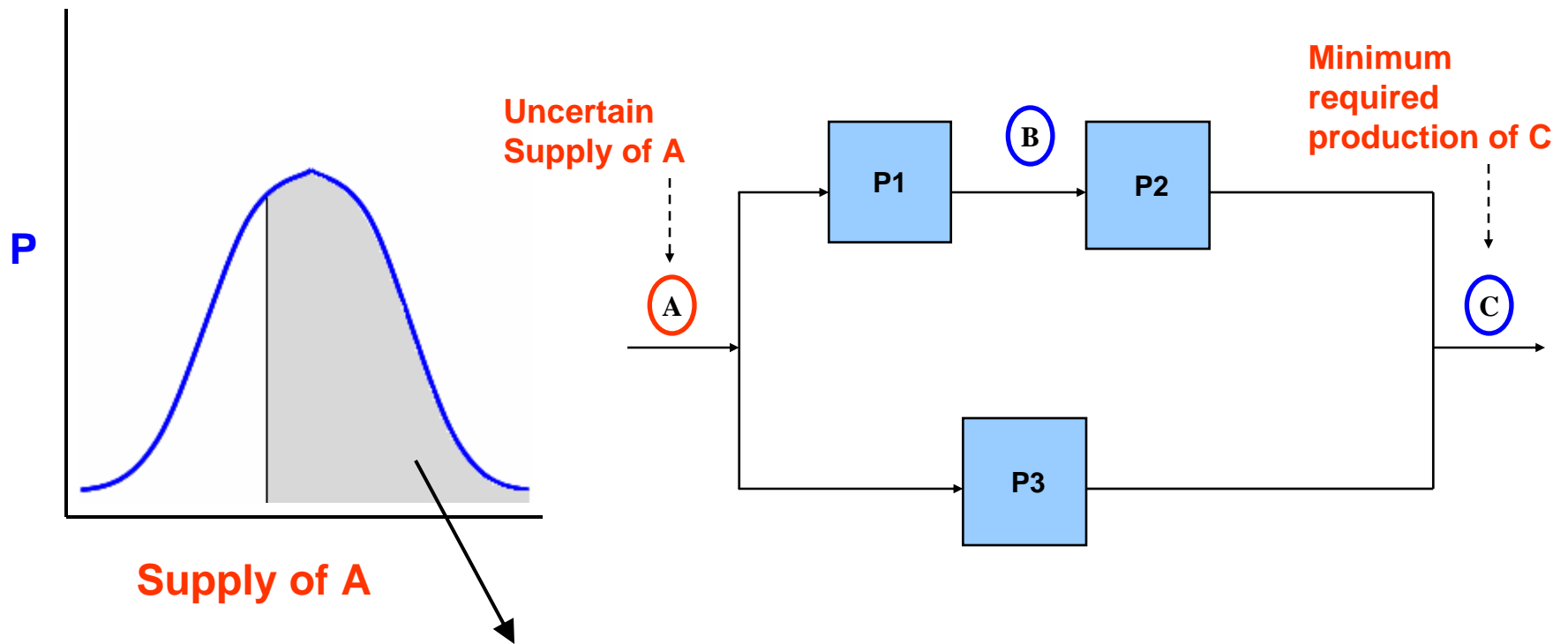
Present a systematic method for optimizing the long term availability of an integrated site.

Model an integrated site as a network of processes



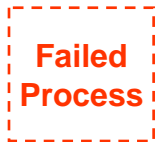
Processes in the network are subject to:

- **Continuous uncertain** (e.g. supply of A, demand of C)
- **Discrete uncertainties** (e.g. failure of process 3)

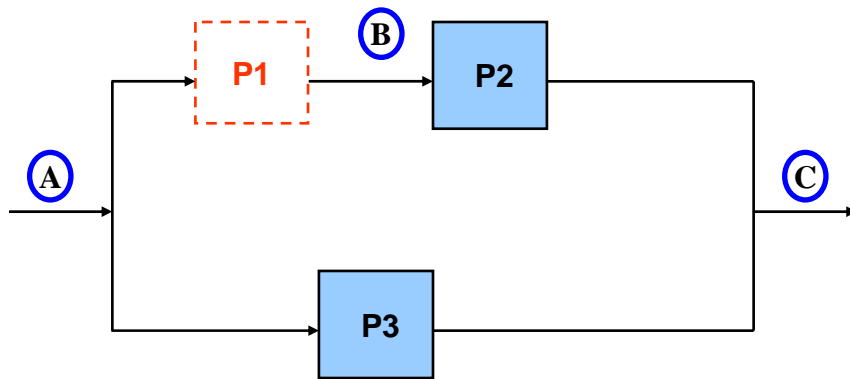


Stochastic Flexibility (SF) represents area of feasible operation under probability distribution

e.g. SF = 0.67



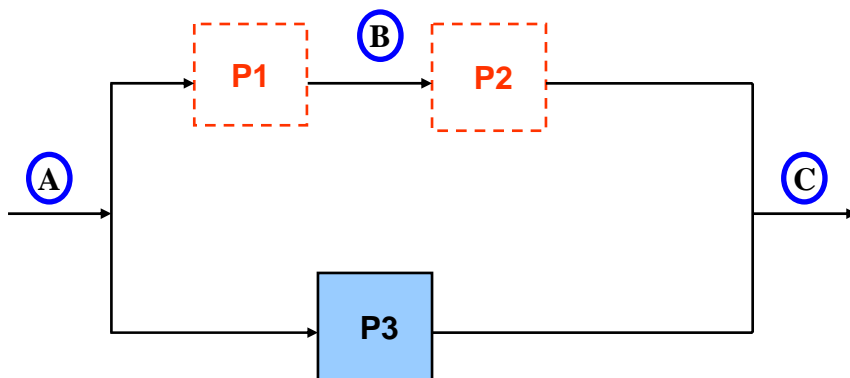
A discrete state is described by sets of failed and active processes



State 1

Active Processes = { P2, P3 }

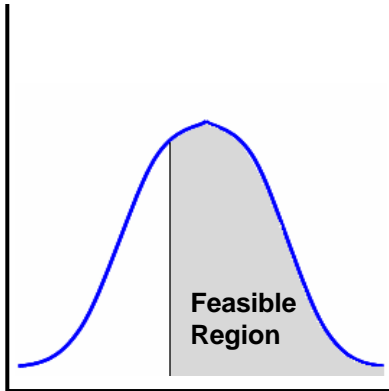
Failed Processes = { P1 }



State 2

Active Processes = { P3 }

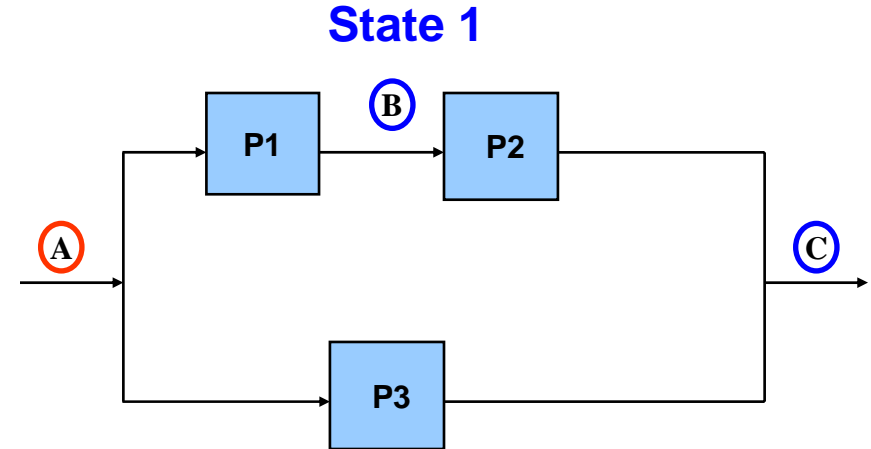
Failed Processes = { P1, P2 }



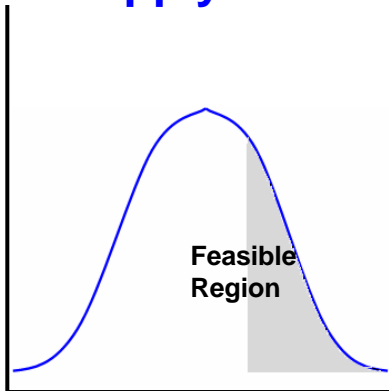
Define:

p_1 : probability of finding system in State 1

SF_1 : Stochastic Flexibility in State 1

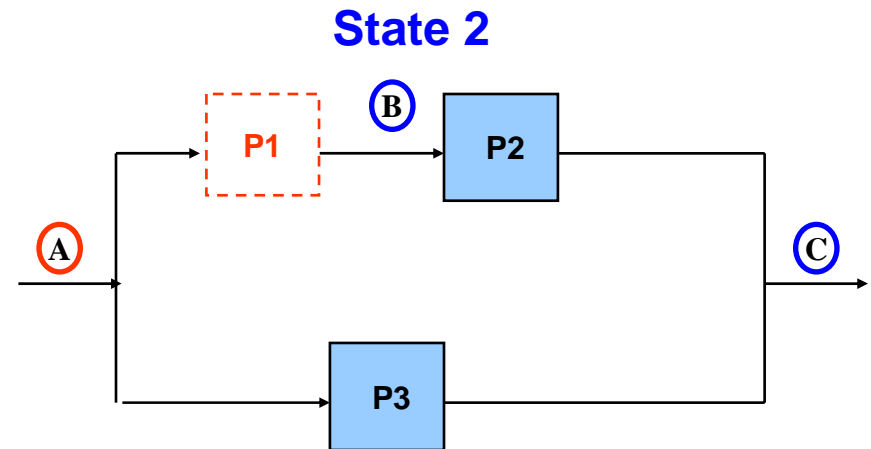


Supply of A



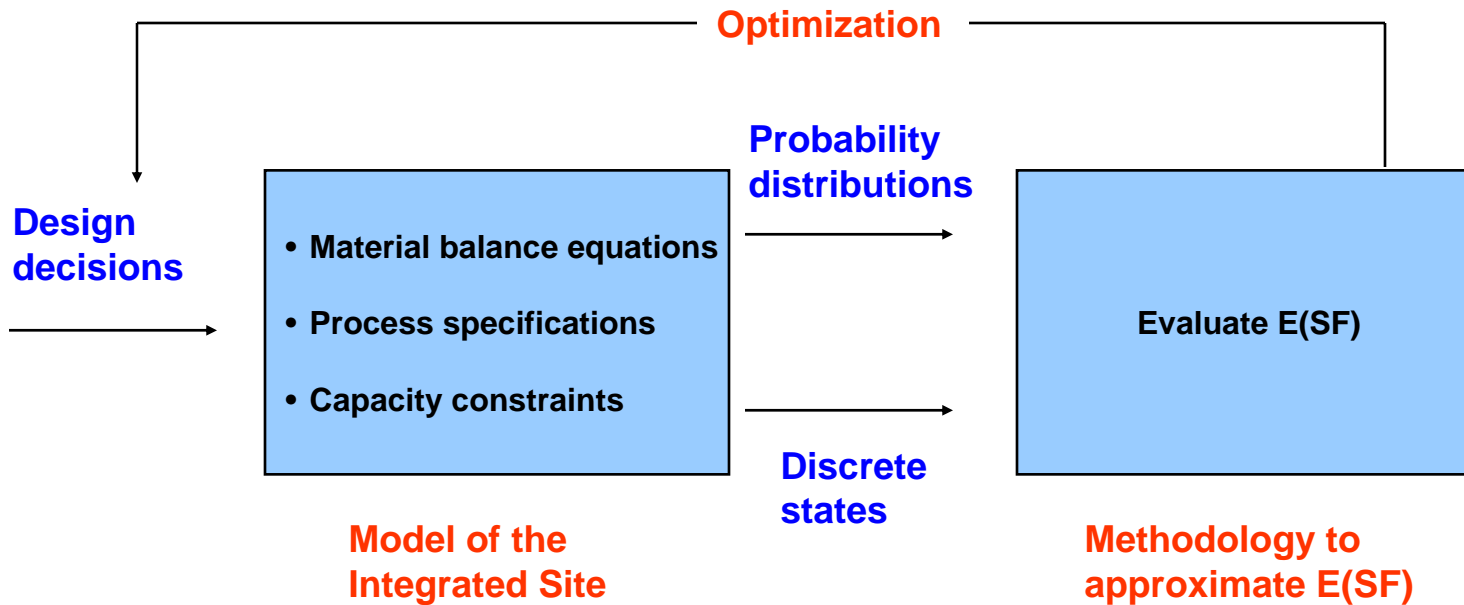
p_2 : probability of finding system in State 2

SF_2 : Stochastic Flexibility in State 2



$$E(SF) = p_1 SF_1 + p_2 SF_2$$

Mathematical programming approach requires a process model and methodology for evaluating E(SF)



Maximizing E(SF) = Maximizing probability of feasible operation

= Maximizing long term availability

Given:

A network of processes and their capacities

A set of products produced in the network

A set of possible failures, failure rates & repair rates

An external demand of products and external supply of raw materials

A set of uncertain parameters and their probability distributions

A maximum amount of resources available for design decisions

Determine:

Location of storage tanks

Maximum storage capacity and steady state inventory levels

Process capacity expansions

In order to maximize the expected stochastic flexibility

Material balance equations

$$h(x, d, \theta, \alpha, y) = 0$$

Process capacity

Intermediate storage capacity

Process specifications

$$g(x, d, \theta, \alpha, y) \leq 0$$

x: continuous process variables (flows)

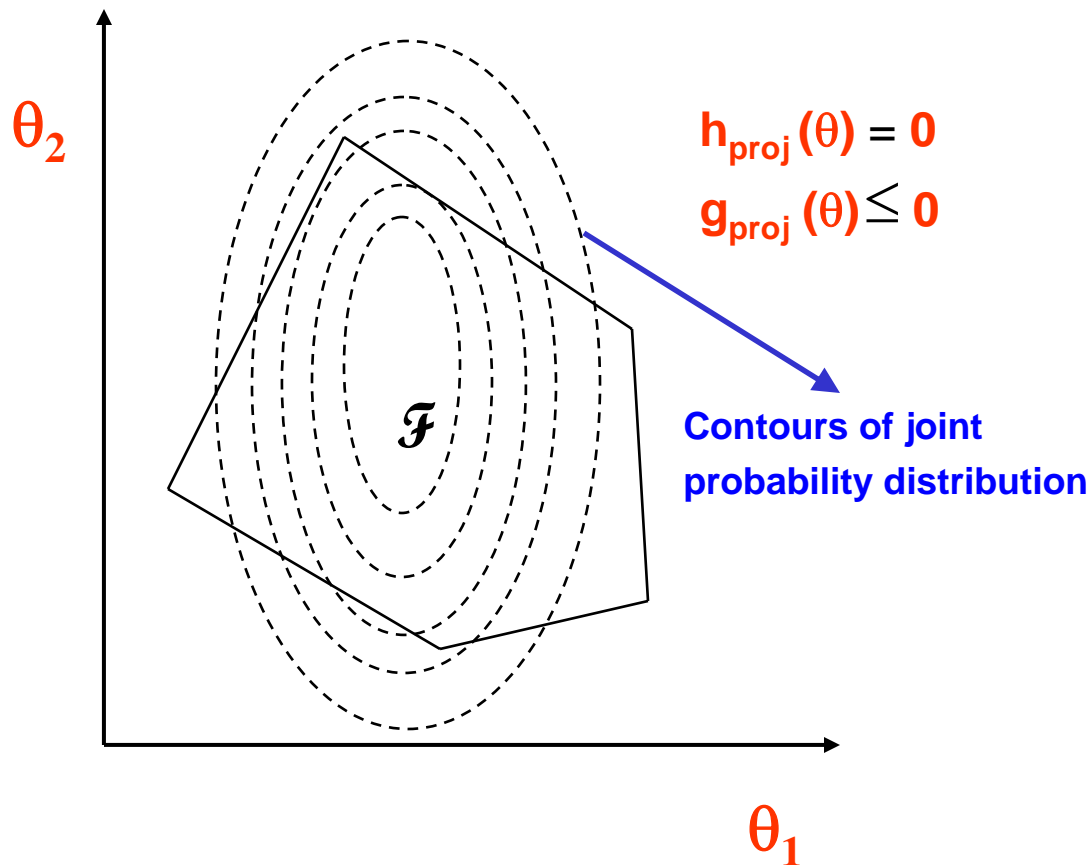
d: integrated design variables (intermediate storage, etc.)

θ : uncertain parameters (demand, supply, etc.)

α : deterministic parameters (unit ratio coefficients, etc.)

y: 0 – 1 parameters; 1 – active process, 0 – failed process

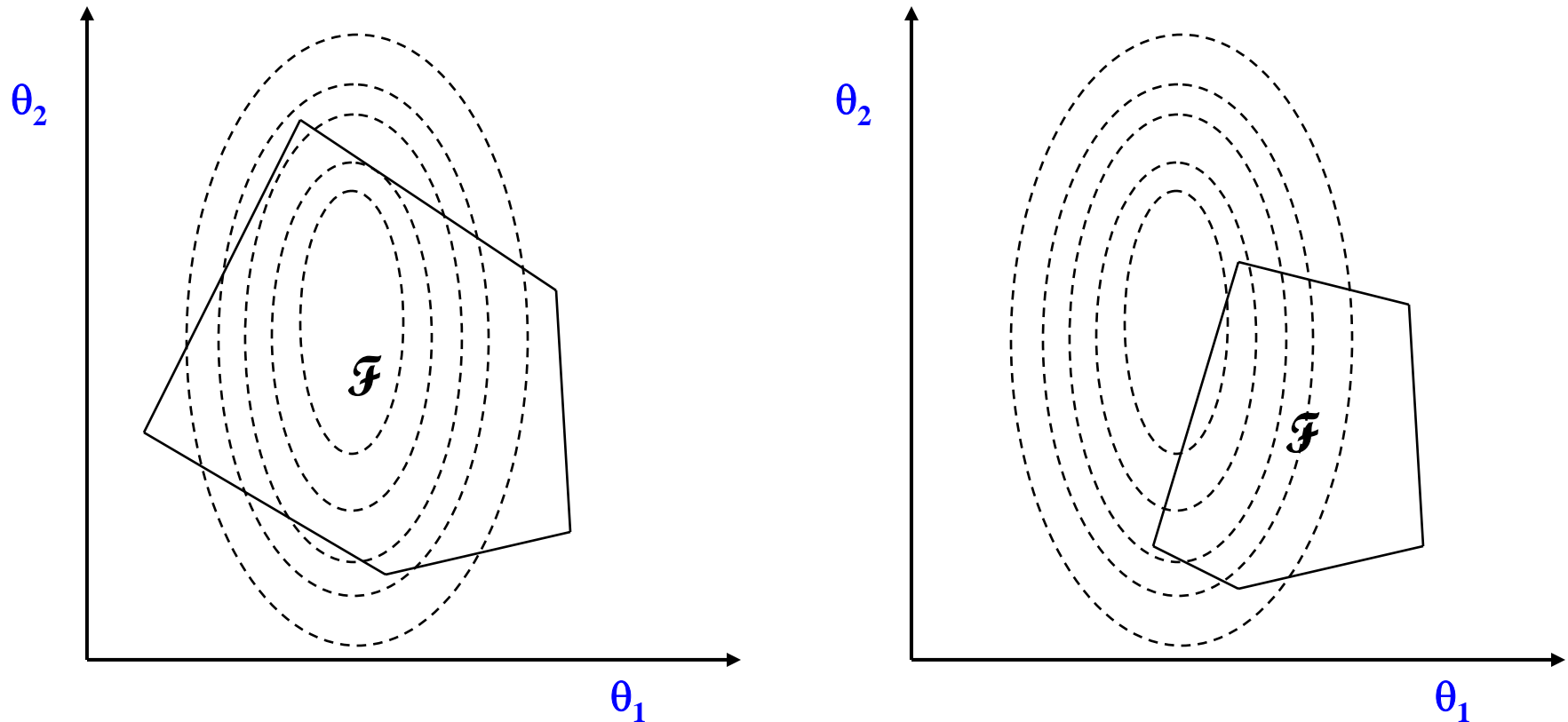
- Consider fixed y (one discrete state)
- Consider fixed d (fixed integrated design)
- Project x onto θ space
- Consider two uncertain parameters $\theta_1 \theta_2$



Stochastic Flexibility (SF) :

**Area under joint
distribution contained in \mathcal{F}**

Impact of discrete states on SF



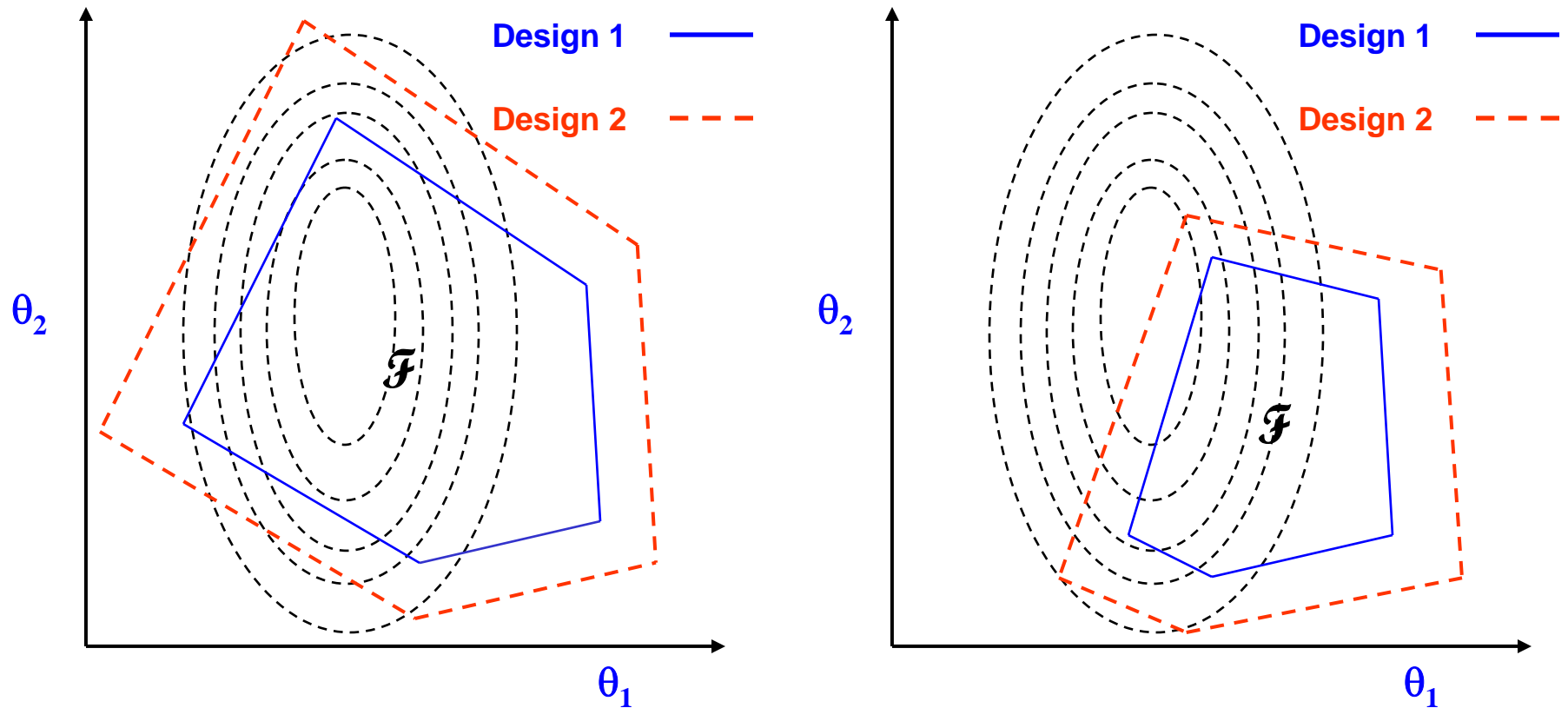
Discrete State A



Discrete State B

Equipment failure

$$SF_A > SF_B$$

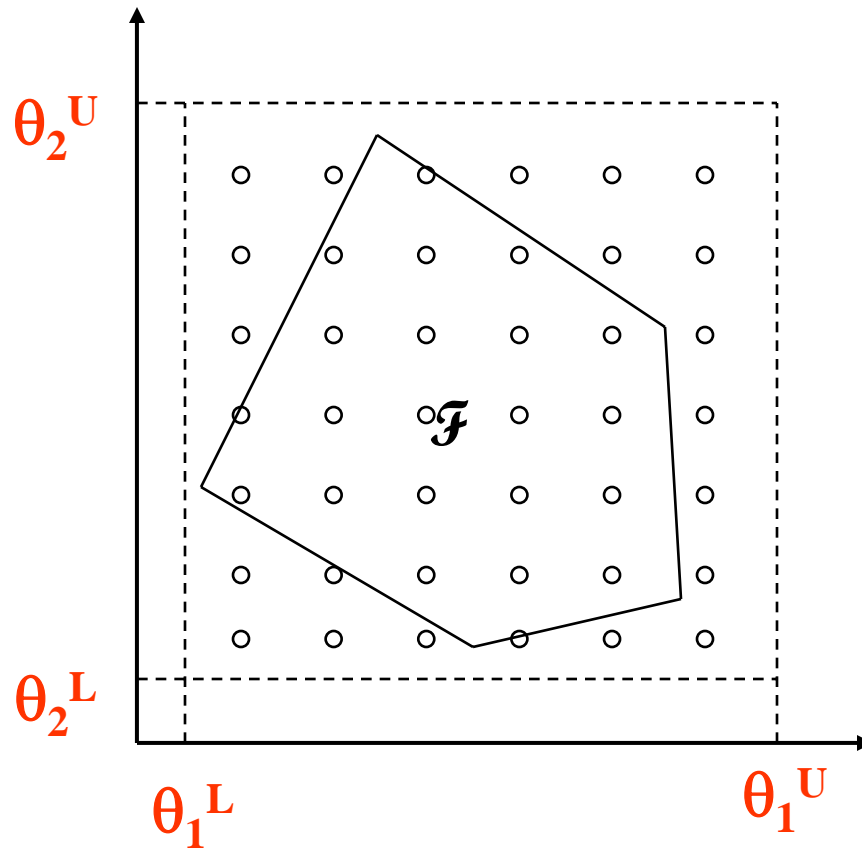


Discrete State A

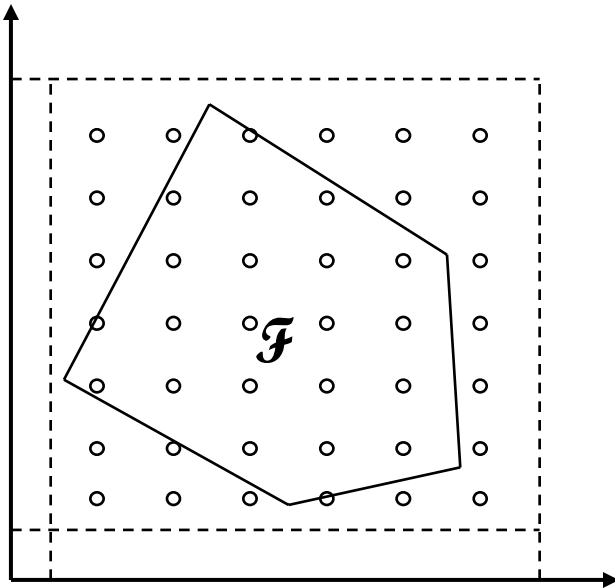


Discrete State B

Equipment failure



- Determine upper and lower bounds (e.g. $\mu \pm 4\sigma$)
- Use (Gaussian) quadrature points
- Use k_i as the index of quadrature points for θ_i



Assume two uncertain parameters

k_1 refers to the k -th discrete value in θ_1

k_2 refers to the k -th discrete value in θ_2

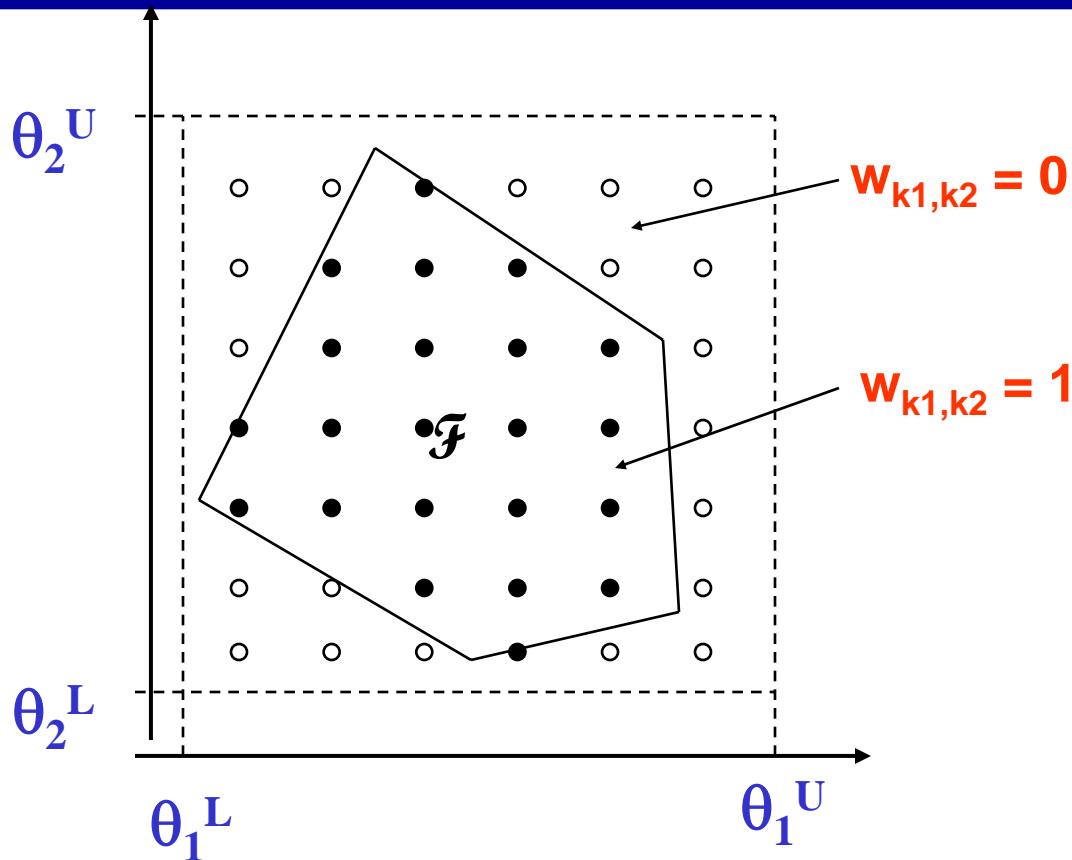
Each discrete point corresponds to a pair (k_1, k_2)

Use the following quadrature formula to assign a probability $\gamma_{k_1 k_2}$ to each point (k_1, k_2)

$$\gamma_{k_1 k_2} = \omega_{k_1} \omega_{k_2} \frac{\theta_1^U - \theta_1^L}{2} \frac{\theta_2^U - \theta_2^L}{2} \times j(\theta_{1_{k_1}}, \theta_{2_{k_2}})$$

$j(\theta_1, \theta_2)$ is the joint probability distribution

ω_{k_i} is the weight of the k -th quadrature point of the i -th uncertain parameter



Process model equations for each (k_1, k_2) point:

$$h_{k_1 k_2}() = 0$$

$$g_{k_1 k_2}() \leq 0$$



$$h_{k_1 k_2}() = 0$$

$$(1) \quad g_{k_1 k_2}() \leq u_{k_1 k_2}$$

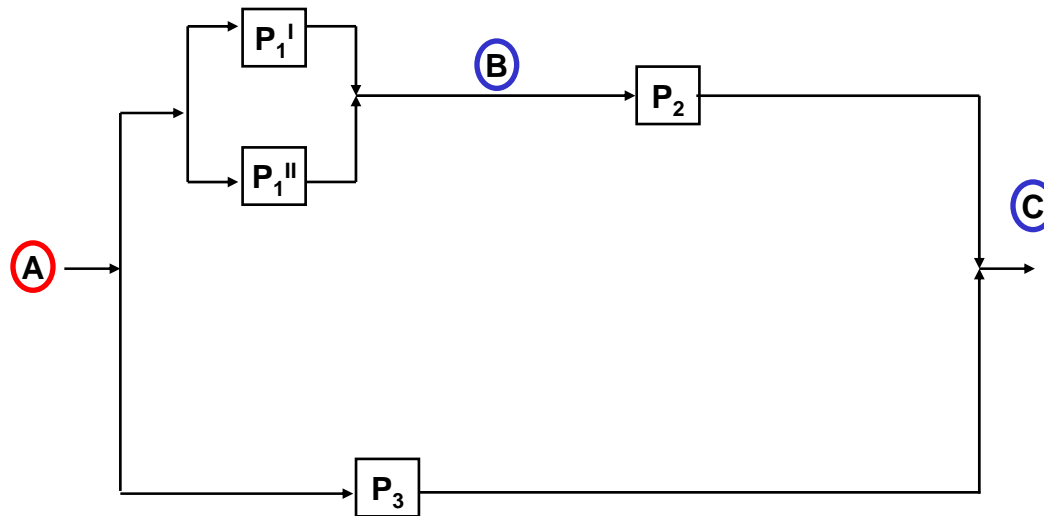
$$(2) \quad u_{k_1 k_2} \leq M(1 - w_{k_1 k_2})$$

Introduce slack variable u for right hand side of constraints (1)

u can be positive only if a binary w is equal to zero (2)

$w = 1$ indicates feasible point $w = 0$ indicates infeasible point

Case Study from Straub & Grossmann 1990



Uncertain supply of A

Uncertain demand of C

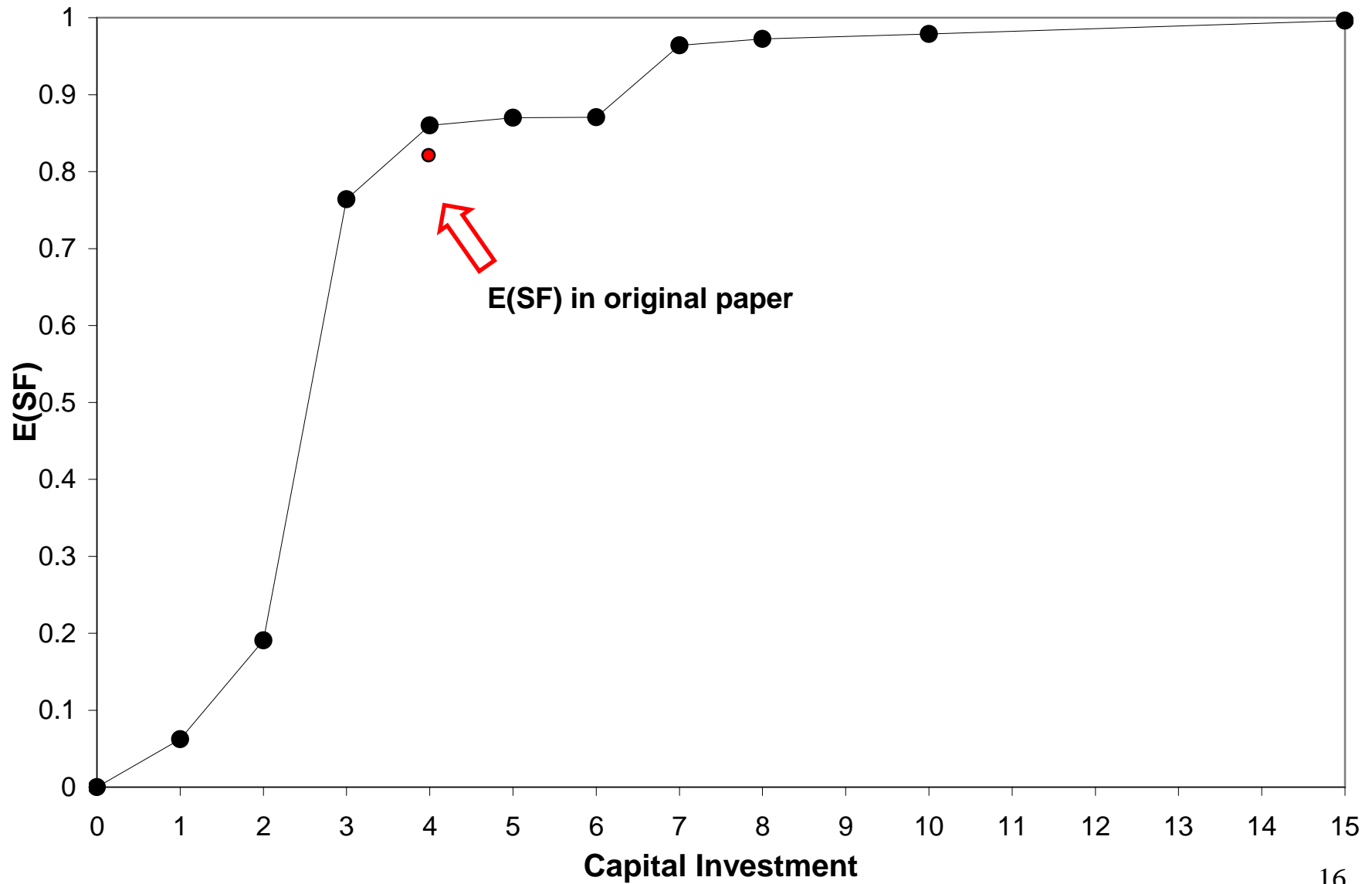
Plants subject to random failures

Base case (Grossmann and Straub 1990):

- All plants installed
- No storage
- No capacity expansions

$E(SF) = 0.804$ present approach
 $= 0.813$ Original paper

Case Study from Straub & Grossmann 1990



Algorithmic techniques such as upper and lower bounding are required to deal with large number of discrete states.

Optimization results will be validated using Stochastic modeling.

The long term goal of the project is to come up with a general tool applicable in other areas such as supply chain.