



KUNGL  
TEKNISKA  
HÖGSKOLAN

# International Master Program

## in System-on-Chip Design

**Evaluation Techniques**

# Two approaches

- Qualitative evaluation
  - aims to identify, classify and rank the failure modes, or event combinations that would lead to system failures
- Quantitative evaluation
  - aims to evaluate in terms of probabilities the attributes of dependability (reliability, availability, safety)

# Common dependability measures

- failure rate
- mean time to failure
- mean time to repair
- mean time between failures
- fault coverage

# Failure rate

- failure rate
  - expected # of failures per time-unit
  - example
    - 1000 controllers working at  $t_0$
    - after 10 hours: 950 working
    - failure rate for each controller:  
0.005 failures / hour  
 $(50 \text{ failures} / 1000 \text{ controllers}) / 10 \text{ hours}$

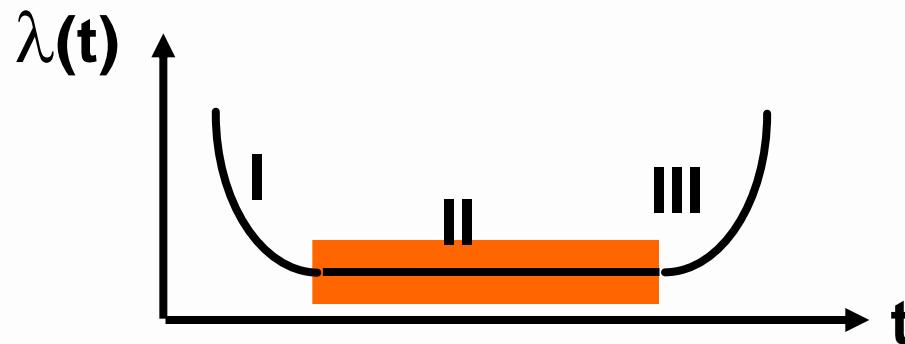
# Failure rate and reliability

Reliability  $R(t)$  is the conditional probability that the system will perform correctly throughout  $[0, t]$ , given that it worked at time 0

$$R(t) = \frac{N_{operating}(t)}{N_{operating}(t) + N_{failed}(t)}$$

# Failure rate

- typical evolution of  $\lambda(t)$  for hardware:



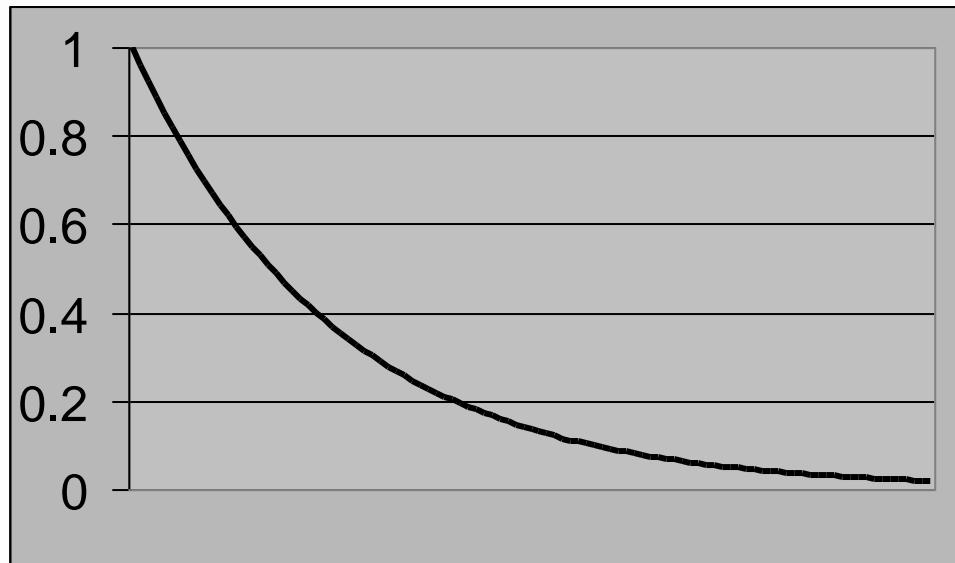
- bathtub: I infant mortality, II useful life, III wear-out
- for useful life period  $\lambda = \text{constant}$ , the reliability is given by

$$R(t) = e^{-\lambda t}$$

# Exponential failure law

$$R(t) = e^{-\lambda t}$$

If  $\lambda$  is constant,  $R(t)$  varies exponentially as a function of time



# Time varying failure rate

- Failure rate is not always constant
  - software failure rate decreases as package matures
- Weibull distribution:

$$z(t) = \alpha \lambda (\lambda t)^{\alpha-1}$$

- if  $\alpha=1$ , then  $z(t) = \text{constant} = \lambda$
- if  $\alpha>1$ , then  $z(t)$  increases as time increases
- if  $\alpha<1$ , then  $z(t)$  decreases as time increases

$$R(t) = e^{-(\lambda t)^\alpha}$$

# Failure rate calculation

- determined for components
  - systems: combination of components
  - $\lambda$  of the system = sum of  $\lambda$  of the components
- determine  $\lambda$  experimentally
  - slow
    - e.g. 1 failure per 100 000 hours (=11.4 years)
  - expensive
    - many components required for significance
- use standards for  $\lambda$

# MTTF

- MTTF: **mean time to failure**
  - expected time until the first failure occurs
- If we have a system of  $N$  identical components and we measure the time  $t_i$  before each component fails, then MTTF is given by

$$MTTF = \frac{1}{N} \cdot \sum_{i=1}^N t_i$$

# MTTF

MTTF is defined in terms of reliability as:

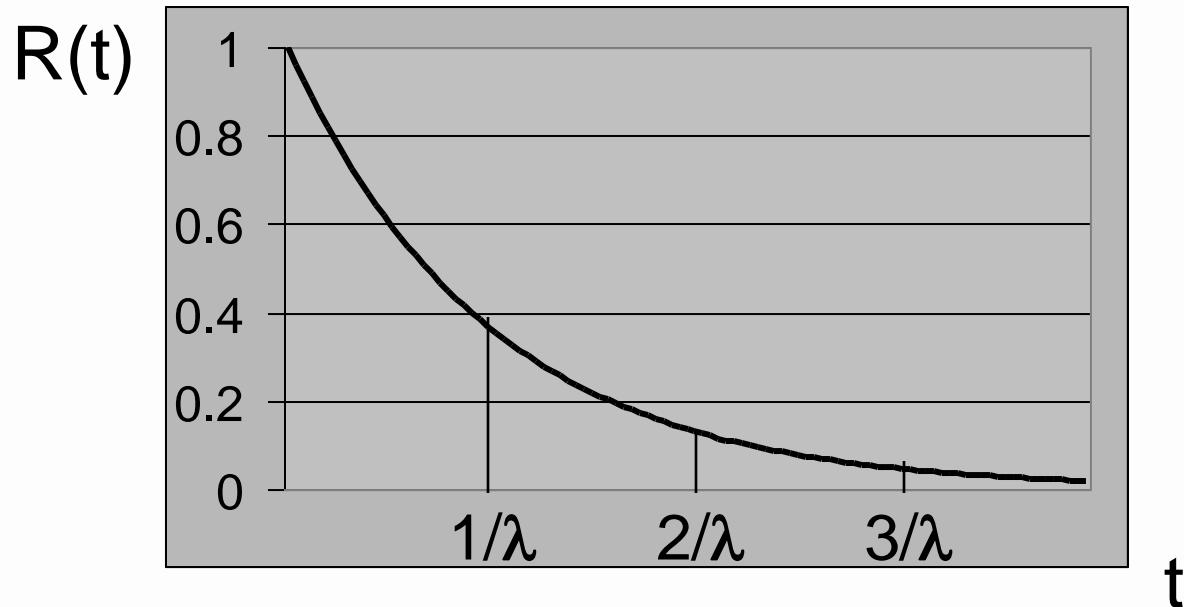
$$MTTF = \int_0^{\infty} R(t)dt$$

If  $R(t)$  obeys the exponential failure law, then  
MTTF is the inverse of the failure rate:

$$MTTF = \int_0^{\infty} e^{-\lambda t} dt = \frac{1}{\lambda}$$

# MTTF

$$R(t) = e^{-\lambda t}$$



# MTTF

- MTTF is meaningful only for systems which operate without repair until they experience a failure
- Most of mission-critical systems undergo a complete check-up before the next mission
  - all failed redundant components are replaced
  - system is returned to fully operational state
- When evaluating reliability of such system, mission time rather than MTTF is used

# MTTR

- MTTR: **mean time to repair**
  - expected time until repaired
- If we have a system of  $N$  identical components and  $i_{th}$  component requires time  $t_i$  to repair, then MTTR is given by

$$MTTR = \frac{1}{N} \cdot \sum_{i=1}^N t_i$$

# MTTR

- difficult to calculate
- determined experimentally
- normally specified in terms of repair rate  
repair rate  $\mu$ , which is the average number  
of repairs that occur per time period

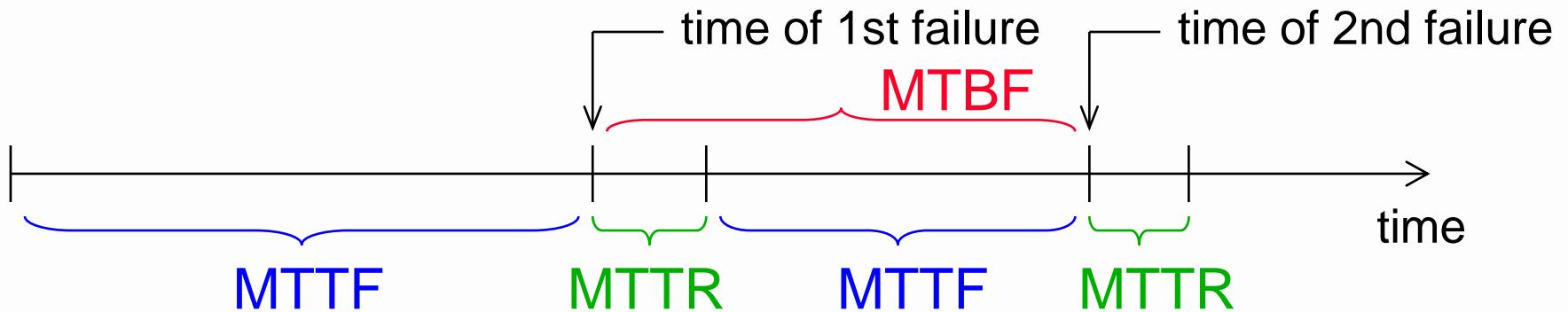
$$MTTR = \frac{1}{\mu}$$

# MTTR

- Low MTTR requirement implies high operational cost
  - if hardware spares are kept on site and the site is maintained 24hr a day, MTTR=30min
  - if the site is maintained 8hr 5 days a week, MTTR = 3 days
  - if system is remotely located MTTR = 2 weeks

# MTBF

- MTBF: **mean time between failures**
  - functional + repair
  - $MTBF = MTTF + MTTR$
  - small time difference:  $MTBF \approx MTTF$
  - conceptual difference



# Fault coverage

- Fault detection coverage is the conditional probability that, given the existence of a fault, the fault is detected
- Difficult to calculate
- Usually computed as

$$C = \frac{\text{number of faults which can be detected}}{\text{total number of faults}}$$

# Example

- Suppose your circuit has 10 lines and you use single-stuck at fault as a model
- Then the total number of faults is 20
- Suppose you have 1 undetectable fault
- Then the coverage is

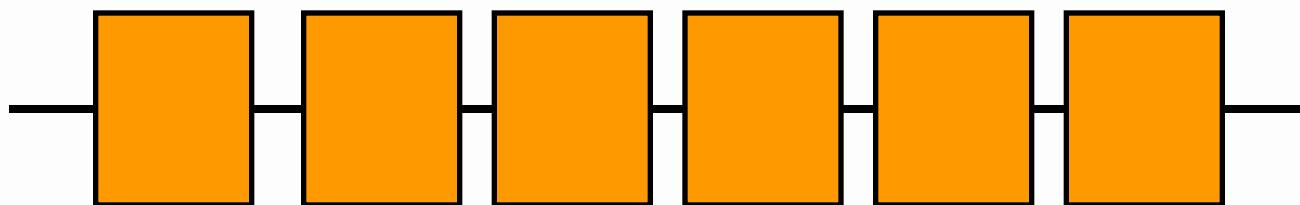
$$C = \frac{19}{20}$$

# Dependability modelling

- up to now:  $\lambda$  and  $R(t)$  for components
- systems are sets of components
- system evaluation approaches:
  - reliability block diagrams
  - Markov processes

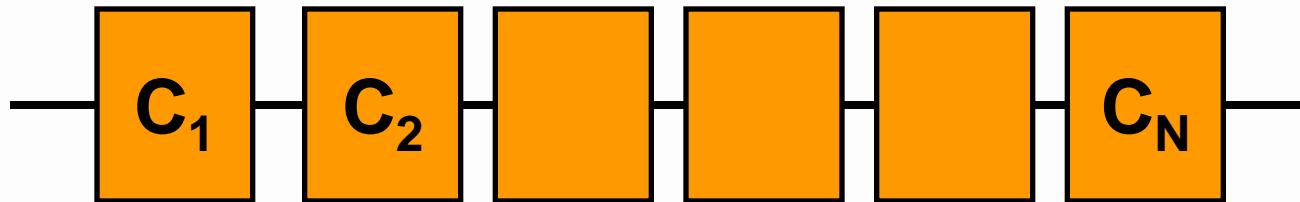
# Serial system

- system functions  
if and only if all components function



**reliability block diagram  
(RBD)**

# Serial system



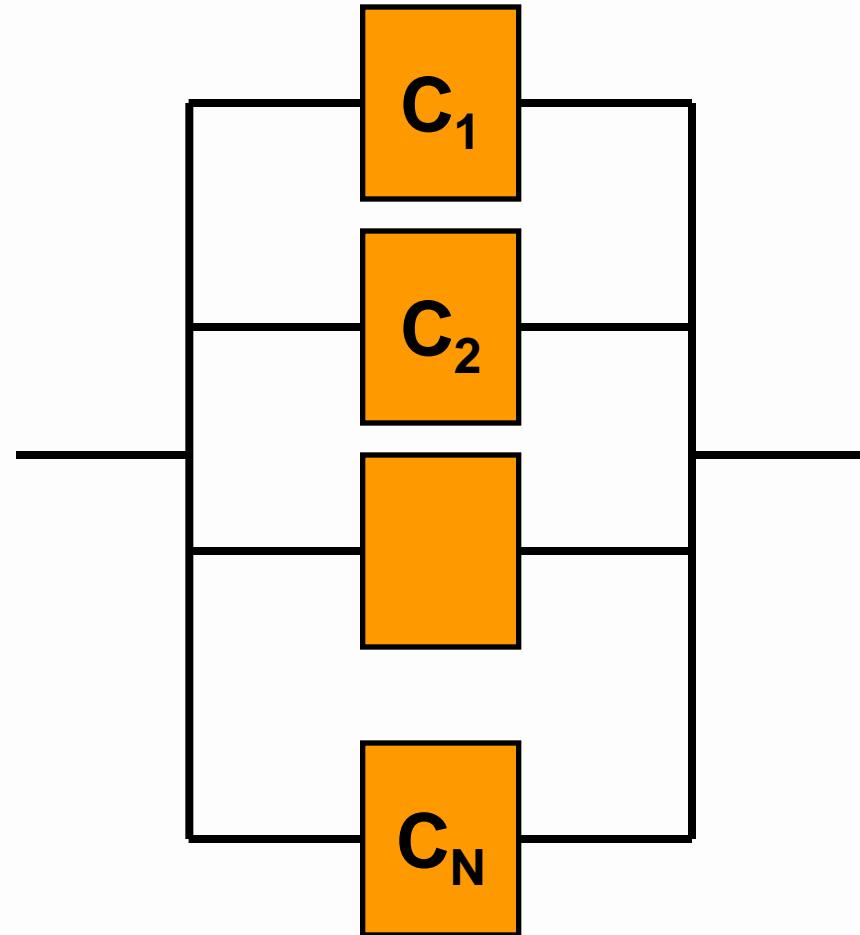
**if  $C_i$  are independent:**

$$R_{\text{series}}(t) = \prod R_i(t)$$

$$\lambda_{\text{series}} = \sum_{i=1}^N \lambda_i$$

# Parallel system

- system works as long as one component works



# Parallel system

**unreliability:**  $Q(t) = 1 - R(t)$

**if  $C_i$  are independent:**  $Q_{parallel}(t) = \prod_{i=1}^N Q_i(t)$

$$R_{parallel}(t) = 1 - \prod_{i=1}^N (1 - R_i(t))$$

# Reliability block diagram

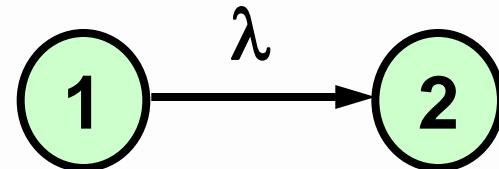
- RBD
  - may be difficult to build
  - equations get complex
  - difficult to take coverage into account
  - difficult to represent repair
  - not possible to represent dependency between components

# Markov chains

- Markov chains
  - illustrated by state transition diagrams
- idea:
  - states
    - components working or not
  - state transitions
    - when components fail or get repaired

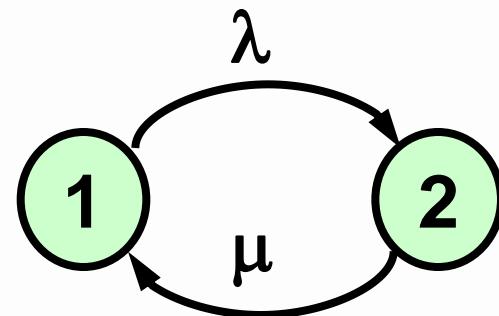
# Single-component system, no repair

- Only two states
  - one operational (state 1) and one failed (state 2)
  - if no repair is allowed, there is a single, non-reversible transition between the states (used in availability analysis)
  - label  $\lambda$  corresponds to the failure rate of the component



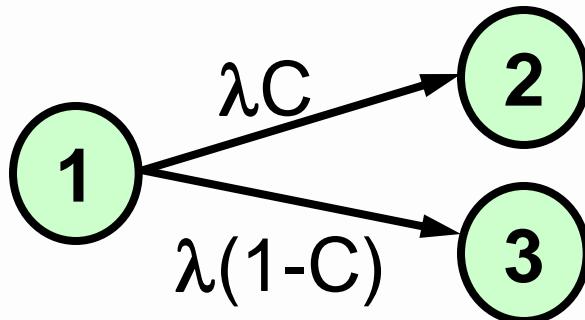
# Single-component system with repair

- If repair is allowed (used in availability analysis)
  - then a transition between the failed and the operational state is possible
  - the label is the repair rate  $\mu$



# Failed-safe and failed-unsafe

- In safety analysis, we need to distinguish between failed-safe and failed-unsafe states
  - let 2 be a failed-safe state and 3 be a failed-unsafe state
  - the transition between the 1 and 2 depends on failure rate and the probability that, if a fault occurs, it is detected and handled appropriately (i.e. fault coverage  $C$ )
  - if  $C$  is the probability that a fault **is** detected,  $1-C$  is the probability that a fault **is not** detected



# Two-component system

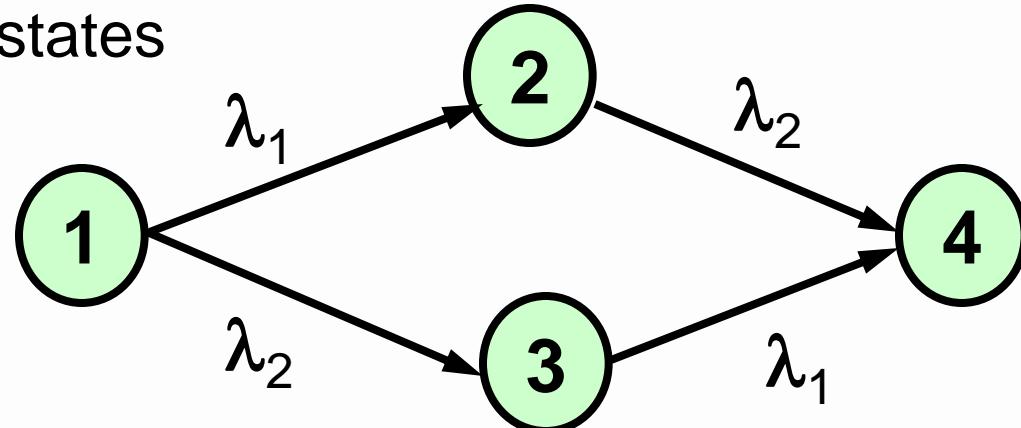
- Has four possible states

O O state 1

F O state 2

O F state 3

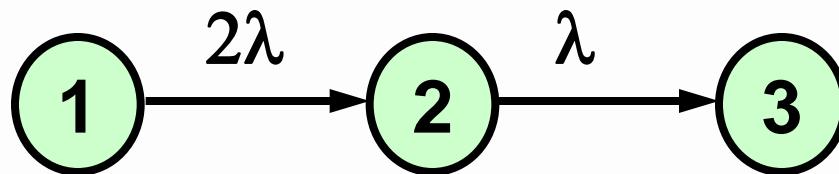
F F state 4



- Components are assumed to be independent and non-repairable
- If components are in serial
  - state 1 is operational state, states 2,3,4 are failed states
- If components are in parallel
  - states 1,2,3 are operational states, state 4 is failed state

# State transition diagram simplification

- Suppose two components are in parallel
- Suppose  $\lambda_1 = \lambda_2 = \lambda$
- Then, it is not necessary to distinguish between the states 2 and 3
  - both represent a condition where one component is operational and one is failed
  - since components are independent events, transition rate from state 1 to 2 is the sum of the two transition rates



# Markov chain analysis

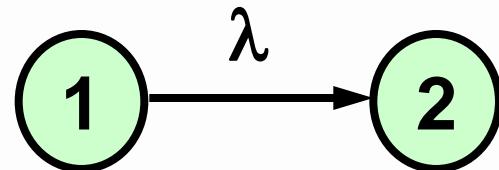
- The aim is to compute  $P_i(t)$ , the probability that the system is in the state  $i$  at time  $t$
- Once  $P_i(t)$  is known, the reliability, availability or safety of the system can be computed as a sum taken over all operating states
- To compute  $P_i(t)$ , we derive a set of differential equations, called **state transition equations**, one for each state of the system

# Transition matrix

- State transition equations are usually presented in matrix form
- Transition matrix  $M$  has entries  $m_{ij}$ , representing the rates of transition between the states  $i$  and  $j$ 
  - index  $i$  is used for the number of columns
  - index  $j$  is used for the number of rows

$$M = \begin{bmatrix} m_{11} & m_{21} \\ m_{12} & m_{22} \end{bmatrix}$$

# Single-component system, no repair

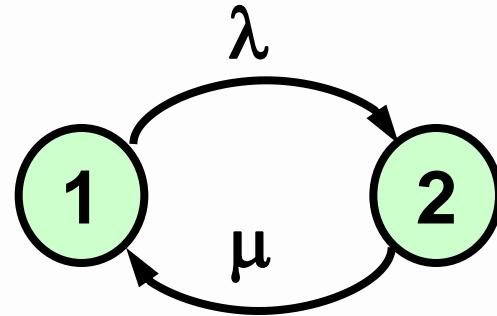


- Transition matrix  $M$  has the form:

$$M = \begin{bmatrix} -\lambda & 0 \\ \lambda & 0 \end{bmatrix}$$

- entries in each columns must sum up to 0
  - entries  $m_{ii}$ , corresponding to self-transitions, are computed as  $-(\text{sum of other entries in this column})$

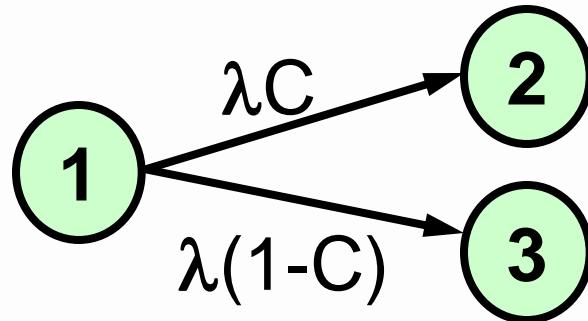
# Single-component system with repair



- Transition matrix  $M$  has the form:

$$M = \begin{bmatrix} -\lambda & \mu \\ \lambda & -\mu \end{bmatrix}$$

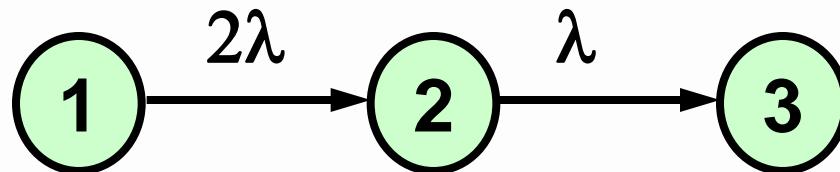
# Single-component system, safety analysis



- Transition matrix  $M$  has the form:

$$M = \begin{bmatrix} -\lambda & 0 & 0 \\ \lambda C & 0 & 0 \\ \lambda(1-C) & 0 & 0 \end{bmatrix}$$

# Two-component parallel system



- Transition matrix  $M$  has the form:

$$M = \begin{bmatrix} -2\lambda & 0 & 0 \\ 2\lambda & -\lambda & 0 \\ 0 & \lambda & 0 \end{bmatrix}$$

# Important properties of matrix M

- Sum of the entries in each column is 0
- Positive sign of an  $ij_{th}$  entry indicates that the transition originates from the  $i_{th}$  state
- In reliability analysis, M allows us to distinguish between the operational and failed states
  - each failed state  $i$  has a zero diagonal element  $m_{ii}$  (a failed state cannot leave)

# State transition equations

- Let  $P(t)$  be a vector whose  $i_{th}$  element is the probability  $P_i(t)$ , the probability that the system is in the state  $i$  at time  $t$
- The matrix representation of a system of state transition equations is given by

$$\frac{d}{dt} P(t) = M \cdot P(t)$$

# Two-component parallel system

- Using transition matrix derived earlier, we get:

$$\frac{d}{dt} \begin{bmatrix} P_1(t) \\ P_2(t) \\ P_3(t) \end{bmatrix} = \begin{bmatrix} -2\lambda & 0 & 0 \\ 2\lambda & -\lambda & 0 \\ 0 & \lambda & 0 \end{bmatrix} \cdot \begin{bmatrix} P_1(t) \\ P_2(t) \\ P_3(t) \end{bmatrix}$$

- This represents the following system of equations

$$\begin{cases} \frac{d}{dt} P_1(t) = -2\lambda P_1(t) \\ \frac{d}{dt} P_2(t) = 2\lambda P_1(t) - \lambda P_2(t) \\ \frac{d}{dt} P_3(t) = \lambda P_2(t) \end{cases}$$

# Solving state transition equations

- By solving these equations, we get

$$P_1(t) = e^{-2\lambda t}$$

$$P_2(t) = 2e^{-\lambda t} - 2e^{-2\lambda t}$$

$$P_3(t) = 1 - 2e^{-\lambda t} + e^{-2\lambda t}$$

- Since the  $P_i(t)$  are known, we can compute the reliability of the system as a sum of probabilities taken over all operating states

$$R_{\text{parallel}}(t) = P_1(t) + P_2(t) = 2e^{-\lambda t} - e^{-2\lambda t}$$

# Comparison to RBD result

- Since  $R = e^{-\lambda t}$ , the previous equation can be written as

$$R_{\text{parallel}}(t) = 2R - R^2$$

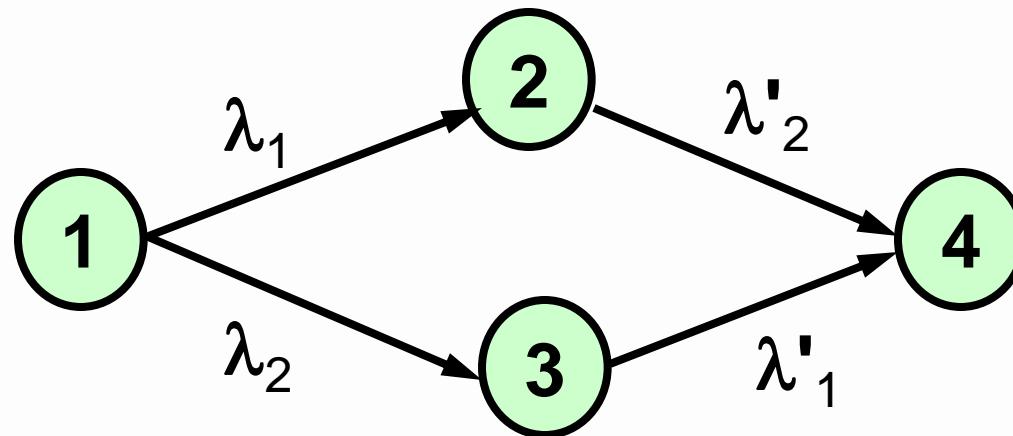
- which agrees with the expression derived using RBD
- two results are the same because we assumed that the failure rates of the two components are independent

# Dependant component case

- The value of Markov chains become evident when component failures cannot be assumed to be independent
  - load-sharing components
  - examples: electrical load, mechanical load, information load
- If two components share the same load and one fails, the additional load on the second component increases its failure rate

# Parallel system with load sharing

- As before, we have four states, but after the 1<sup>st</sup> component failure, the failure rate of the 2<sup>nd</sup> component increases



# Parallel system with load sharing

- State transition equations are:

$$\frac{d}{dt} \begin{bmatrix} P_1(t) \\ P_2(t) \\ P_3(t) \\ P_4(t) \end{bmatrix} = \begin{bmatrix} -\lambda_1 - \lambda_2 & 0 & 0 & 0 \\ \lambda_1 & -\lambda'_2 & 0 & 0 \\ \lambda_2 & 0 & -\lambda'_1 & 0 \\ 0 & \lambda'_2 & \lambda'_1 & 0 \end{bmatrix} \cdot \begin{bmatrix} P_1(t) \\ P_2(t) \\ P_3(t) \\ P_4(t) \end{bmatrix}$$

$$\left\{ \begin{array}{l} \frac{d}{dt} P_1(t) = (-\lambda_1 - \lambda_2)P_1(t) \\ \frac{d}{dt} P_2(t) = \lambda_1 P_1(t) - \lambda'_2 P_2(t) \\ \frac{d}{dt} P_3(t) = \lambda_2 P_1(t) - \lambda'_1 P_3(t) \\ \frac{d}{dt} P_4(t) = \lambda'_2 P_2(t) + \lambda'_1 P_3(t) \end{array} \right.$$

# Effect of the load

- If  $\lambda'_1 = \lambda_1$  and  $\lambda'_2 = \lambda_2$ , the equation of load sharing parallel system reduces to well-known

$$R_{\text{parallel}}(t) = 2e^{-\lambda t} - e^{-2\lambda t}$$

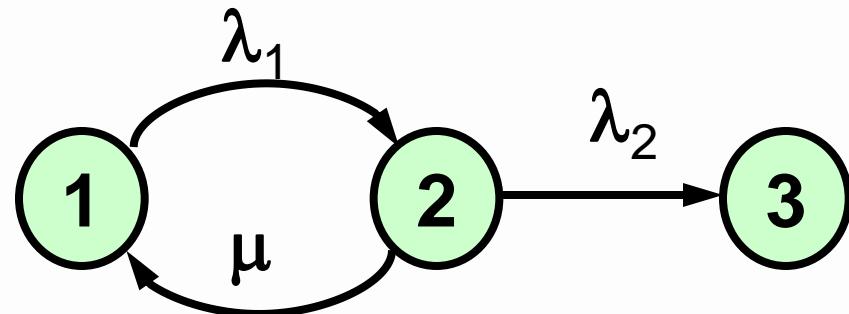
# Availability evaluation

- Difference with reliability analysis:
  - in reliability analysis components are allowed to be repaired as long as the system has not failed
  - in availability analysis components can also be repaired after the system failure

# Two-component standby system

- First component is primary
- Second is held in reserve and only brought to operation if the first component fails
- We assume that
  - fault detection unit which detect failure of the primary component are replace is with standby is perfect
  - standby component cannot fail while in the standby mode

# State transition diagram for reliability analysis with repair



state 1: both OK

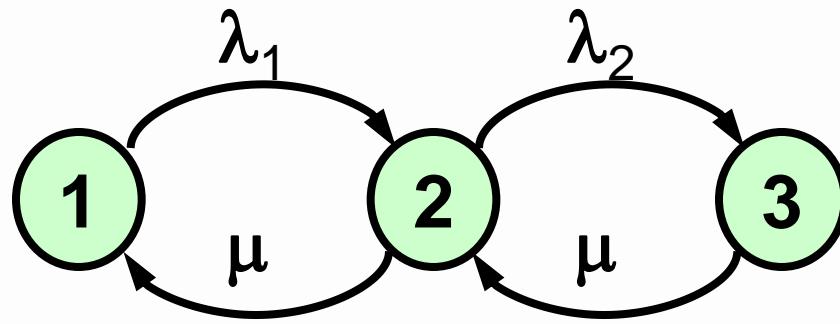
state 2: primary failed and replaced by spare

state 3: both failed

$$M = \begin{bmatrix} -\lambda_1 & \mu & 0 \\ \lambda_1 & -\lambda_2 - \mu & 0 \\ 0 & \lambda_2 & 0 \end{bmatrix}$$

Repair replaces a broken component by a working one.

# State transition diagram for availability analysis with repair

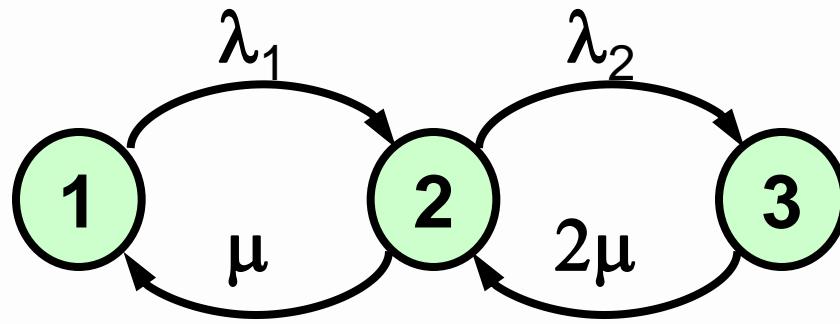


$$M = \begin{bmatrix} -\lambda_1 & \mu & 0 \\ \lambda_1 & -\lambda_2 - \mu & \mu \\ 0 & \lambda_2 & -\mu \end{bmatrix}$$

States are the same.

Repair replaces a broken component by a working one. Here we assume that there is only one repair team.

# State transition diagram for availability analysis with repair



If we assume that there are two independent repair teams, then  $\mu$  on the edge from 3 to 2 gets the coefficient 2 (the rate doubles).

$$M = \begin{bmatrix} -\lambda_1 & \mu & 0 \\ \lambda_1 & -\lambda_2 - \mu & 2\mu \\ 0 & \lambda_2 & -2\mu \end{bmatrix}$$

# Availability analysis

- None of the diagonal elements of  $M$  are 0
- By solving the system, we can get  $P_i(t)$  and compute the availability as a sum of probabilities taken over all operating states
- Usually steady-state availability rather than time dependent one is of interest
- As time approaches infinity, the derivative of the right-hand side of the equation  $d/dt P(t) = M \cdot P(t)$  vanishes and we get time-independent relationship

$$M \cdot P(\infty) = 0$$

# Two-component standby system

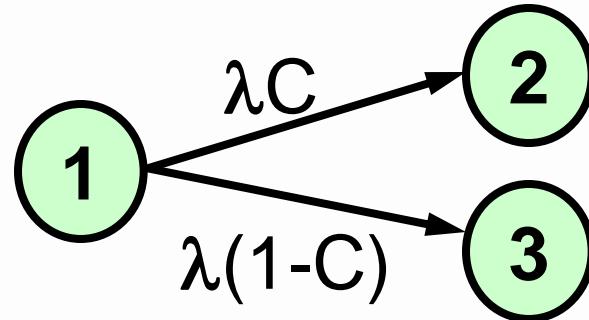
- Using transition matrix derived earlier, we get the following system of equations

$$\begin{cases} -\lambda_1 P_1(\infty) + \mu P_2(\infty) = 0 \\ \lambda_1 P_1(\infty) - (\lambda_2 + \mu) P_2(\infty) + \mu P_3(\infty) = 0 \\ \lambda_2 P_2(\infty) - \mu P_3(\infty) = 0 \end{cases}$$

- By solving the equations, we get

$$A(\infty) \approx 1 - (\lambda/\mu)^2$$

# Safety evaluation



- The state transition equations are:

$$\frac{d}{dt} \begin{bmatrix} P_1(t) \\ P_2(t) \\ P_3(t) \end{bmatrix} = \begin{bmatrix} -\lambda & 0 & 0 \\ \lambda C & 0 & 0 \\ \lambda(1-C) & 0 & 0 \end{bmatrix} \cdot \begin{bmatrix} P_1(t) \\ P_2(t) \\ P_3(t) \end{bmatrix}$$

# Safety evaluation

- By solving these equations, we get

$$P_1(t) = e^{-\lambda t}$$

$$P_2(t) = C(1 - e^{-\lambda t})$$

$$P_3(t) = (1 - C) - (1 - C)e^{-\lambda t}$$

- Since the  $P_i(t)$  are known, we can compute the reliability of the system as a sum of probabilities of being the operational and fail-safe states

$$R(t) = P_1(t) + P_2(t) = C + (1 - C)e^{-\lambda t}$$

- At time  $t=0$ , the safety is 1. As time approaches infinity, the safety approaches C

# How to deal with cases of systems with “k out of n choices”

- Suppose we want to solve the following task:  
*What is the probability that more than two engines in a 4-engine airplane will fail during a t-hour flight if the failure rate of a single engine is  $\lambda$  per hour?*
- The probability that more than two engines fail can be expressed as:

$$\begin{aligned} P_{>2 \text{ failed}} &= \binom{4}{1} P_{1 \text{ works}} P_{3 \text{ failed}} + P_{4 \text{ failed}} \\ &= 1 - (P_{4 \text{ work}} + \binom{4}{3} P_{3 \text{ work}} P_{1 \text{ failed}} + \binom{4}{2} P_{2 \text{ work}} P_{2 \text{ failed}}) \end{aligned}$$

- Only probabilities of mutually exclusive events can be summed up like this

# “k out of n choices”

- “k out of n choices” can be computed as

$$\binom{n}{k} = \frac{n!}{(n-k)! k!}$$

- For example

$$\binom{4}{2} = \frac{4!}{(4-2)! 2!} = 6$$

## Example cont.

So, we get

$$P_{>2 \text{ failed}} = 4 P_{1 \text{ works 3 failed}} + P_{4 \text{ failed}}$$

where

$$P_{1 \text{ works 3 failed}} = R (1-R)^3$$

$$P_{4 \text{ failed}} = (1-R)^4$$

where  $R$  is the reliability of a single engine  
computed as  $R = e^{-\lambda t}$

# Summary

- Methods for evaluating the reliability, availability and safety of a system
  - RBDs
  - Markov chains

# Next lecture

- Hardware redundancy

**Read chapter 4  
of the text book**