

Lecture 07

Software Reliability

Software Reliability

- In the broad sense
 - S/w design will operate (execute) well for a substantial time period.
- In a narrow sense
 - Reliability is a metric which is probability of operational success of the s/w
- Probabilistic and Deterministic Model
 - **Example** – Winding of a motor bourns out
 - Due to temperature above melting point

(It will burns – it is certainly deterministic, But we are unable to predict when they will occur – it is probabilistic)

Software Reliability

- Thus the model of failures become a probabilistic one and the random element (random variable) is the time to failure
- Failure Modes
 - Hardware
 - Software
 - Skinware(Human)
- Reliability Theory
 - Modeling of failure and prediction of success probability
 - Based on the concept of random/continuous variable, probability density function, probability distribution function

Software Reliability

- Probability concept :
- Discrete random variable, X is a discrete random variable
 - 1) Probability Density Function :
Probability of occurrence $P(x_i) = f(x_i)$
 - 2) Distribution Function : Defined in term of the probability that $X \leq x$:

$$P(X \leq x) \equiv F(x) = \sum_{X \leq x} f(x)$$

- Continuous random variable,
The variable is continuous over some range of definition
Density function, $P(t < x < t + \Delta t) = f(t)\Delta t$

Software Reliability

- **Distribution Function:** $F(x) = \int_a^x f(x) dx$

– Let x take on all values between points a and b ,

$$\text{So, } P(X \leq x) = F(x) = \int_a^x f(x) dx \quad \text{for } a < x \leq b$$

$$dF(x) / dx = f(x)$$

– The Probability that X lies in an interval $x < X < x+dx$,

$$P(x < X < x + dx) = P(X \leq x + dx) - P(X \leq x)$$

$$= \int_a^{x+dx} f(x) dx - \int_a^x f(x) dx$$

$$= \int_x^{x+dx} f(x) dx = F(x + dx) - F(x)$$

$F(x)$ is a continuous, $dx \rightarrow 0$; $P(X=x)$ is zero

Definition of Reliability

- $R(t)$ = Reliability is the mathematical probability is function of time t , is the success.

$$R(t) = P(x > t) = 1 - P(x \leq t) \text{ where } x \text{ is continuous}$$

$$= P(x >= t) = 1 - P(x <= t) \text{ where } x \text{ is discrete}$$

- Hazard rate (conditional failure rate), $z(t)$
 - Defined in term of the probability that a failure occurs in the same interval t to $t + \Delta t$, given that system has survival up to time t

$$Z(t) \Delta t = P(t < x \leq t + \Delta t \mid x > t)$$

We know, $F(t) = 1 - R(t)$

Relationship R(t) and Z(t)

- We know, from definition of Hazard rate

$$Z(t) \Delta t = P(t < x < t + \Delta t \mid x > t) \quad [P(AB) = P(A) \cdot P(B/A)]$$

$$Z(t) = P(t < x < t + \Delta t) / \Delta t \cdot P(x > t) \quad [P(AB) = P(B) \cdot P(A/B)]$$

$$= \{ P(x < t + \Delta t) - P(x < t) \} / \Delta t \cdot P(x > t)$$

$$= \{ 1 - P(x > t + \Delta t) \} - \{ 1 - P(x > t) \} / \Delta t \cdot P(x > t)$$

$$= -1 / R(t) \times [R(t + \Delta t) - R(t)] / \Delta t$$

$$= -R'(t) / R(t)$$

$$\Rightarrow \int_0^t Z(t) dt = - \int_0^t R'(t) / R(t) dt$$

$$R(t) = e^{- \int_0^t Z(t) dt}$$

Mean Time To Failure (MTTF)

- Estimation theory:
 - How one determines the parameters in a probabilistic model from statistical data taken on the items governed by the model
 - Specifically, in reliability work we place a group of components on **life test** and observe the sequence of failure times t_1, t_2, \dots, t_n
 - On the basis of these data we compute time-to-failure models and hazard models
 - Estimation theory provides guidelines for efficient and accurate components

Mean Time To Failure (MTTF)

- $n \leq 5$ (few data) \rightarrow result must be questioned
- $n \geq 100$ \rightarrow result should be good
- $n \rightarrow \infty$ \rightarrow many of the different computational scheme is need
- $10 \leq n \leq 50$ \rightarrow best for estimation theory
- A point estimation formula, $MTTF = (10 + 20 + 25 + 35 + 40) / 5$
- $= 26h$
- It is often convenient to characterize a failure model using set of failure data by a single parameter (MTTF/MTBF)

Mean Time To Failure (MTTF)

- If we have life test information on a population of n items with failure times t_1, t_2, \dots, t_n

- $MTTF = \frac{1}{n} \sum_{i=1}^n t_i$

- Expected value, $E(x) = \sum_{i=1}^n x_i f(x_i); x = x_1, x_2, \dots, x_n$ where x

is discrete random variable

$$E(x) = \int_a^b x f(x) dx; a \leq x \leq b$$

where x

is continuous random variable

Mean Time To Failure (MTTF)

- Using Hazard model, the MTTF for the probability distribution defined by the model is,

- $$\text{MTTF} = E(t) = \int_0^{\infty} t f(t) dt$$

- Let, $I = \int t f(t) dt$

$$= t \int f(t) dt - \int [d/dt(t) \int f(t) dt] dt$$

$$= -tR(t) + \int R(t) dt$$

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

$$f(t) = dF(t)/dt$$

$$= d(1-R(t))/dt$$

$$= -dR(t)/dt$$

Mean Time To Failure (MTTF)

- Case I : Constant hazard, $R(t) = e^{-\lambda t}$
- Case II : Linearly increasing hazard, $R(t) = e^{-kt^2/2}$
- Case III : Weibull distribution, $R(t) = e^{-kt^{(m+1)}/(m+1)}$