

Normal (Gaussian) dist<sup>n</sup>  
 $f(x) = \frac{1}{\sqrt{2\sigma}} e^{-\frac{x^2}{2\sigma}}$   
 (10) : 68%  
 (20) : 95%  
 (30) : 99.7%

### QQ Plots

to check if two datasets are from similar dist<sup>n</sup>  
 (diagonal plot)  $\Rightarrow$  similar dist<sup>n</sup>  
 to check Normality (normal dist<sup>n</sup>)

Quantile of data

$z = \frac{x_i - \mu}{\sigma}$

Z score of data

Black Swan Theory :- Rare events are more likely to occur Vs as predicted by normal distribution

Student's t distribution  
 tails size controlled by add<sup>n</sup>  
 Parameter  $\nu$  : degrees of freedom  
 $\nu = n - 1$  (normal dist)  
 (short tail)

95% Confidence interval around Sample mean =  $\bar{x} \pm t_{n-1} (0.05) \frac{s}{\sqrt{n}}$

### Common Types of distribution

Binomial distribution  
 $f(k) = \binom{n}{k} p^k (1-p)^{n-k}$   
 $n, p$   
 $q = 1-p$   
 $\rightarrow$  probab<sup>y</sup> of success in  $n$  trials  
 $\rightarrow$  Success =  $p$   
 $\rightarrow$  mean =  $np$   
 $\rightarrow$  std =  $\sqrt{npq}$   
 $\rightarrow$  (for  $n/2, n \rightarrow \infty$ )  
 Binomial  $\rightarrow$  Normal

Poisson distribution  
 $f(k) = \frac{\lambda^k e^{-\lambda}}{k!}$   
 $\rightarrow$  probab<sup>y</sup> of  $k$  events occurrence in unit time  
 $\rightarrow$  mean  $\rightarrow \lambda$   
 $\rightarrow$  std  $\rightarrow \lambda$

$\rightarrow$  gives an event occurring  $\lambda$  times per interval of time/space on avg  
 $\rightarrow$  probab<sup>y</sup> of  $k$  events occurrence in unit time  
 $\rightarrow$   $\lambda$  might not be const at smaller time frame but over sufficiently larger time frame  
 $\rightarrow$  Can be used to measure request Capacity measure

Exponential dist<sup>n</sup>  
 $f(x) = e^{-\lambda x}$   
 $\rightarrow$  time b/w two consecutive events

Chi-square statistics  
 Statistics to measure how far different is the observed distribution from given dist<sup>n</sup>  
 to check the null hypothesis of independence  
 $\chi^2$  Statistic :  $\sum \frac{(O-E)^2}{E}$   
 $O$  : Observed frequency  
 $E$  : expected frequency  
 for Categorical variables  
 the  $\chi^2$  statistic follows the  $\chi^2$  distribution

Goodness of fit

Its actually distribution of  $\chi^2$  values  
 $\chi^2 = \sum \frac{(x_i - \mu)^2}{\sigma^2}$   
 where  $x_i$  : standard Normal dist<sup>n</sup>

F-distribution  
 $\rightarrow$  Similar to Chi-square statistics but instead of dealing with counts dealing with continuous variables  
 $\rightarrow$  F-statistics  
 $\sim$  Variability among group mean  
 Variability within each group  
 $\rightarrow$  Residual Variability

Weibull Distribution  
 $\left\{ \begin{array}{l} \text{time to failure} \\ \text{when } \lambda \text{ is increasing/decreasing} \end{array} \right.$

$f(t) = \frac{\beta}{\eta} \left(\frac{t}{\eta}\right)^{\beta-1} \exp\left(-\left(\frac{t}{\eta}\right)^\beta\right)$   
 $\rightarrow$  Probab<sup>y</sup> of failure at time  $t$   
 $\eta$  :  $\frac{1}{\lambda}$  : characteristic life

$\beta \rightarrow$  Shape parameter  
 $> 1$  : Probab<sup>y</sup> of failure increase over time  
 $< 1$  : Probab<sup>y</sup> of failure decreases over time

