

Exam I

Question I

Data Center Reliability

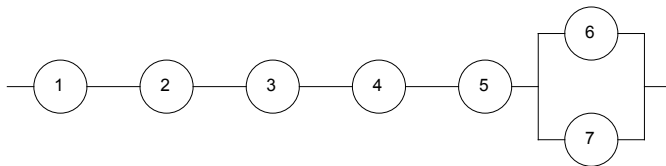
```
Show[Graphics[{
  Circle[{-18, 0}, 1],
  Circle[{-14, 0}, 1],
  Circle[{-10, 0}, 1],
  Circle[{-6, 0}, 1],
  Circle[{-2, 0}, 1],
  Circle[{2, 2}, 1],
  Circle[{2, -2}, 1],

  Text["1", {-18, 0}],
  Text["2", {-14, 0}],
  Text["3", {-10, 0}],
  Text["4", {-6, 0}],
  Text["5", {-2, 0}],
  Text["6", {2, 2}],
  Text["7", {2, -2}],

  Line[{{-20, 0}, {-19, 0}},
  Line[{{-17, 0}, {-15, 0}},
  Line[{{-13, 0}, {-11, 0}},
  Line[{{-9, 0}, {-7, 0}},
  Line[{{-5, 0}, {-3, 0}},
  Line[{{-1, 0}, {0, 0}},
  Line[{{4, 0}, {5, 0}},

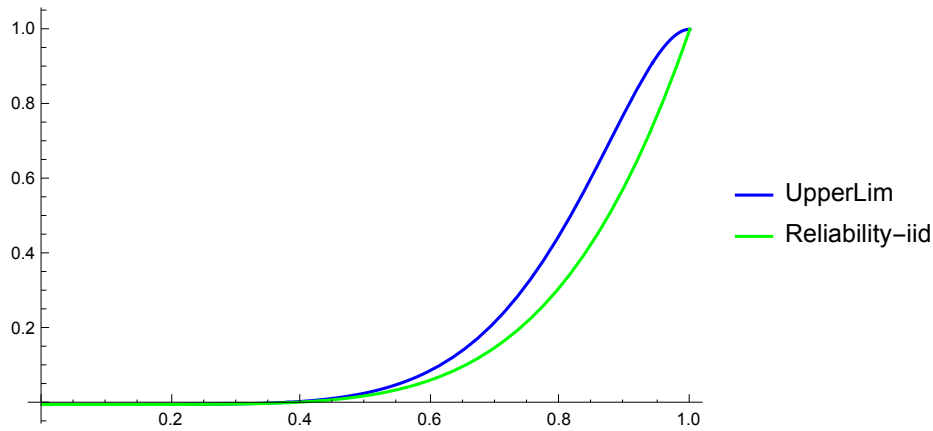
  Line[{{0, -2}, {0, 2}},
  Line[{{4, -2}, {4, 2}},

  Line[{{0, 2}, {1, 2}},
  Line[{{3, 2}, {4, 2}},
  Line[{{0, -2}, {1, -2}},
  Line[{{3, -2}, {4, -2}}
}]]
```



```
ClearAll["Global`*"]
dist = BernoulliDistribution[p];
Rdatacenter =
  ReliabilityDistribution[X1 & X2 & X3 & X4 & X5 & (X6 ∨ X7), {{X1, dist},
    {X2, dist}, {X3, dist}, {X4, dist}, {X5, dist}, {X6, dist}, {X7, dist}}];
r[p_] = FullSimplify[Mean[Rdatacenter]]
- (-2 + p) p6
```

```
Plot[{p^6 (2 - p^6), r[p]}, {p, 0, 1},
  PlotLegends -> Placed[{"UpperLim", "Reliability-iid"}, Right],
  PlotStyle -> {Blue, Green}]
```



```
ClearAll["Global`*"]
dist = ExponentialDistribution[λ];
dist4 = ErlangDistribution[2, λ];
Rdatacenter =
  ReliabilityDistribution[X1 & X2 & X3 & X4 & X5 & (X6 ∨ X7), {{X1, dist},
    {X2, dist}, {X3, dist}, {X4, dist4}, {X5, dist}, {X6, dist}, {X7, dist}}];
FullSimplify[SurvivalFunction[Rdatacenter, t]]
FullSimplify[HazardFunction[Rdatacenter, t]]
h[t_] = FullSimplify[HazardFunction[Rdatacenter, t]];
FullSimplify[∂t h[t]]
Mean[Rdatacenter]

$$\begin{cases} 1 & t \leq 0 \\ e^{-6t\lambda} (-1 + 2 e^{t\lambda}) \text{GammaRegularized}[2, t\lambda] & \text{True} \end{cases}$$


$$\begin{cases} \lambda \left( 6 + \frac{1}{1-2 e^{t\lambda}} - \frac{1}{1+t\lambda} \right) & t \geq 0 \\ 0 & \text{True} \end{cases}$$


$$\begin{cases} \text{Indeterminate} & t == 0 \\ \lambda^2 \left( \frac{1}{(1-2 e^{t\lambda})^2} + \frac{1}{-1+2 e^{t\lambda}} + \frac{1}{(1+t\lambda)^2} \right) & t > 0 \\ 0 & \text{True} \end{cases}$$



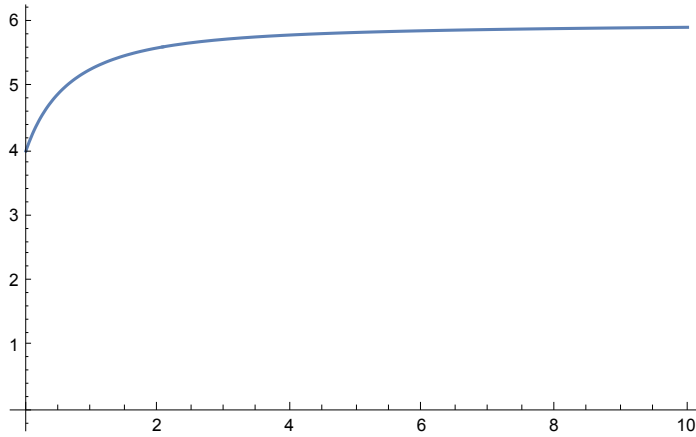
---


199
882 λ
```

```

ClearAll["Global`*"]
λ = 1.;
R[t_] = Exp[-6 λ t] * (1 + λ t) * (2 - Exp[-λ t]);
f[t_] = -D[R[t], t];
h[t_] = f[t] / R[t];
Plot[h[t], {t, 0.001, 10.}, AxesOrigin -> {0, 0}]

```



```

ClearAll["Global`*"]
λ = 1.;
dist = ExponentialDistribution[λ];
dist4 = ErlangDistribution[2, λ];
Rdatacenter =
  ReliabilityDistribution[X1 & X2 & X3 & X4 & X5 & (X6 ∨ X7), {{X1, dist},
    {X2, dist}, {X3, dist}, {X4, dist4}, {X5, dist}, {X6, dist}, {X7, dist}}];

p = 0.5;
ξ = Quantile[Rdatacenter, p]
If[0.5 ≤ 1 - Exp[-1.], Print[{- (p * ξ) / Log[1 - p], - ξ / Log[1 - p]}], Print[{- (p * ξ) / Log[1 - p], ξ}]]

Mean[Rdatacenter]
  2
  11 λ
0.164423

{0.118606, 0.237213}

0.225624

0.181818

```

Question 2 — Type II/item censoring without replacement

```
ClearAll["Global`*"]
data = {10.2, 89.6, 54.0, 96.0, 23.3, 30.4, 41.2, 0.8, 73.2, 3.6, 28.0, 31.6};
data = Sort[data]
```

```
n = 20.;
r = 12;
```

```
tau =  $\sum_{i=1}^r$  data[[i]] + (n - r) * data[[r]]
- (tau / r) * Log[1 - 0.75]
```

```
2 * tau
Quantile[ChiSquareDistribution[2 * r], 0.025]
Quantile[ChiSquareDistribution[2 * r], 0.975]
```

```
 $\lambda_L$  = Quantile[ChiSquareDistribution[2 * r], 0.025] / (2 * tau)
 $\lambda_U$  = Quantile[ChiSquareDistribution[2 * r], 0.975] / (2 * tau)
- 1 /  $\lambda_U$  * Log[1 - 0.75]
- 1 /  $\lambda_L$  * Log[1 - 0.75]
```

```
 $\lambda_L$  = 12.40 / 2499.8
 $\lambda_U$  = 39.36 / 2499.8
- 1 /  $\lambda_U$  * Log[1 - 0.75]
- 1 /  $\lambda_L$  * Log[1 - 0.75]
```

```
{0.8, 3.6, 10.2, 23.3, 28., 30.4, 31.6, 41.2, 54., 73.2, 89.6, 96.}
```

```
1249.9
```

```
144.394
```

```
2499.8
```

```
12.4012
```

```
39.3641
```

```
0.00496086
```

```
0.0157469
```

```
88.0361
```

```
279.447
```

```
0.0049604
```

```
0.0157453
```

```
88.0452
```

```
279.472
```

Question 3

```
In[13]:= ClearAll["Global`*"]  
data = {2, 3, 8, 1, 1, 4, 1, 4, 5, 1};  
 $\lambda_0$  = Mean[data]  
LCL = Ceiling[Max[0,  $\lambda_0 - 3 \times \sqrt{\lambda_0}$ ]]  
UCL = Floor[ $\lambda_0 + 3 \times \sqrt{\lambda_0}$ ]
```

Out[15]= 3

Out[16]= 0

Out[17]= 8

In[30]=

```

L = 0;
U = 10;
gammaL = 0.038733;
gammaU = 0.591561;
ξ[θ_] = 1 -
  (CDF[PoissonDistribution[λ0 + θ], U] - CDF[PoissonDistribution[λ0 + θ], L - 1]) +
  gammaL * PDF[PoissonDistribution[λ0 + θ], L] +
  gammaU * PDF[PoissonDistribution[λ0 + θ], U];

(*ξstar[0]
  1 - (0.9997 - 0) + 0.038733 * (0.0498 - 0) + 0.591561 * (0.9997 - 0.9989) *)

ξ[-1]
1 - (1.0000 - 0) + 0.038733 * (0.1353 - 0) + 0.591561 * (1.0000 - 0.9998)

ARL[θ_] = 1 / ξ[θ];
ARL[0]
ARL[-1]
1 / 0.00535889

Plot[{ARL[θ], ARL[θ]}, {θ, -3, 3}]

```

Out[35]= 0.00527284

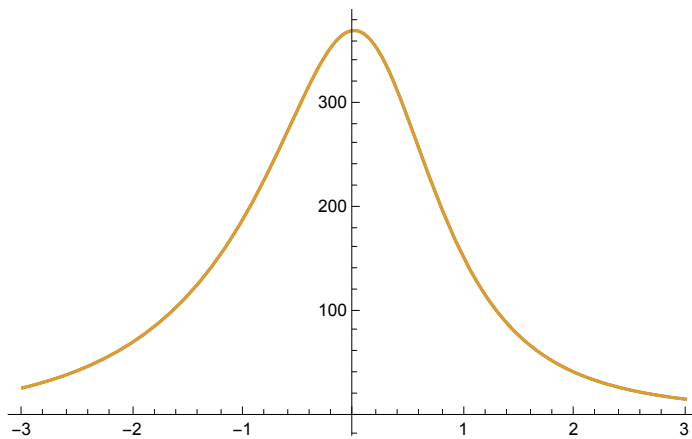
Out[36]= 0.00535889

Out[38]= 370.371

Out[39]= 189.651

Out[40]= 186.606

Out[41]=



Question 4 — (\bar{X}, S^2) joint scheme for μ and σ^2

```

n = 4;
μ₀ = 0;
σ₀ = 1;
incontrolARLindiv = 500.;

distmu = NormalDistribution[0, 1];
γmu = Quantile[distmu, 1 - 1 / (2 * incontrolARLindiv)]
ξmu[δ_, θ_] = 1 - (CDF[distmu,  $\frac{\gamma\mu - \delta}{\theta}$ ] - CDF[distmu,  $\frac{-\gamma\mu - \delta}{\theta}$ ]);
ARLmu[δ_, θ_] = 1 / ξmu[δ, θ];
ARLmu[0, 1]

distsigma = ChiSquareDistribution[n - 1];
γsigma = Quantile[distsigma, 1 - 1 / incontrolARLindiv]
ξsigma[θ_] = 1 - CDF[distsigma,  $\frac{\gamma\sigma}{\theta^2}$ ];
ARLsigma[θ_] = 1 / ξsigma[θ];
ARLsigma[1]

ξmusigma[δ_, θ_] = ξmu[δ, θ] + ξsigma[θ] - ξmu[δ, θ] * ξsigma[θ];
ARLmusigma[δ_, θ_] = 1 / ξmusigma[δ, θ];
ARLmusigma[0, 1]

ARLmusigma[δ_, θ_] =  $\frac{ARLmu[\delta, \theta] * ARLsigma[\theta]}{ARLmu[\delta, \theta] + ARLsigma[\theta] - 1}$ ;
shiftmu = 0.1;
ARLmusigma[shiftmu, 1]

ARLmu[shiftmu, 1]
ARLsigma[1]


$$\frac{(1 - \xi\mu[\text{shiftmu}, 1]) \times \xi\sigma[1]}{\xi\mu\sigma[\text{shiftmu}, 1]}$$


$$\frac{(1 - 1 / ARLmu[\text{shiftmu}, 1]) \times 1 / ARLsigma[1]}{1 / ARLmusigma[\text{shiftmu}, 1]}$$


$$\frac{\left(1 - \frac{1}{475.1454}\right) * \frac{1}{500}}{\frac{1}{475.1454} + \frac{1}{500} - \frac{1}{475.1454} \times \frac{1}{500}}$$

3.09023
500.
14.7955
500.
250.25
243.878
475.145
500.

```

0.48673

0.48673

0.48673

Question 5 — Single sampling plan for attributes with RECTIFYING INSPECTION

```

Clear[Evaluate[Context[] <> "*"]];
p1 = 0.005; (* AQL *)
α = 1 - 0.95; (* producer's risk *)
p2 = 0.175; (* LTPD *)
β = 0.15; (* consumer's risk *)

Q[c_, x_] = Quantile[ChiSquareDistribution[2 × (c + 1)], x];
r[c_] =  $\frac{N[Q[c, 1 - \beta], 5]}{N[Q[c, \alpha], 5]}$ ;
i = 0;
While[r[i] >  $\frac{p_2}{p_1}$ , Print["Do not use acceptance number c=", i, " because r(c)=",
  N[Q[i, 1 - β], 5], "/", N[Q[i, α], 5], "=", r[i], ">  $\frac{p_2}{p_1}$ =",  $\frac{p_2}{p_1}$ ];
  i++]
Print["Use the acceptance number c=", i, " because r(c)=",
  N[Q[i, 1 - β], 5], "/", N[Q[i, α], 5], "=", r[i], "≤  $\frac{p_2}{p_1}$ =",  $\frac{p_2}{p_1}$ ]
samplesize[c_] = Ceiling[ $\frac{Q[i, 1 - \beta]}{2 \times p_2}$ ];
If[Ceiling[ $\frac{Q[i, 1 - \beta]}{2 \times p_2}$ ] ≤ Floor[ $\frac{Q[i, \alpha]}{2 \times p_1}$ ],
  Print["Use the sample size n=", samplesize[i], "."],
  Print["Houston, we have a problem!"]]

Do not use acceptance number c=0 because r(c)=3.79424/0.102587=36.9857 >  $\frac{p_2}{p_1}$ =35.
Use the acceptance number c=1 because r(c)=6.74488/0.710723=9.49017 ≤  $\frac{p_2}{p_1}$ =35.
Use the sample size n=20.

(* Unnecessary verification *)
ntot = 1000; (* lot size *)
exactdist[p_] = HypergeometricDistribution[samplesize[i], Round[ntot × p], ntot];
If[CDF[exactdist[p1], i] ≥ 1 - α && CDF[exactdist[p2], i] ≤ β,
  Print["The single sampling plan for attributes
  complies with the producer's and consumer's risk points."],
  Print["The single sampling plan for attributes does not comply
  with the producer's or the consumer's risk points."]]

The single sampling plan for attributes
  complies with the producer's and consumer's risk points.

```



```
(* Average Outgoing Quality (AOQ) or percentage of defective due to rectifying inspection in a
single sampling plan and using the binomial approximation to the acceptance probability *)
ATI[n_, c_, p_] = n * CDF[BinomialDistribution[n, p], c] +
  ntot * (1 - CDF[BinomialDistribution[n, p], c]);
midp = 0.1;
ATI[samplesize[i], i, midp]
(* Results using the tables *)
20 * 0.3917 + 1000 * (1 - 0.3917)
616.088
616.134
```

Question 5 — Single sampling plan for variables with UNKNOWN STANDARD DEVIATION

Exercício 7(b)

```
Clear[Evaluate[Context[] <> "*"]];
p1 = 0.01; (* AQL *)
α = 1 - 0.95; (* producer's risk *)
p2 = 0.05; (* LTPD *)
β = 0.1; (* consumer's risk *)

ns = 59;
ks = 1.937713;
```

```
gdist = NormalDistribution[0, 1];
Ω[x_] := Quantile[gdist, x];
```

$$\text{Round}\left[\frac{\Omega[1 - p_1] - k_s \times \sqrt{\frac{3 \times n_s - 4}{3 \times n_s - 3}}}{\sqrt{\frac{1 + \frac{3 \times n_s \times k_s^2}{6 \times n_s - 8}}{n_s}}}, 0.01\right]$$

```
CDF[NormalDistribution[0, 1], %]
```

$$\text{Round}\left[\frac{\Omega[1 - p_2] - k_s \times \sqrt{\frac{3 \times n_s - 4}{3 \times n_s - 3}}}{\sqrt{\frac{1 + \frac{3 \times n_s \times k_s^2}{6 \times n_s - 8}}{n_s}}}, 0.01\right]$$

```
CDF[NormalDistribution[0, 1], %]
```

```
1.77
```

```
0.961636
```

```
-1.29
```

```
0.0985253
```

```
U = 3;
sigma = 0.1;
dist = NormalDistribution[U + 2 * sigma, sigma];
data = RandomVariate[dist, ns + 1];
xbar = Round[Mean[data], 0.001]
sx = Round[ $\sqrt{\text{Variance}[data]}$ , 0.001]

$$\frac{U - \bar{x}}{s_x}$$

If[ $\frac{U - \bar{x}}{s_x} \geq k_s$ , Print["We should accept the lot."],
  Print["We should reject the lot."]]
3.207
0.12
-1.725
We should reject the lot.
```