

מכאן 2 :

1 :

אם $\hat{\theta}$ הוא הערכה של θ על בסיס M_1, \dots, M_n אזי $\hat{\theta} = g(M_1, \dots, M_n)$ ←

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אם $\tilde{g} = Tg$ (הערכה של \tilde{g} על בסיס M_1, \dots, M_n) אזי $\tilde{g} = Tg$ ←

$$(1) \tau(\hat{\theta}) = \tau(g(M_1, \dots, M_n)) = Tg(M_1, \dots, M_n)$$

$$(2) \tau(\theta) = \tau(g(M_1, \dots, M_n)) = Tg(M_1, \dots, M_n)$$

אם $\tau(\theta)$ היא הערכה של $\tau(\theta)$ על בסיס M_1, \dots, M_n אזי $\tau(\theta) = Tg(M_1, \dots, M_n)$

אם T_1, T_2 הם הערכות של θ על בסיס M_1, \dots, M_n אזי $T_1 = T_2$ ←

$$\theta = f(t_1, t_2, \dots, t_n)$$

$$\theta = g(t_1^2, t_2^2, \dots, t_n^2)$$

אם T_1, T_2 הם הערכות של θ על בסיס M_1, \dots, M_n אזי $T_1 = T_2$

$$T_1 = f(T_1^1, T_1^2, \dots, T_1^n)$$

$$T_2 = g(T_2^1, T_2^2, \dots, T_2^n)$$

$$T_3 = \frac{T_1 + T_2}{2} = \frac{1}{2} (f(T_1^1, T_1^2, \dots, T_1^n) + g(T_2^1, T_2^2, \dots, T_2^n))$$

$$\frac{1}{2} (f(t_1, t_2, \dots, t_n) + g(t_1^2, t_2^2, \dots, t_n^2)) = \frac{1}{2} (\theta + \theta) = \frac{1}{2} \cdot 2\theta = \theta$$

$$T_3 = \frac{T_1 + T_2}{2}$$

$$E(\bar{X}) = \frac{2}{\lambda} = M_1 = \bar{X} \Rightarrow \boxed{\frac{1}{\lambda} = \frac{\bar{X}}{2}}$$

$$\text{var}(X) = \frac{2}{\lambda^2} = M_2 - M_1^2 \Rightarrow$$

∴ $\frac{2}{\lambda^2} = c_1$

(1c)

$$T_1 = \left(\frac{\sum_{i=1}^n X_i}{n} \right)^2 = (\bar{X})^2 = \left(\frac{2}{\lambda} \right)^2 = \frac{4}{\lambda^2}$$

$$c_1 = \frac{1}{4} \quad \text{p.d.}$$

$$T_2 = \frac{\sum_{i=1}^n X_i^2}{n} = \frac{\sum_{i=1}^n (X_i^2 - \bar{X}^2 + \bar{X}^2)}{n} = \frac{\sum_{i=1}^n (X_i^2 - \bar{X}^2)}{n} + \frac{\sum_{i=1}^n \bar{X}^2}{n} =$$

$$\frac{\sum_{i=1}^n (X_i^2 - \bar{X}^2)}{n} + \bar{X}^2 = \frac{2}{\lambda^2} + \left(\frac{2}{\lambda} \right)^2 = \frac{2}{\lambda^2} + \frac{4}{\lambda^2} = \frac{6}{\lambda^2}$$

$$c_2 = \frac{1}{6} \quad \text{p.d.}$$

$$T_3 = \frac{\sum_{i=1}^n (X_i - \bar{X})^2}{n} = \frac{2}{\lambda^2} \Rightarrow \boxed{c_3 = \frac{1}{2}}$$

$$E(c_1 T_1) = \frac{1}{4} E(T_1) = \frac{1}{4} E(\bar{X}^2) = \frac{1}{4} [\text{var}(\bar{X}) + E(\bar{X})^2]$$

$$= \frac{1}{4} \cdot \left[\frac{2}{\lambda^2 n} + \left(\frac{2}{\lambda} \right)^2 \right] = \frac{1}{4} \left[\frac{2}{\lambda^2 n} + \frac{4}{\lambda^2} \right] = \frac{1}{2\lambda^2} \left[\frac{1}{n} + 2 \right]$$

$$= \frac{1}{2} \left[\frac{1+2n}{\lambda^2 n} \right]$$

$$E(c_2 T_2) = c_2 E(T_2) = \frac{1}{6} E(T_2) = \frac{1}{6} E\left(\frac{\sum_{i=1}^n X_i^2}{n} \right) =$$

$$\frac{1}{6} E\left(\frac{\sum_{i=1}^n (X_i^2 - \bar{X}^2 + \bar{X}^2)}{n} \right) = \frac{1}{6} \text{var}(X) + \frac{1}{6} E(\bar{X}^2) =$$

$$\frac{1}{6} \cdot \left[\frac{2}{\lambda^2} + \left(\frac{2}{\lambda} \right)^2 \right] = \frac{1}{6} \cdot \frac{6}{\lambda^2} = \frac{1}{\lambda^2}$$

∴ $\frac{1}{\lambda^2} = c_3$

$$E(c_3 T_3) = c_3 E(T_3) = c_3 E\left(\frac{n-1}{n-1} \cdot \frac{1}{n} \cdot \sum_{i=1}^n (X_i - \bar{X})^2 \right) = \frac{(n-1)}{n} \cdot \text{var}(X) \cdot c_3$$

$$\boxed{\frac{(n-1)}{n} \cdot \frac{1}{\lambda^2}}$$

$$= \frac{(n-1)}{n} \cdot \frac{2}{\lambda^2} \cdot c_3$$

$$T_1 = \bar{X}(\bar{X} - 1) = \mu_1(\mu_1 - 1) = \mu_1^2 - \mu_1 = g(\mu_1)$$

$$\mu_1 = E(X) = \frac{1}{p}$$

$$g(\mu_1) = \left(\frac{1}{p}\right)^2 - \frac{1}{p} = \frac{1-p}{p^2} = \text{var}(X)$$

$$T_2 = \frac{1}{n} \sum_{i=1}^n (X_i - \bar{X})^2 = \mu_2 - \mu_1^2 = g(\mu_1, \mu_2)$$

$$\mu_1 = E(X) = \frac{1}{p}$$

$$\mu_2 = E(X^2) = \text{var}(X) + E(X)^2 = \frac{1-p}{p^2} + \left(\frac{1}{p}\right)^2 = \frac{2-p}{p^2}$$

$$g(\mu_1, \mu_2) = \mu_2 - \mu_1^2 = \frac{2-p}{p^2} - \left(\frac{1}{p}\right)^2 = \frac{2-p}{p^2} - \frac{1}{p^2} = \frac{1-p}{p^2} = \text{var}(X)$$

$$T_3 = \frac{1}{2n} \sum_{i=1}^n X_i^2 - \frac{1}{2} \bar{X} = \frac{1}{2} \left[\frac{1}{n} \sum_{i=1}^n X_i^2 - \bar{X} \right] = \frac{1}{2} \left[\frac{1}{n} \sum_{i=1}^n (X_i^2 - \bar{X}^2 + \bar{X}^2 - \bar{X}) \right]$$

$$= \frac{1}{2} \left[\frac{1}{n} \sum_{i=1}^n (X_i^2 - \bar{X}^2) + \frac{1}{n} \sum_{i=1}^n \bar{X}^2 - \bar{X} \right]$$

$$= \frac{1}{2} [\mu_2 - \mu_1^2 + \bar{X}^2 - \bar{X}] = \frac{1}{2} [\mu_2 - \mu_1^2 + \mu_1^2 - \mu_1]$$

$$= \frac{1}{2} [\mu_2 - \mu_1] = f(\mu_1, \mu_2)$$

$$f(\mu_1, \mu_2) = \frac{1}{2} [\mu_2 - \mu_1] = \frac{1}{2} \left[\frac{2-p}{p^2} - \frac{1}{p} \right] = \frac{1}{2} \left[\frac{2-p-p}{p^2} \right] =$$

$$\frac{1}{2} \cdot \frac{2-2p}{p^2} = \frac{1-p}{p^2} = \text{var}(X)$$

$$E(T_1) = E(\bar{X}(\bar{X}-1)) = E(\bar{X}^2 - \bar{X}) = E(\bar{X}^2) - E(\bar{X}) \quad (2)$$

$$= \text{var}(\bar{X}) + E(\bar{X})^2 - E(\bar{X}) = \frac{1}{n} \text{var}(X) + E(X)^2 - E(X)$$

$$= \frac{1}{n} \left[\frac{(1-p)}{p^2} \right] + \left(\frac{1}{p} \right)^2 - \frac{1}{p} = \frac{1-p}{n \cdot p^2} + \frac{1}{p^2} - \frac{1}{p} =$$

$$\frac{1-p+n-np}{np^2} = \frac{(1-p) + n(1-p)}{np^2} = \frac{(1-p)}{p^2} \cdot \frac{n+1}{n} =$$

$$\frac{n+1}{n} \text{var}(X) \geq \text{var}(X)$$

$$T_1^* = \frac{n}{n+1} T_1$$

• 1174

$$E(T_2) = E\left(\frac{1}{n} \sum_{i=1}^n (X_i - \bar{X})^2\right) = E\left(\frac{1}{n} \sum_{i=1}^n [X_i - E(X) + E(X) - \bar{X}]^2\right)$$

$$= E\left[\frac{1}{n} \sum_{i=1}^n (X_i - E(X))^2 - (\bar{X} - E(X))^2\right] = E\left(\frac{1}{n} \sum_{i=1}^n (X_i - E(X))^2\right)$$

$$- E(\bar{X} - E(\bar{X}))^2 = \text{var}(X) - \text{var}(\bar{X}) = \text{var}(X) - \frac{1}{n} \text{var}(X)$$

$$= \frac{n-1}{n} \text{var}(X) \leq \text{var}(X)$$

$$T_2^* = \frac{n}{n-1} T_2$$

• 1174

$$E(T_3) = E\left(\frac{1}{2n} \sum_{i=1}^n X_i^2 - \frac{1}{2} \bar{X}\right) = \frac{1}{2} E\left(\frac{1}{n} \sum_{i=1}^n X_i^2\right) - \frac{1}{2} E(\bar{X})$$

$$= \frac{1}{2} [E(X^2) - E(X)] = \frac{1}{2} [\text{var}(X) + E(X)^2 - E(X)]$$

$$= \frac{1}{2} \left[\frac{(1-p)}{p^2} + \frac{1}{p^2} - \frac{1}{p} \right] = \frac{1}{2} \left[\frac{1-p+1-p}{p^2} \right] = \frac{1}{2} \cdot \frac{2-2p}{p^2} =$$

$$\frac{1}{2} \cdot \frac{2(1-p)}{p^2} = \frac{1-p}{p^2} = \text{var}(X) = \text{var}(X)$$

$$T_3^* = T_3$$

• 1174 1083

```

N = 3000
n = 10
p = 0.2

mat = matrix(0, n, N)

for (i in 1:N) {
  sample = 1 + rgeom(n, p)
  mat[,i] = sample
}

t1 = rep(0, N)
t2 = rep(0, N)
t3 = rep(0, N)
m = rep(0, N)

T1 = function (arg1) {
  temp1 = mean(arg1)
  return(temp1*(temp1 - 1))
}

T2 = function(arg2) {
  mean2 = mean(arg2)
  temp2 = arg2 - mean2
  temp2 = temp2 ^ 2
  return(sum(temp2) / length(temp2))
}

T3 = function (arg3) {
  temp3 = arg3 ^ 2
  return((sum(temp3) / (2*length(temp3))) - 0.5 * mean(arg3))
}

for (i in 1:N) {
  t1[i] = T1(mat[, i])
  t2[i] = T2(mat[, i])
  t3[i] = T3(mat[, i])
  m[i] = mean(mat[, i])
}

mean(t1)
mean(t2)
mean(t3)

```

$$\sigma^2 = \text{Var}(\text{Geo}(0.2)) = \frac{1-0.2}{(0.2)^2} = 20$$

3 of 10

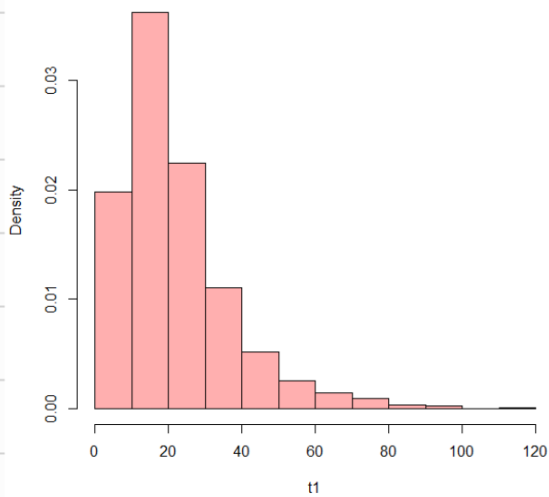
(2)

```

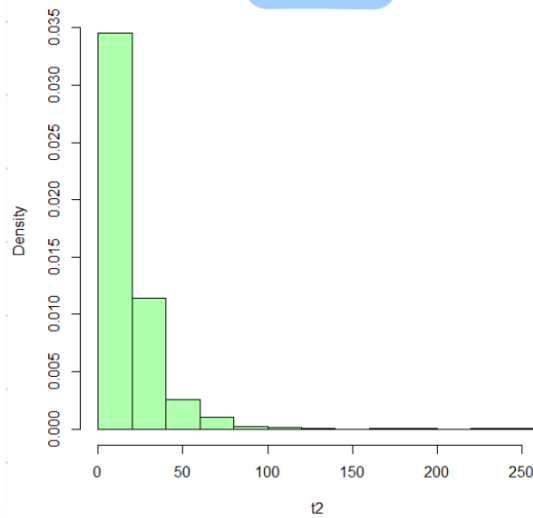
> mean(t1)
[1] 22.01214
> mean(t2)
[1] 18.01473
> mean(t3)
[1] 20.01343

```

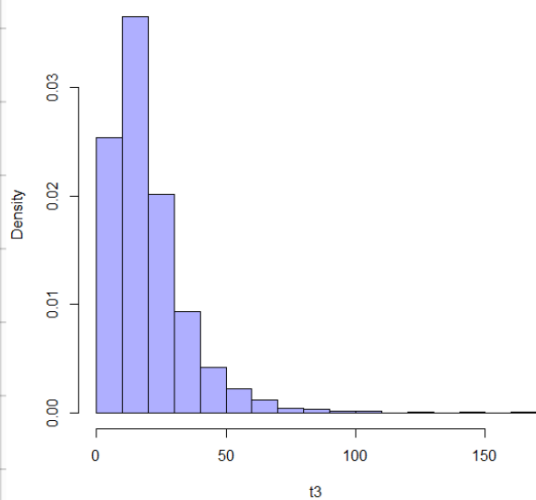
Histogram of t1



Histogram of t2



Histogram of t3



עבור $\delta = 10$

$$|t_i - \sigma^2| \leq \delta \iff t_i \in [\sigma^2 - \delta, \sigma^2 + \delta]$$

$$P(|t_i - \sigma^2| \leq \delta) = P(\sigma^2 - \delta \leq t_i \leq \sigma^2 + \delta)$$

$$= F_{t_i}(\sigma^2 + \delta) - F_{t_i}(\sigma^2 - \delta)$$

```
> sum(t1 > 0 & t1 < 30) / length(t1)
[1] 0.7663333
> sum(t1 > 0 & t1 < 10) / length(t1)
[1] 0.198
> 0.7663333 - 0.198 <- F_{t_i}(\sigma^2 + \delta) - F_{t_i}(\sigma^2 - \delta)
[1] 0.5683333
> sum(t1 > 10 & t1 < 30) / length(t1)
[1] 0.5683333
> sum(t1 > 10 & t1 < 30) / length(t1)
[1] 0.5683333
> sum(t2 > 10 & t2 < 30) / length(t2)
[1] 0.4743333
> sum(t3 > 10 & t3 < 30) / length(t3)
[1] 0.5663333
```

(3)

```
> snow <- c(1,1,3,3,3,4,5,5,5,5,5,5,5,5,6,10,10,10,15,15,15,15,20,20,20,20, 20,30,30,35,40,40,40,40,45,65)
> n.snow = length(snow)
>
> T1(snow)
[1] 278.279
> T2(snow)
[1] 233.0723
> T3(snow)
[1] 255.6757
>
> (n.snow/(n.snow+1)) * T1(snow)
[1] 270.9559
> (n.snow/(n.snow-1)) * T2(snow)
[1] 239.5465
> T3(snow)
[1] 255.6757
```

מיוזנים

```
> snow <- c(1,1,3,3,3,4,5,5,5,5,5,5,5,5,6,10,10,10,15,15,15,15,20,20,20,20, 20,30,30,35,40,40,40,40,45,65)
> n.snow = length(snow)
>
> sqrt(T1(snow))
[1] 16.6817
> sqrt(T2(snow))
[1] 15.26671
> sqrt(T3(snow))
[1] 15.98986
>
> sqrt((n.snow/(n.snow+1)) * T1(snow))
[1] 16.46074
> sqrt((n.snow/(n.snow-1)) * T2(snow))
[1] 15.47729
> sqrt(T3(snow))
[1] 15.98986
```

מיוזנים

אני חושב כי היעדר בין האומנים לבין הכסאות המיוזנים
בוא מסביר זאת י"ס