

# DS\_Assignment\_1

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```
library("ie2misc")
```

```
## Warning: package 'ie2misc' was built under R version 4.2.2
```

```
library("stats")
```

## Question:

Carry out *steps* 1 – 5 for the following distributions:

1.  $N(0, 1)$ ,
2.  $Cauchy(0, 1)$ ,
3.  $LogNormal(0, 1)$ ,
4.  $Gamma(1, \frac{1}{2})$ ,
5.  $Beta(3, 3)$ ,
6.  $Beta(3, 1)$ ,
7.  $Beta(\frac{1}{2}, 3)$ ,
8.  $t_5$ ,
9.  $t_{20}$ .

## Steps:

1. Draw a random sample of size  $n$  from  $f_\theta(x)$ .
2. Compute the following quantities:
  - i) the ratio  $\frac{\bar{x} - mode(x)}{\bar{x} - median(x)}$ ,
  - ii) the ratio  $\frac{MD_{\bar{x}}}{sd(x)}$ ,
  - iii) the ratio  $\frac{SIQR(x)}{SD(x)}$ ,
  - iv) % of observations in the interval  $\bar{x} \pm 6 \times SD$ ,
  - v) % of observations in the interval  $\bar{x} \pm 7.5 \times MD_{\bar{x}}$ ,
  - vi) % of observations in the interval  $\bar{x} \pm 9 \times QD$ .
3. Repeat *steps* 1 – 2 for  $R = 1000$  times. This will generate  $R$  values for each sets of quantities computed in *step* 2.
4. For each of the quantities obtained in *step* 2, repeat the minimum and maximum value in the form of an interval.
5. Carry out *steps* 1 – 4 for  $n = 10, 100, 1000, 5000$ .

Answer:

```
n = c(10, 100, 1000, 5000)
R = 1000
```

Data collection and preparation for manipulation:

```
data1 = array(NA, c(9, 10, 1000))
data2 = array(NA, c(9, 100, 1000))
data3 = array(NA, c(9, 1000, 1000))
data4 = array(NA, c(9, 5000, 1000))

data1[1, , ] = replicate(R, rnorm(n[1], 0, 1))
data1[2, , ] = replicate(R, rcauchy(n[1], 0, 1))
data1[3, , ] = replicate(R, rlnorm(n[1], 0, 1))
data1[4, , ] = replicate(R, rgamma(n[1], shape = 1, scale = 0.5))
data1[5, , ] = replicate(R, rbeta(n[1], 3, 3))
data1[6, , ] = replicate(R, rbeta(n[1], 3, 1))
data1[7, , ] = replicate(R, rbeta(n[1], 0.5, 3))
data1[8, , ] = replicate(R, rt(n[1], 5))
data1[9, , ] = replicate(R, rt(n[1], 20))

data2[1, , ] = replicate(R, rnorm(n[2], 0, 1))
data2[2, , ] = replicate(R, rcauchy(n[2], 0, 1))
data2[3, , ] = replicate(R, rlnorm(n[2], 0, 1))
data2[4, , ] = replicate(R, rgamma(n[2], shape = 1, scale = 0.5))
data2[5, , ] = replicate(R, rbeta(n[2], 3, 3))
data2[6, , ] = replicate(R, rbeta(n[2], 3, 1))
data2[7, , ] = replicate(R, rbeta(n[2], 0.5, 3))
data2[8, , ] = replicate(R, rt(n[2], 5))
data2[9, , ] = replicate(R, rt(n[2], 20))

data3[1, , ] = replicate(R, rnorm(n[3], 0, 1))
data3[2, , ] = replicate(R, rcauchy(n[3], 0, 1))
data3[3, , ] = replicate(R, rlnorm(n[3], 0, 1))
data3[4, , ] = replicate(R, rgamma(n[3], shape = 1, scale = 0.5))
data3[5, , ] = replicate(R, rbeta(n[3], 3, 3))
data3[6, , ] = replicate(R, rbeta(n[3], 3, 1))
data3[7, , ] = replicate(R, rbeta(n[3], 0.5, 3))
data3[8, , ] = replicate(R, rt(n[3], 5))
data3[9, , ] = replicate(R, rt(n[3], 20))

data4[1, , ] = replicate(R, rnorm(n[4], 0, 1))
data4[2, , ] = replicate(R, rcauchy(n[4], 0, 1))
data4[3, , ] = replicate(R, rlnorm(n[4], 0, 1))
data4[4, , ] = replicate(R, rgamma(n[4], shape = 1, scale = 0.5))
data4[5, , ] = replicate(R, rbeta(n[4], 3, 3))
data4[6, , ] = replicate(R, rbeta(n[4], 3, 1))
data4[7, , ] = replicate(R, rbeta(n[4], 0.5, 3))
data4[8, , ] = replicate(R, rt(n[4], 5))
data4[9, , ] = replicate(R, rt(n[4], 20))
```

## Functions initialization:

```
getmode <- function(v) #for mode calculation
{
  uniqv <- unique(v)
  uniqv[which.max(tabulate(match(v, uniqv)))]
}

f1 = function(data) #1_to_calculate_2(i)
{
  return(mean(data) - getmode(data)/(mean(data) - median(data)))
}
f2 = function(data) #2_to_calculate_2(ii)
{
  return(madstat(data)/sd(data))
}
f3 = function(data) #3_to_calculate_2(iii)
{
  return((quantile(data, 0.75)-quantile(data, 0.25))/2*sd(data))
}
f4 = function(data) #4_to_calculate_2(iv)
{
  return(100*sum(data > (mean(data) - 6*sd(data)) & data < (mean(data) + 6*sd(data)) )/1000)
}
f5 = function(data) #5_to_calculate_2(v)
{
  return(100*sum(data > (mean(data) - 7.5*madstat(data)) & data < (mean(data) + 7.5*madstat(data)) )/1000)
}
f6 = function(data) #6_to_calculate_2(vi)
{
  return(100*sum(data > (mean(data) - 9*(quantile(data, 0.75)-quantile(data, 0.25))/2) & data < (mean(data) + 9*(quantile(data, 0.75)-quantile(data, 0.25))/2) )/1000)
}
```

(i) For the ratio  $\frac{\bar{x} - \text{mode}(x)}{\bar{x} - \text{median}(x)}$ :

```
vect1 = array(NA, c(4, 9, 1000))
for (i in 1:9)
{
  for (j in 1:1000)
  {
    vect1[1, i, j] = f1(data1[i, , j]) #1
  }
}
for (i in 1:9)
{
  for (j in 1:1000)
  {
    vect1[2, i, j] = f1(data2[i, , j]) #2
  }
}
```

```

for (i in 1:9)
{
  for (j in 1:1000)
  {
    vect1[3, i, j] = f1(data3[i, , j])  #3
  }
}
for (i in 1:9)
{
  for (j in 1:1000)
  {
    vect1[4, i, j] = f1(data4[i, , j])  #4
  }
}

m1 = array(NA, c(4, 9, 2))
for (i in 1:4)
{
  for (j in 1:9)
  {
    m1[i, j, ] = c(min(vect1[i, j, ]), max(vect1[i, j, ]))
  }
}

data.frame("Sample_size" = n, "dist1" = m1[, 1, ], "dist2" = m1[, 2, ], "dist3" = m1[, 3, ], "dist4" = m1[, 4, ])

```

```

##   Sample_size   dist1.1   dist1.2   dist2.1   dist2.2   dist3.1   dist3.2
## 1          10  -2385.372   3235.396 -1260.477   1374.079 -4980.78074  461.129423
## 2          100 -4177.250 252518.940 -1838.584   1024.689  -30.27572   2.020648
## 3         1000 -9590.448 10506.006 -2616.712  91142.986  -26.74634   1.680487
## 4         5000 -59649.799 99839.764 -1350.426 19730.654  -22.34310   1.626133
##      dist4.1      dist4.2      dist5.1   dist5.2      dist6.1      dist6.2
## 1 -791.95250 3480.5689832  -3105.523  86486.96 -66044.452144 48904.43204
## 2  -34.22956   0.5981343  -42496.874  39097.97  -2091.136338   331.61820
## 3  -17.83711   0.5315052  -52680.869  64675.81    2.903958    31.81685
## 4  -20.58158   0.5026713 -192146.637  97899.79    2.991550    26.65018
##      dist7.1      dist7.2      dist8.1      dist8.2      dist9.1      dist9.2
## 1 -1186.93343 1161.3728197  -4981.359    943.6147  -1327.248    3166.241
## 2  -19.62041   0.1843997  -37542.472 100195.9177  -11348.913   53094.740
## 3  -14.05999   0.1496239  -11397.854 316128.7946  -37393.182    5770.495
## 4  -13.32226   0.1473737 -29422.936 17598.2859  -36662.627   390953.368

```

(ii) For the ratio  $\frac{MD_{\bar{x}}}{sd(x)}$ :

```

vect2 = array(NA, c(4, 9, 1000))
for (i in 1:9)
{
  for (j in 1:1000)
  {
    vect2[1, i, j] = f2(data1[i, , j])  #1
  }
}

```

```

}
for (i in 1:9)
{
  for (j in 1:1000)
  {
    vect2[2, i, j] = f2(data2[i, , j]) #2
  }
}
for (i in 1:9)
{
  for (j in 1:1000)
  {
    vect2[3, i, j] = f2(data3[i, , j]) #3
  }
}
for (i in 1:9)
{
  for (j in 1:1000)
  {
    vect2[4, i, j] = f2(data4[i, , j]) #4
  }
}

m2 = array(NA, c(4, 9, 2))
for (i in 1:4)
{
  for (j in 1:9)
  {
    m2[i, j, ] = c(min(vect2[i, j, ]), max(vect2[i, j, ]))
  }
}

```

```

data.frame("Sample_size" = n, "dist1" = m2[, 1, ], "dist2" = m2[, 2, ], "dist3" = m2[, 3, ], "dist4" = m2[, 4, ])

```

```

##   Sample_size  dist1.1  dist1.2  dist2.1  dist2.2  dist3.1  dist3.2
## 1          10 0.5815849 0.9380359 0.48687182 0.8974847 0.5409454 0.9066274
## 2          100 0.7023616 0.8617572 0.17210605 0.6619077 0.3494411 0.7999285
## 3         1000 0.7728000 0.8202457 0.06001619 0.3590393 0.3113774 0.6896659
## 4         5000 0.7875683 0.8078909 0.02463003 0.2377993 0.3795404 0.6490933
##   dist4.1  dist4.2  dist5.1  dist5.2  dist6.1  dist6.2  dist7.1
## 1 0.5478607 0.9191967 0.5987904 0.9290239 0.5854471 0.9155295 0.5568261
## 2 0.4782323 0.8636560 0.7635084 0.8737319 0.7458167 0.8724665 0.6711109
## 3 0.6774581 0.7729831 0.8042235 0.8467731 0.7936889 0.8383448 0.7410292
## 4 0.7124787 0.7579053 0.8185974 0.8345745 0.8070518 0.8268661 0.7579687
##   dist7.2  dist8.1  dist8.2  dist9.1  dist9.2
## 1 0.9333868 0.5535477 0.9104509 0.5607459 0.9139227
## 2 0.8507628 0.5309310 0.8365171 0.7110904 0.8437977
## 3 0.8009158 0.5805285 0.7750741 0.7584286 0.8047708
## 4 0.7854867 0.6812949 0.7554412 0.7752010 0.7964594

```

(iii) For the ratio  $\frac{SIQR(x)}{SD(x)}$ :

```
vect3 = array(NA, c(4, 9, 1000))
for (i in 1:9)
{
  for (j in 1:1000)
  {
    vect3[1, i, j] = f3(data1[i, , j]) #1
  }
}
for (i in 1:9)
{
  for (j in 1:1000)
  {
    vect3[2, i, j] = f3(data2[i, , j]) #2
  }
}
for (i in 1:9)
{
  for (j in 1:1000)
  {
    vect3[3, i, j] = f3(data3[i, , j]) #3
  }
}
for (i in 1:9)
{
  for (j in 1:1000)
  {
    vect3[4, i, j] = f3(data4[i, , j]) #4
  }
}

m3 = array(NA, c(4, 9, 2))
for (i in 1:4)
{
  for (j in 1:9)
  {
    m3[i, j, ] = c(min(vect3[i, j, ]), max(vect3[i, j, ]))
  }
}

data.frame("Sample_size" = n, "dist1" = m3[, 1, ], "dist2" = m3[, 2, ], "dist3" = m3[, 3, ], "dist4" = m3[, 4, ],
```

##	Sample_size	dist1.1	dist1.2	dist2.1	dist2.2	dist3.1	dist3.2
## 1	10	0.02902172	2.2484074	0.0497183	1174.442	0.03219825	22.210182
## 2	100	0.35855148	1.1161474	1.5482953	21091.341	0.43126469	5.188449
## 3	1000	0.56325700	0.7896943	8.2080569	2858325.559	1.01198837	3.515421
## 4	5000	0.62633537	0.7246010	14.8713182	18516.742	1.27366147	2.545687

##	dist4.1	dist4.2	dist5.1	dist5.2	dist6.1	dist6.2
## 1	0.006538689	1.0798837	0.002106836	0.08755664	0.001672582	0.11484747
## 2	0.052625234	0.2710546	0.015839954	0.03980542	0.013884537	0.05172857
## 3	0.107435582	0.1750219	0.021978540	0.03218715	0.022160075	0.03207551

```
## 4 0.122544163 0.1539028 0.024515873 0.02846028 0.025052779 0.02905401
##      dist7.1    dist7.2    dist8.1  dist8.2    dist9.1  dist9.2
## 1 0.0001609526 0.07072990 0.05393821 5.475987 0.04641612 3.0187895
## 2 0.0061472585 0.03196685 0.46711083 1.615937 0.35973515 1.2607216
## 3 0.0128489621 0.02068306 0.75460127 1.221756 0.60188121 0.8385299
## 4 0.0142481733 0.01772583 0.84741197 1.044987 0.66407517 0.7800005
```

(iv) For % of observations in the interval  $\bar{x} \pm 6 \times SD$ :

```
vect4 = array(NA, c(4, 9, 1000))
for (i in 1:9)
{
  for (j in 1:1000)
  {
    vect4[1, i, j] = f4(data1[i, , j]) #1
  }
}
for (i in 1:9)
{
  for (j in 1:1000)
  {
    vect4[2, i, j] = f4(data2[i, , j]) #2
  }
}
for (i in 1:9)
{
  for (j in 1:1000)
  {
    vect4[3, i, j] = f4(data3[i, , j]) #3
  }
}
for (i in 1:9)
{
  for (j in 1:1000)
  {
    vect4[4, i, j] = f4(data4[i, , j]) #4
  }
}

m4 = array(NA, c(4, 9, 2))
for (i in 1:4)
{
  for (j in 1:9)
  {
    m4[i, j, ] = c(min(vect4[i, j, ]), max(vect4[i, j, ]))
  }
}

data.frame("Sample_size" = n, "dist1" = m4[, 1, ], "dist2" = m4[, 2, ], "dist3" = m4[, 3, ], "dist4" = m4[, 4, ])

## Sample_size dist1.1 dist1.2 dist2.1 dist2.2 dist3.1 dist3.2 dist4.1 dist4.2
## 1      10      1      1      1.0      1.0      1.0      1.0      1.0      1
```

```
## 2      100      10      10      9.8      10.0      9.9      10.0      9.9      10
## 3      1000     100     100     99.0     99.9     99.1     100.0     99.6     100
## 4      5000     500     500    496.9    499.9    497.3    499.4    498.7     500
## dist5.1 dist5.2 dist6.1 dist6.2 dist7.1 dist7.2 dist8.1 dist8.2 dist9.1
## 1      1      1      1      1      1      1      1.0      1      1.0
## 2      10      10      10      10      10      10      9.9      10     10.0
## 3      100     100     100     100     100     100     99.7     100     99.9
## 4      500     500     500     500     500     500    499.2     500    499.9
## dist9.2
## 1      1
## 2      10
## 3      100
## 4      500
```

(v) For % of observations in the interval  $\bar{x} \pm 7.5 \times MD_{\bar{x}}$ :

```
vect5 = array(NA, c(4, 9, 1000))
for (i in 1:9)
{
  for (j in 1:1000)
  {
    vect5[1, i, j] = f5(data1[i, , j]) #1
  }
}
for (i in 1:9)
{
  for (j in 1:1000)
  {
    vect5[2, i, j] = f5(data2[i, , j]) #2
  }
}
for (i in 1:9)
{
  for (j in 1:1000)
  {
    vect5[3, i, j] = f5(data3[i, , j]) #3
  }
}
for (i in 1:9)
{
  for (j in 1:1000)
  {
    vect5[4, i, j] = f5(data4[i, , j]) #4
  }
}

m5 = array(NA, c(4, 9, 2))
for (i in 1:4)
{
  for (j in 1:9)
  {
    m5[i, j, ] = c(min(vect5[i, j, ]), max(vect5[i, j, ]))
  }
}
```



```

    }
}

data.frame("Sample_size" = n, "dist1" = m5[, 1, ], "dist2" = m5[, 2, ], "dist3" = m5[, 3, ], "dist4" = m5[, 4, ])

```

```

## Sample_size dist1.1 dist1.2 dist2.1 dist2.2 dist3.1 dist3.2 dist4.1 dist4.2
## 1 10 1 1 1.0 1.0 1.0 1.0 1.0 1.0
## 2 100 10 10 9.5 10.0 9.7 10.0 9.8 10.0
## 3 1000 100 100 97.2 99.9 98.4 99.8 99.5 100.0
## 4 5000 500 500 489.2 499.9 494.7 497.6 498.3 499.9
## dist5.1 dist5.2 dist6.1 dist6.2 dist7.1 dist7.2 dist8.1 dist8.2 dist9.1
## 1 1 1 1 1 1 1 1.0 1 1.0
## 2 10 10 10 10 10 10 9.8 10 9.9
## 3 100 100 100 100 100 100 99.4 100 99.9
## 4 500 500 500 500 500 500 498.8 500 499.8
## dist9.2
## 1 1
## 2 10
## 3 100
## 4 500

```

(vi) For % of observations in the interval  $\bar{x} \pm 9 \times QD$ :

```

vect6 = array(NA, c(4, 9, 1000))
for (i in 1:9)
{
  for (j in 1:1000)
  {
    vect6[1, i, j] = f6(data1[i, , j]) #1
  }
}
for (i in 1:9)
{
  for (j in 1:1000)
  {
    vect6[2, i, j] = f6(data2[i, , j]) #2
  }
}
for (i in 1:9)
{
  for (j in 1:1000)
  {
    vect6[3, i, j] = f6(data3[i, , j]) #3
  }
}
for (i in 1:9)
{
  for (j in 1:1000)
  {
    vect6[4, i, j] = f6(data4[i, , j]) #4
  }
}

```

```

}

m6 = array(NA, c(4, 9, 2))
for (i in 1:4)
{
  for (j in 1:9)
  {
    m6[i, j, ] = c(min(vect6[i, j, ]), max(vect6[i, j, ]))
  }
}

data.frame("Sample_size" = n, "dist1" = m6[, 1, ], "dist2" = m6[, 2, ], "dist3" = m6[, 3, ], "dist4" = m6[, 4, ])

```

```

##   Sample_size dist1.1 dist1.2 dist2.1 dist2.2 dist3.1 dist3.2 dist4.1 dist4.2
## 1         10      0.7      1      0      1.0      0.0      1.0      0.7      1.0
## 2        100     10.0     10      0      9.9      9.3     10.0      9.5     10.0
## 3       1000    100.0    100      0     95.4     96.8     99.2     98.9    100.0
## 4       5000   500.0    500      0    469.7    488.4    494.1    497.5    499.8
##   dist5.1 dist5.2 dist6.1 dist6.2 dist7.1 dist7.2 dist8.1 dist8.2 dist9.1
## 1      0.8      1      0.8      1      0.6      1      0.7      1      0.8
## 2     10.0     10     10.0     10      9.7     10      9.8     10      9.9
## 3    100.0    100    100.0    100     99.8    100     99.3    100     99.9
## 4    500.0    500    500.0    500    499.9    500    498.5    500    499.9
##   dist9.2
## 1        1
## 2       10
## 3      100
## 4     500

```