

## Assignment 2 - Ram Yoogesh (20867060)

### Question 1

(A)

From the question we can consider that

$$p = Pr(Y \leq Q_Y(p))$$

Upon substituting the values of Y we get,

$$p = Pr(aX + b \leq Q_Y(p))$$

$$p = Pr(X \leq \frac{Q_Y(p) - b}{a})$$

$$p = F_X(\frac{Q_Y(p) - b}{a})$$

From the above 2 equations, we can write this as

$$F_X^{-1}(p) = F_X^{-1}(F_X(\frac{Q_Y(p) - b}{a}))$$

$$Q_Y(p) = aQ_X(p) + b$$

Hence proved.

(b)

$$Pr(Q_X(U) \leq x) = Pr(F_X^{-1}(U) \leq x)$$

$$Pr(Q_X(U) \leq x) = Pr(F_X^{-1}(U) \leq x)$$

$$Pr(Q_X(U) \leq x) = Pr(U \leq F_X(x))$$

$$Pr(Q_X(U) \leq x) = G_U(F_X(x))$$

$$Pr(Q_X(U) \leq x) = F_X(x)$$

(c) (i)

Writing the `r_unifgenFX` function to return random derivatives

```
r_unifgenFX <- function(n, qfunction = qnorm)
{
  randomderivates <- runif(n, min =0, max = 1)
  return (randomderivates)
}
```

(c) (ii)

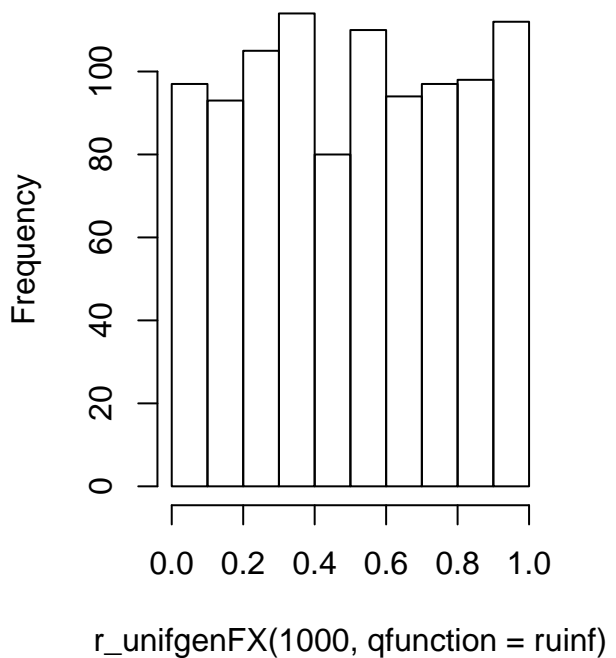
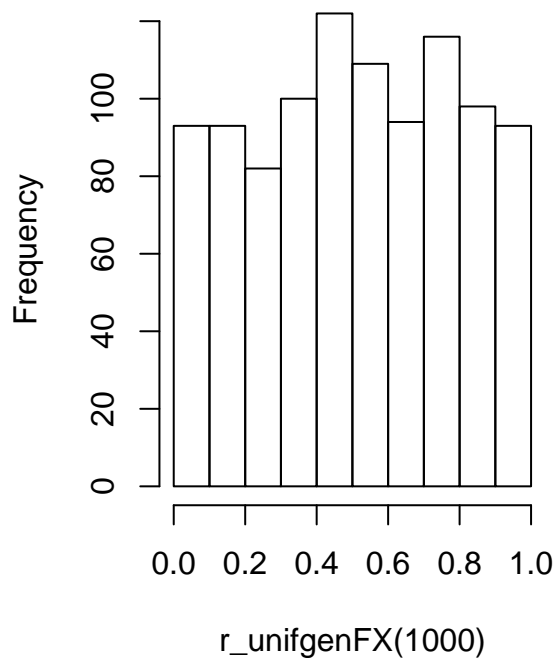
Executing the following snippets as instructed

```
set.seed(1234567)

oldpar <- par(mfrow = c(1, 2))
hist(r_unifgenFX(1000))

hist(r_unifgenFX(1000, qfunction = ruinf))
```

**Histogram of `r_unifgenFX(1000)`gram of `r_unifgenFX(1000, qfunction`**



```
par(oldpar)
```

(c)(iii)

Generating pseudo-random observations from a student t distribution

```
r_unifgen <- function (n, dof)
{
  robs_student_t <- rt(n, df=dof)
  return (robs_student_t)
}
```

```
sam_obs <- r_unifgen(1000, 3)
```

```
# Now passing the vector of random values to the qt fucntion to find the quantile values.
```

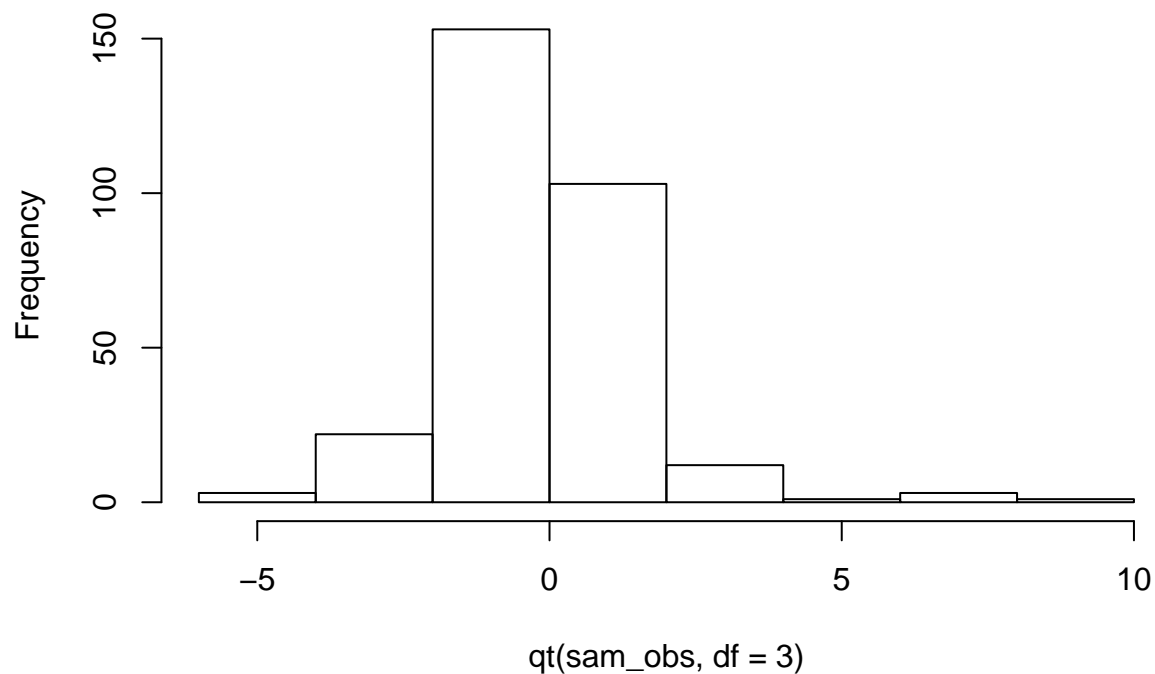
```
x_val <- qt(sam_obs, df= 3, lower.tail = TRUE)
```

```
## Warning in qt(sam_obs, df = 3, lower.tail = TRUE): NaNs produced
```

```
hist(qt(sam_obs, df = 3))
```

```
## Warning in qt(sam_obs, df = 3): NaNs produced
```

**Histogram of qt(sam\_obs, df = 3)**



(d)

```
# Just analyzing the dataset before interpreting the questions !
```

```
str(mtcars)
```

```
## 'data.frame': 32 obs. of 11 variables:
## $ mpg : num 21 21 22.8 21.4 18.7 18.1 14.3 24.4 22.8 19.2 ...
## $ cyl : num 6 6 4 6 8 6 8 4 4 6 ...
## $ disp: num 160 160 108 258 360 ...
## $ hp : num 110 110 93 110 175 105 245 62 95 123 ...
## $ drat: num 3.9 3.9 3.85 3.08 3.15 2.76 3.21 3.69 3.92 3.92 ...
## $ wt : num 2.62 2.88 2.32 3.21 3.44 ...
## $ qsec: num 16.5 17 18.6 19.4 17 ...
## $ vs : num 0 0 1 1 0 1 0 1 1 1 ...
## $ am : num 1 1 1 0 0 0 0 0 0 0 ...
## $ gear: num 4 4 4 3 3 3 3 4 4 4 ...
## $ carb: num 4 4 1 1 2 1 4 2 2 4 ...
```

```
# we can infer that it has 11 variables and 32 rows . Names of these are mpg, cyl, disp, hp, drat, wt,
head(mtcars)
```

```
##           mpg cyl disp  hp drat   wt  qsec vs am gear carb
## Mazda RX4      21.0   6  160 110 3.90 2.620 16.46 0  1    4    4
## Mazda RX4 Wag  21.0   6  160 110 3.90 2.875 17.02 0  1    4    4
## Datsun 710      22.8   4  108  93 3.85 2.320 18.61 1  1    4    1
## Hornet 4 Drive  21.4   6  258 110 3.08 3.215 19.44 1  0    3    1
## Hornet Sportabout 18.7   8  360 175 3.15 3.440 17.02 0  0    3    2
## Valiant         18.1   6  225 105 2.76 3.460 20.22 1  0    3    1
```

(d)(i)

In order to answer this question ,lets use the summary function to get some insights. We can infer from summary of qsec ( min value- 14.50, 1st quartile - 16.89 and so on).

```
summary(mtcars)
```

```
##           mpg           cyl           disp           hp
## Min.   :10.40   Min.   :4.000   Min.   : 71.1   Min.   : 52.0
## 1st Qu.:15.43   1st Qu.:4.000   1st Qu.:120.8   1st Qu.: 96.5
## Median :19.20   Median :6.000   Median :196.3   Median :123.0
## Mean   :20.09   Mean   :6.188   Mean   :230.7   Mean   :146.7
## 3rd Qu.:22.80   3rd Qu.:8.000   3rd Qu.:326.0   3rd Qu.:180.0
## Max.   :33.90   Max.   :8.000   Max.   :472.0   Max.   :335.0
##           drat           wt           qsec           vs
## Min.   : 2.760   Min.   :1.513   Min.   :14.50   Min.   :0.0000
## 1st Qu.: 3.080   1st Qu.:2.581   1st Qu.:16.89   1st Qu.:0.0000
## Median : 3.695   Median :3.325   Median :17.71   Median :0.0000
```

```
## Mean :3.597 Mean :3.217 Mean :17.85 Mean :0.4375
## 3rd Qu.:3.920 3rd Qu.:3.610 3rd Qu.:18.90 3rd Qu.:1.0000
## Max. :4.930 Max. :5.424 Max. :22.90 Max. :1.0000
##      am      gear      carb
## Min. :0.0000 Min. :3.000 Min. :1.000
## 1st Qu.:0.0000 1st Qu.:3.000 1st Qu.:2.000
## Median :0.0000 Median :4.000 Median :2.000
## Mean :0.4062 Mean :3.688 Mean :2.812
## 3rd Qu.:1.0000 3rd Qu.:4.000 3rd Qu.:4.000
## Max. :1.0000 Max. :5.000 Max. :8.000
```

```
# Now lets use the Quantile function to answer the question
quantile(mtcars$qsec)
```

```
##      0%      25%      50%      75%     100%
## 14.5000 16.8925 17.7100 18.9000 22.9000
```

Qsec - The variable denotes the 1/4 mile time took by cars.

Quantile function basically produces the values at the respective quartiles. Say 0, 25, 50, 75 and 100. What it means for Qsec is that at 25th percent quartile the value of qsec is 16.89 (It tooks a car 16.89s to reach 1/4 mile). At 50th percent, also called as the median of the dataset, the value at 50th percent quartile is 17.71s, likewise for the remaining quartiles.

d(ii)

we can pass `mtcars$qsec` in the X argument and for the probability we can use `runif` to generate 1000 observations

```
samobs_fromqsec <- quantile(mtcars$qsec, prob = runif(1000))
samobs_fromqsec
```

```
## 82.56272019% 55.43537533% 25.26772677% 28.88000016% 56.21380128% 48.23564820%
## 19.45783 17.98370 16.89499 17.01434 17.98853 17.59155
## 27.80648707% 93.64463692% 16.56456550% 1.18075679% 2.22223180% 48.60856393%
## 16.97440 20.01627 16.49240 14.53660 14.56889 17.61510
## 39.00647471% 17.17150714% 96.79571246% 23.35539414% 60.25798912% 2.62663255%
## 17.30920 16.53756 20.23788 16.87721 18.20399 14.58143
## 15.67640505% 31.63931684% 11.19827144% 60.73381805% 4.63054057% 97.24547171%
## 16.37301 17.02000 15.66030 18.24825 14.95273 20.61154
## 68.23912810% 68.41987553% 35.84659728% 52.15265874% 86.24651006% 41.70240827%
## 18.60154 18.60210 17.07811 17.84677 19.78666 17.39277
## 23.88116326% 83.79194455% 8.69364745% 76.68455190% 4.22513322% 79.65354961%
## 16.88209 19.46927 15.47255 18.90000 14.85093 19.27400
## 37.64747565% 8.97748205% 47.33024519% 84.31135849% 67.42749871% 38.16132473%
## 17.21768 15.48047 17.54103 19.52870 18.59220 17.25750
## 14.64794646% 4.94745784% 33.35952470% 12.02421337% 91.91094527% 25.07700184%
## 16.17534 15.03231 17.03024 15.74735 20.00492 16.89322
```

##	59.71529905%	29.30435641%	73.51423618%	33.78774493%	26.55488406%	97.56579010%
##	18.15352	17.02000	18.83893	17.03423	16.92784	20.87766
##	25.48536612%	29.33372608%	53.45269199%	65.57387246%	69.89205494%	34.77731368%
##	16.89701	17.02000	17.91125	18.54623	18.60667	17.04343
##	43.90856060%	17.13485958%	19.73887279%	25.05627696%	8.32081330%	66.50878717%
##	17.41223	16.53483	16.72024	16.89302	15.46215	18.56942
##	55.26927833%	26.58797766%	91.64741621%	5.32587986%	19.63268633%	61.39929765%
##	17.98267	16.92907	20.00411	15.12733	16.71464	18.30743
##	18.96279245%	82.42500224%	0.58414079%	86.01700889%	29.71542988%	57.77933504%
##	16.67083	19.45655	14.51811	19.75607	17.02000	17.99823
##	49.98949510%	36.16389406%	23.97085091%	63.16237934%	39.99208040%	97.30362794%
##	17.70928	17.10270	16.88293	18.42767	17.33975	20.65985
##	76.90825551%	88.04478203%	9.35689837%	47.29932379%	61.35180183%	61.18349247%
##	18.90000	19.92939	15.49106	17.53930	18.30419	18.29006
##	81.59333244%	97.56507545%	41.85280132%	76.74217701%	99.85544763%	99.14391234%
##	19.44882	20.87706	17.39744	18.90000	22.77991	22.18876
##	35.03837977%	4.08151275%	96.18970093%	13.72014924%	54.62887629%	9.24995951%
##	17.04586	14.81487	20.18195	15.99701	17.96959	15.48807
##	9.57933494%	86.29760696%	52.09488226%	81.45562124%	58.72106701%	12.85156454%
##	15.49726	19.79347	17.84391	19.44754	18.06106	15.83455
##	50.87438170%	72.53598233%	13.05597590%	4.89486768%	44.62784743%	97.92759221%
##	17.76963	18.75098	15.86936	15.01910	17.41669	21.17824
##	74.42114935%	89.96940241%	54.45640450%	90.44165486%	59.44494465%	42.64210395%
##	18.90000	19.98905	17.96104	20.00037	18.12838	17.40438
##	89.97218476%	63.65564424%	98.26391332%	3.76559685%	15.85332071%	59.63822310%
##	19.98914	18.46131	21.45766	14.73554	16.40701	18.14635
##	70.40746557%	19.02837143%	30.85463264%	12.50515957%	29.92917513%	11.11923300%
##	18.60826	16.67571	17.02000	15.79804	17.02000	15.65197
##	16.12096806%	13.10673954%	60.93945040%	18.51580143%	3.11191664%	42.97619576%
##	16.45845	15.87912	18.26737	16.63758	14.59647	17.40645
##	90.81213993%	55.85000983%	36.22468703%	55.41786146%	28.97401412%	83.11288627%
##	20.00152	17.98627	17.10741	17.98359	17.01783	19.46295
##	34.28008170%	45.95526094%	18.82636116%	27.78894154%	10.51862428%	50.46582369%
##	17.03880	17.46430	16.66068	16.97375	15.58866	17.74177
##	40.00436899%	46.61056800%	61.94783424%	52.03986771%	91.47919260%	79.96346029%
##	17.34014	17.50087	18.34484	17.84118	20.00359	19.32588
##	41.47342306%	32.03756956%	93.84348404%	30.03271334%	59.69637299%	23.96383011%
##	17.38568	17.02000	20.02921	17.02000	18.15176	16.88286
##	10.13508830%	44.45318650%	60.48490664%	15.83801790%	42.91884473%	58.31603471%
##	15.54824	17.41561	18.22510	16.40407	17.40610	18.02339
##	21.68378509%	19.76913752%	97.30679605%	60.18004464%	21.48169677%	54.16405422%
##	16.82274	16.72183	20.66249	18.19674	16.81209	17.94654
##	30.52252387%	52.81632212%	7.18180772%	32.22636990%	68.67917217%	12.07703731%
##	17.02000	17.87969	15.43037	17.02000	18.60291	15.75292
##	21.96024337%	56.93879572%	39.68072014%	77.84693227%	7.78091645%	96.77311194%
##	16.83730	17.99302	17.33010	18.97158	15.44709	20.21993
##	62.56218953%	5.35220518%	76.08212270%	72.01643141%	28.49725410%	3.43481961%
##	18.38674	15.13394	18.90000	18.70428	17.00010	14.65248
##	4.25333907%	85.87827445%	55.56464666%	56.16303030%	54.82415550%	89.18526161%
##	14.85801	19.73757	17.98450	17.98821	17.97928	19.96474
##	39.17641195%	6.61698075%	0.52501378%	60.79356549%	52.81056943%	30.50776182%
##	17.31447	15.41461	14.51628	18.25380	17.87940	17.02000
##	10.21651097%	31.64146643%	70.74450289%	79.02869396%	57.32117717%	17.50590673%
##	15.55682	17.02000	18.60931	19.16940	17.99539	16.56244

##	80.33520530%	27.67639011%	78.53511046%	84.19519907%	60.64314751%	97.29950440%
##	19.38811	16.96956	19.08678	19.51322	18.23981	20.65643
##	14.43649335%	44.71281979%	42.56872064%	48.42840463%	12.77212040%	59.03215043%
##	16.13469	17.41722	17.40393	17.60282	15.82618	18.08999
##	99.94138249%	35.24499438%	46.80802277%	56.60681063%	29.94399627%	25.77130969%
##	22.85130	17.04778	17.51189	17.99096	17.02000	16.89967
##	16.32712027%	99.45530235%	22.34047703%	96.98357540%	1.75254641%	63.25369675%
##	16.47474	22.44747	16.85734	20.39395	14.55433	18.43390
##	3.89308827%	3.83684309%	65.16432751%	73.16015372%	26.20007875%	40.79077656%
##	14.76755	14.75343	18.53608	18.80710	16.91464	17.36451
##	98.29375474%	73.41952804%	63.97085178%	88.11367962%	4.58539762%	47.65324581%
##	21.48245	18.83042	18.48281	19.93152	14.94139	17.55905
##	62.49401374%	30.27512734%	71.14703904%	29.84693730%	42.12123554%	6.29242084%
##	18.38209	17.02000	18.62612	17.02000	17.40115	15.37003
##	95.52195999%	2.50572101%	86.83456127%	59.68835352%	38.12123768%	8.11396807%
##	20.13848	14.57768	19.86505	18.15102	17.25440	15.45638
##	9.58540177%	95.15380803%	6.21218453%	78.64528128%	64.12392831%	43.46727473%
##	15.49743	20.11451	15.34988	19.10522	18.49325	17.40950
##	78.74434425%	95.78504299%	67.11769404%	11.66728989%	71.27704201%	13.61519306%
##	19.12180	20.15561	18.58452	15.70973	18.63781	15.97684
##	3.32151200%	57.99432388%	35.22213146%	21.98217737%	71.34997507%	3.56033260%
##	14.62403	17.99956	17.04757	16.83846	18.64436	14.68400
##	72.32125557%	27.39335718%	91.71770946%	33.38422896%	99.33601078%	44.12672729%
##	18.73168	16.95903	20.00432	17.03047	22.34836	17.41359
##	32.38993089%	82.13553496%	0.13211018%	35.75784068%	15.85842727%	69.85904539%
##	17.02123	19.45386	14.50410	17.07123	16.40799	18.60656
##	28.75272119%	92.32198375%	75.85187803%	52.15039302%	96.91810017%	32.01746123%
##	17.00960	20.00620	18.90000	17.84666	20.33956	17.02000
##	36.59960355%	60.72604805%	55.74902296%	60.73641607%	52.42969303%	67.94086182%
##	17.13647	18.24752	17.98564	18.24849	17.86051	18.60062
##	94.38108227%	97.19274095%	56.16172699%	76.81455724%	33.90080368%	20.82112569%
##	20.06421	20.56773	17.98820	18.90000	17.03528	16.77727
##	72.53334927%	36.64303324%	22.17044423%	60.30403329%	69.60928398%	53.59754032%
##	18.75075	17.13984	16.84838	18.20828	18.60579	17.91844
##	81.12893058%	9.77212514%	41.87956522%	47.28574241%	33.18166195%	67.44142794%
##	19.44450	15.50998	17.39827	17.53854	17.02859	18.59255
##	82.93805579%	48.08529613%	1.30763687%	76.81563040%	6.92166009%	33.77703801%
##	19.46132	17.58316	14.54054	18.90000	15.42311	17.03413
##	14.39823096%	1.50932721%	99.53401121%	3.05180331%	79.75445329%	31.66873844%
##	16.12734	14.54679	22.51286	14.59461	19.29090	17.02000
##	34.37210552%	72.88307224%	8.10276193%	9.35531305%	78.51319511%	10.54438695%
##	17.03966	18.78219	15.45607	15.49101	19.08311	15.59138
##	72.84079292%	86.44529721%	1.40377271%	34.13213578%	11.44323149%	7.16045424%
##	18.77839	19.81316	14.54352	17.03743	15.68612	15.42978
##	19.58475069%	76.13811439%	90.71499419%	52.21936542%	11.03495799%	26.99951492%
##	16.71212	18.90000	20.00122	17.85008	15.64308	16.94438
##	26.58121542%	76.58006530%	60.09711921%	4.26129107%	45.40479179%	90.48830299%
##	16.92882	18.90000	18.18903	14.86001	17.43359	20.00051
##	18.30044126%	27.45125513%	83.46215112%	5.39561377%	34.88374229%	95.29818657%
##	16.62155	16.96119	19.46620	15.14484	17.04442	20.12391
##	28.02474026%	77.93569483%	44.25319396%	91.97108389%	95.98682306%	37.29054451%
##	16.98252	18.98644	17.41437	20.00511	20.16874	17.19002
##	75.06954707%	69.45112473%	14.14768214%	87.51211669%	73.04591804%	68.16971598%
##	18.90000	18.60530	16.07918	19.91288	18.79683	18.60133

##	96.62130100%	46.81982549%	22.41111279%	37.31333553%	13.41029117%	72.31173571%
##	20.21005	17.51255	16.86107	17.19178	15.93746	18.73083
##	24.01478505%	6.60919521%	93.22497279%	4.72624635%	26.77474464%	0.79783136%
##	16.88334	15.41440	20.00900	14.97676	16.93602	14.52473
##	2.78722225%	77.10185470%	60.30889172%	20.73760142%	56.35227684%	53.29456567%
##	14.58640	18.90000	18.20873	16.77287	17.98938	17.90341
##	95.04546374%	3.24914595%	70.16517220%	13.15136352%	72.33748881%	0.16035712%
##	20.10746	14.60586	18.60751	15.88769	18.73314	14.50497
##	28.34046830%	47.39938923%	87.03422311%	34.36332927%	45.03041860%	21.73480587%
##	16.99427	17.54489	19.89166	17.03958	17.41919	16.82542
##	16.61894140%	14.54678690%	53.45341756%	6.95348750%	12.78197691%	69.79420648%
##	16.49645	16.15589	17.91129	15.42400	15.82722	18.60636
##	72.62953056%	22.03578202%	10.66227849%	96.46102663%	39.27409495%	38.01164555%
##	18.75939	16.84129	15.60380	20.19961	17.31750	17.24590
##	58.38360724%	34.80120264%	73.31172081%	22.90898398%	53.42700039%	1.84633748%
##	18.02968	17.04365	18.82072	16.87305	17.90998	14.55724
##	15.08750771%	63.29574878%	17.24004459%	33.91181941%	30.51505988%	93.33004639%
##	16.25982	18.43677	16.54266	17.03538	17.02000	20.00932
##	46.75896422%	70.92740706%	95.40609394%	84.90456820%	27.15169690%	87.29395294%
##	17.50915	18.60987	20.13094	19.60778	16.95004	19.90611
##	38.25742987%	13.65208260%	65.83302207%	9.65588801%	80.08583849%	2.32546174%
##	17.26495	15.98393	18.55266	15.49940	19.34637	14.57209
##	58.55699237%	55.30167094%	89.49665471%	76.27238836%	82.44216554%	79.31343431%
##	18.04580	17.98287	19.97440	18.90000	19.45671	19.21707
##	18.22255831%	60.08912972%	93.48142892%	17.35302303%	41.98244275%	12.19641664%
##	16.61576	18.18829	20.00979	16.55106	17.40029	15.76550
##	26.47810658%	42.74377776%	80.79843645%	95.90880473%	4.35019971%	97.45289234%
##	16.92499	17.40501	19.44143	20.16366	14.88234	20.78386
##	73.61760666%	69.47557076%	25.35871693%	25.29182946%	12.13420972%	67.97527664%
##	18.84822	18.60537	16.89584	16.89521	15.75895	18.60072
##	7.65236567%	56.63355999%	35.78151937%	45.26173177%	53.13574385%	61.09007385%
##	15.44350	17.99113	17.07307	17.42560	17.89553	18.28138
##	39.66444773%	36.98221291%	79.51229669%	93.28805828%	73.40476594%	88.85469981%
##	17.32960	17.16612	19.25036	20.00919	18.82909	19.95450
##	93.58648648%	17.18177774%	10.95236172%	76.61375646%	47.08925670%	74.05835572%
##	20.01248	16.53832	15.63438	18.90000	17.52758	18.88785
##	74.75191192%	42.31789222%	26.81746774%	87.59383108%	45.83986071%	22.99319741%
##	18.90000	17.40237	16.93761	19.91541	17.45786	16.87384
##	1.41022040%	67.78565834%	98.73533929%	73.29129106%	37.71403006%	29.70401703%
##	14.54372	18.60014	21.84932	18.81889	17.22284	17.02000
##	30.47169046%	9.74121834%	95.07730026%	42.72531895%	47.90643668%	76.87365452%
##	17.02000	15.50672	20.10953	17.40490	17.57318	18.90000
##	47.22851661%	75.55451011%	75.90153133%	55.41112577%	54.00159298%	34.52233209%
##	17.53535	18.90000	18.90000	17.98355	17.93848	17.04106
##	47.08693470%	12.83334901%	27.03833464%	15.93411532%	9.90233915%	63.29211826%
##	17.52745	15.83263	16.94583	16.42254	15.52371	18.43652
##	16.77585389%	31.74803695%	70.65067308%	29.44693493%	37.09916805%	42.24277923%
##	16.50812	17.02000	18.60902	17.02000	17.17519	17.40191
##	8.54225080%	34.23246755%	19.08565883%	36.18063692%	93.47018781%	82.50684077%
##	15.46833	17.03836	16.67997	17.10400	20.00976	19.45731
##	63.21110416%	72.14571137%	62.12100883%	66.16487014%	67.66258357%	57.21479284%
##	18.43100	18.71590	18.35665	18.56089	18.59803	17.99473
##	5.36231538%	38.48180657%	53.67963444%	10.90712373%	98.32699744%	45.25751041%
##	15.13648	17.28234	17.92251	15.62961	21.51007	17.42537



##	22.78159168%	52.73905243%	16.94112117%	31.96576117%	81.51818509%	66.50885914%
##	16.87187	17.87586	16.52042	17.02000	19.44812	18.56942
##	13.57316636%	72.02767946%	95.75497692%	85.96036911%	41.74553351%	75.75423541%
##	15.96876	18.70529	20.15365	19.74852	17.39411	18.90000
##	12.89697641%	84.40036611%	48.22059954%	73.86768851%	73.69824101%	5.75277749%
##	15.83934	19.54057	17.59071	18.87071	18.85547	15.23452
##	25.82413440%	61.11394193%	95.29512636%	84.01542867%	46.99528466%	96.75052785%
##	16.90066	18.28360	20.12371	19.48926	17.52234	20.21846
##	46.84289142%	23.36761556%	84.24755870%	43.43918408%	24.22170751%	1.71236598%
##	17.51383	16.87732	19.52020	17.40932	16.88526	14.55308
##	85.24440827%	58.27056135%	44.43893421%	97.71033176%	97.17268494%	43.72285171%
##	19.65308	18.01916	17.41552	20.99774	20.55107	17.41108
##	18.71776013%	76.24595363%	57.17659069%	66.35393803%	63.68160241%	84.55621938%
##	16.65260	18.90000	17.99449	18.56558	18.46309	19.56134
##	60.06125291%	65.00372610%	74.43937077%	6.23836399%	32.96195886%	42.14920430%
##	18.18570	18.53209	18.90000	15.35645	17.02655	17.40133
##	12.64910898%	1.81188858%	32.50640330%	85.90682147%	27.74250342%	22.88595950%
##	15.81322	14.55617	17.02231	19.74138	16.97202	16.87284
##	64.14323167%	16.91371752%	24.95216082%	28.92098962%	9.79609087%	36.06261581%
##	18.49457	16.51838	16.89206	17.01586	15.51251	17.09485
##	75.45577143%	39.29699399%	66.22044703%	1.77657036%	95.96122208%	79.98571936%
##	18.90000	17.31821	18.56227	14.55507	20.16708	19.32961
##	66.72027924%	41.24587236%	60.62650210%	9.99697151%	18.15322030%	16.98545185%
##	18.57466	17.37862	18.23826	15.53368	16.61060	16.52372
##	41.12361702%	84.56324122%	47.37622682%	75.14393383%	65.30329937%	15.62882415%
##	17.37483	19.56228	17.54359	18.90000	18.53952	16.36386
##	15.09863986%	44.41795330%	31.24085737%	12.40868827%	80.10751610%	77.32013552%
##	16.26196	17.41539	17.02000	15.78788	19.35000	18.90000
##	50.06167986%	67.83466404%	48.18816858%	84.23408743%	72.55344826%	53.59719626%
##	17.71421	18.60029	17.58890	19.51840	18.75255	17.91842
##	16.33490126%	50.49642548%	51.64629957%	59.75686992%	43.05594647%	92.04971453%
##	16.47532	17.74386	17.82166	18.15739	17.40695	20.00535
##	85.31953988%	31.67108058%	34.41605130%	96.39332630%	2.97899190%	24.32897761%
##	19.66309	17.02000	17.04007	20.19521	14.59235	16.88626
##	25.48749412%	24.09915086%	94.47427723%	96.59550174%	81.15423890%	99.85084964%
##	16.89703	16.88412	20.07028	20.20837	19.44473	22.77609
##	97.60253797%	3.14412797%	79.93268368%	55.07816458%	59.18669226%	47.88220010%
##	20.90819	14.59747	19.32073	17.98148	18.10436	17.57183
##	24.91296099%	46.38534256%	35.99187839%	45.90941428%	98.89915055%	64.06740695%
##	16.89169	17.48830	17.08937	17.46175	21.98541	18.48940
##	70.36883677%	50.94557947%	22.72563661%	16.62752677%	41.22941352%	74.28119148%
##	18.60814	17.77449	16.87135	16.49709	17.37811	18.90000
##	87.53741237%	88.58820109%	67.99261391%	74.94103750%	63.08594604%	24.34056166%
##	19.91366	19.94623	18.60078	18.90000	18.42246	16.88637
##	80.14161089%	85.96292431%	92.31216845%	46.04137866%	7.87999530%	91.09166777%
##	19.35571	19.74886	20.00617	17.46911	15.44985	20.00238
##	9.49554904%	28.32591338%	79.40266200%	54.38385801%	31.61785780%	44.82378322%
##	15.49493	16.99372	19.23201	17.95744	17.02000	17.41791
##	4.26726425%	30.74625696%	85.35248607%	38.83596610%	28.42529260%	30.33586959%
##	14.86151	17.02000	19.66749	17.30391	16.99742	17.02000
##	62.00674351%	86.75340994%	20.40956866%	26.56454812%	87.85976169%	31.21055092%
##	18.34886	19.85423	16.75558	16.92820	19.92365	17.02000
##	67.31209850%	57.36703717%	21.98351745%	71.48725458%	13.06888685%	18.91626215%
##	18.58934	17.99568	16.83853	18.65670	15.87184	16.66737

##	80.24957601%	41.20583872%	22.39940511%	93.57218065%	98.73882155%	45.31057237%
##	19.37378	17.37738	16.86045	20.01155	21.85221	17.42833
##	12.81808245%	18.01249799%	44.47380307%	58.58124807%	63.32538957%	63.98238875%
##	15.83103	16.60013	17.41574	18.04806	18.43879	18.48360
##	12.16778138%	29.05457439%	12.31505962%	40.91488745%	99.67018734%	76.10014169%
##	15.76248	17.02000	15.77801	17.36836	22.62599	18.90000
##	37.00816529%	24.56743084%	67.42958466%	26.44265243%	83.73999661%	8.90877424%
##	17.16813	16.88848	18.59225	16.92367	19.46878	15.47855
##	46.72322555%	66.06355556%	53.30952778%	91.19596991%	51.07847746%	10.50131891%
##	17.50716	18.55838	17.90415	20.00271	17.78355	15.58684
##	4.79479257%	54.75345890%	71.30119833%	74.11295131%	83.15522433%	49.14441174%
##	14.99397	17.97577	18.63998	18.89275	19.46334	17.65165
##	76.08220354%	80.09538068%	77.08752465%	17.06873046%	14.92410402%	85.47089314%
##	18.90000	19.34797	18.90000	16.52991	16.22841	19.68327
##	24.67310040%	39.86026088%	63.77517125%	82.30975408%	83.46085041%	2.02628975%
##	16.88946	17.33567	18.46947	19.45548	19.46619	14.56281
##	2.96028783%	53.83577924%	19.33191991%	20.53826631%	38.83300663%	3.57876588%
##	14.59177	17.93025	16.69829	16.76237	17.30382	14.68863
##	65.28212179%	80.93495762%	7.22998099%	77.01999543%	39.99942837%	79.26487084%
##	18.53900	19.44270	15.43172	18.90000	17.33998	19.20894
##	53.40622498%	37.87850908%	63.99103454%	16.44384186%	80.80983381%	28.09660786%
##	17.90895	17.23558	18.48419	16.48342	19.44153	16.98519
##	87.32806256%	93.31556375%	65.04365678%	11.40901975%	86.95976383%	65.43348215%
##	19.90717	20.00928	18.53308	15.68251	19.88174	18.54275
##	95.39866608%	49.40109081%	98.39737427%	4.04826868%	56.32452504%	5.57796340%
##	20.13045	17.66915	21.56854	14.80652	17.98921	15.19063
##	21.52089642%	74.90079550%	43.62576238%	70.41066410%	35.86152999%	39.38994452%
##	16.81415	18.90000	17.41048	18.60827	17.07927	17.32109
##	31.86560806%	15.69787287%	91.38069924%	72.28258590%	46.89795701%	85.03222708%
##	17.02000	16.37713	20.00328	18.72820	17.51691	19.62480
##	33.02136210%	37.55417708%	3.94666980%	69.79467750%	72.05057438%	64.72545194%
##	17.02710	17.21045	14.78101	18.60636	18.70735	18.52519
##	85.84579923%	27.52170900%	34.16188948%	90.09796020%	65.11145337%	46.34797394%
##	19.73325	16.96381	17.03771	19.99304	18.53476	17.48622
##	44.88396195%	73.40279100%	87.35191254%	54.45923128%	28.59221858%	23.11727786%
##	17.41828	18.82891	19.90791	17.96118	17.00363	16.87499
##	21.28031352%	18.48676102%	35.14927663%	12.66192277%	58.92328313%	61.42355395%
##	16.80147	16.63542	17.04689	15.81457	18.07987	18.30909
##	70.32375119%	46.27673177%	61.57636507%	83.19388423%	20.68892031%	74.24138442%
##	18.60800	17.48224	18.31951	19.46370	16.77031	18.90000
##	47.12421673%	53.25396592%	20.11453786%	74.80059259%	95.31693645%	35.32434523%
##	17.52953	17.90140	16.74004	18.90000	20.12513	17.04852
##	87.33771513%	13.70111536%	33.15433143%	71.66129923%	74.41812137%	18.53249948%
##	19.90747	15.99335	17.02834	18.67235	18.90000	16.63882
##	17.73185916%	86.99647773%	93.63507691%	56.21887785%	25.97823131%	83.69040547%
##	16.57925	19.88663	20.01564	17.98856	16.90639	19.46832
##	22.02233742%	83.39316403%	64.81756696%	27.13593268%	15.06997556%	56.01573207%
##	16.84058	19.46556	18.52748	16.94946	16.25645	17.98730
##	59.87513752%	85.82230527%	7.91459973%	71.09776745%	27.67979351%	58.26617382%
##	18.16839	19.73011	15.45082	18.62169	16.96969	18.01875
##	98.70134722%	69.24041132%	4.92303916%	85.29697452%	11.07348304%	56.63433259%
##	21.82108	18.60465	15.02618	19.66009	15.64715	17.99113
##	37.92338141%	59.32995910%	79.39228364%	81.30657717%	10.98658799%	68.42494446%
##	17.23906	18.11769	19.23027	19.44615	15.63799	18.60212

##	37.41018709%	65.33071480%	27.19118700%	62.65377004%	23.32095359%	80.02774257%
##	17.19929	18.54020	16.95151	18.39299	16.87688	19.33664
##	19.95808426%	25.72483297%	99.51649935%	33.78074255%	36.43465349%	63.07213523%
##	16.73179	16.89924	22.49831	17.03416	17.12369	18.42152
##	44.43080260%	97.41456355%	64.61671668%	72.28058111%	1.73730173%	54.55520772%
##	17.41547	20.75202	18.52249	18.72802	14.55386	17.96594
##	61.41715152%	75.15509601%	94.66101383%	22.54937261%	14.52931121%	56.81358271%
##	18.30865	18.90000	20.08243	16.86835	16.15253	17.99224
##	70.93339919%	40.63691117%	91.75832681%	72.55729849%	78.08536687%	46.20462921%
##	18.60989	17.35974	20.00445	18.75290	19.01149	17.47822
##	22.50396991%	76.93846370%	66.98903928%	22.16803334%	29.07169119%	75.58811470%
##	16.86596	18.90000	18.58133	16.84826	17.02000	18.90000
##	73.63500544%	2.14465510%	39.42447419%	73.48798802%	90.55221451%	80.10338463%
##	18.84979	14.56648	17.32216	18.83657	20.00071	19.34931
##	3.41605686%	5.69554651%	82.49853631%	54.33428572%	69.03089781%	94.94171189%
##	14.64777	15.22015	19.45724	17.95498	18.60400	20.10071
##	0.06568639%	72.88971250%	77.78107417%	83.54128832%	7.20302784%	43.21466640%
##	14.50204	18.78279	18.96055	19.46693	15.43096	17.40793
##	3.64648220%	20.12152942%	70.23718113%	50.54231337%	45.92351597%	11.44708442%
##	14.70563	16.74040	18.60774	17.74699	17.46253	15.68652
##	54.81209792%	94.53195620%	97.71798649%	91.69078285%	73.80243479%	55.56996779%
##	17.97868	20.07403	21.00410	20.00424	18.86484	17.98453
##	97.57585758%	8.70030054%	31.77857948%	11.53389849%	38.38203405%	53.95998186%
##	20.88602	15.47274	17.02000	15.69567	17.27461	17.93642
##	25.71577791%	8.40294303%	49.98907296%	93.10448200%	28.68647466%	49.54541053%
##	16.89916	15.46444	17.70925	20.00862	17.00714	17.67900
##	20.35663442%	64.99704269%	5.45324318%	65.19135209%	85.10193222%	73.31093068%
##	16.75279	18.53193	15.15931	18.53675	19.63409	18.82065
##	8.17208826%	15.19526031%	91.54207751%	37.29624164%	31.14801680%	87.10475930%
##	15.45800	16.28053	20.00378	17.19046	17.02000	19.90025
##	45.04128115%	63.69198360%	54.01141618%	45.99868061%	5.10073281%	97.10975897%
##	17.41926	18.46379	17.93897	17.46673	15.07079	20.49879
##	6.05379744%	90.78179481%	76.71940716%	24.72775439%	69.98805224%	56.39949203%
##	15.31011	20.00142	18.90000	16.88997	18.60696	17.98968
##	47.78090077%	53.10578591%	77.94339205%	25.25776459%	47.38254880%	44.61515930%
##	17.56617	17.89405	18.98772	16.89490	17.54395	17.41661
##	86.15671732%	31.45588804%	35.46103423%	24.78235832%	54.04915076%	33.35702941%
##	19.77469	17.02000	17.04979	16.89048	17.94084	17.03022
##	62.74363173%	18.14192550%	75.55908652%	85.25939037%	84.09904470%	89.40786717%
##	18.39912	16.60976	18.90000	19.65508	19.50040	19.97164
##	13.47049223%	60.51653360%	23.12048471%	41.96974782%	85.32504595%	87.69730539%
##	15.94903	18.22804	16.87502	17.40021	19.66383	19.91862
##	31.87065898%	7.27591109%	76.11394958%	34.80541727%	58.23198906%	55.26489369%
##	17.02000	15.43300	18.90000	17.04369	18.01557	17.98264
##	39.07513544%	35.03422923%	11.03386143%	46.58671003%	19.76789541%	57.03613772%
##	17.31133	17.04582	15.64297	17.49954	16.72177	17.99362
##	82.14986406%	94.39426269%	79.65792359%	54.57027170%		
##	19.45399	20.06507	19.27474	17.96669		

(d)(iii)

Before answering the question, let us first analyse the cylinders data from mtcars dataset.

```
# Printing all 32 rows of data with respect to number of cylinders  
print(mtcars$cyl)
```

```
## [1] 6 6 4 6 8 6 8 4 4 6 6 8 8 8 8 8 4 4 4 4 8 8 8 8 4 4 4 8 6 8 4
```

```
# Running the summary to get more info  
summary(mtcars$cyl)
```

```
##      Min. 1st Qu.  Median    Mean 3rd Qu.    Max.  
##      4.000   4.000   6.000   6.188   8.000   8.000
```

```
#calculating the mode of cyl  
y <- mtcars$cyl  
y <- table(y)  
names(y)[which.max(y==max(y))]
```

```
## [1] "8"
```

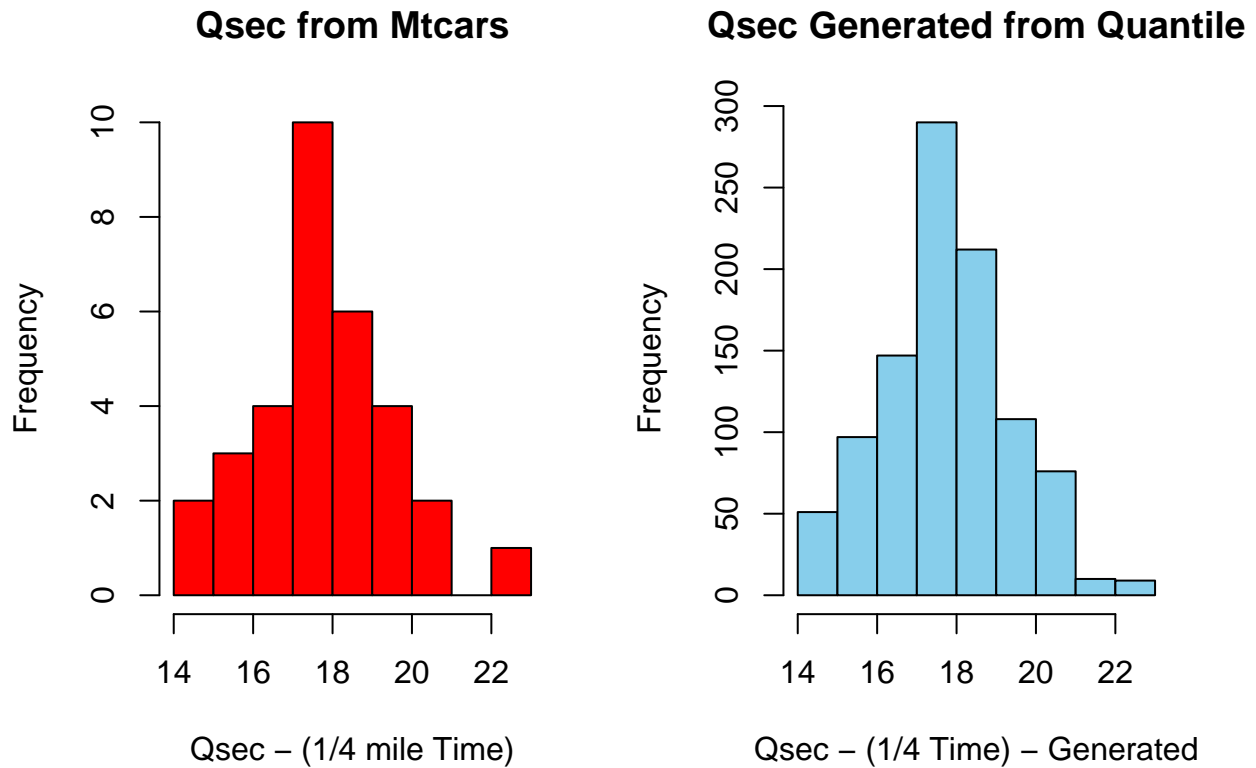
```
## Using the quantile function for cyl  
quantile(mtcars$cyl, prob=runif(10))
```

```
## 38.4802% 78.30924% 36.98379% 39.03124% 34.94357% 5.158628% 87.79199% 7.831603%  
## 6.000000 8.000000 6.000000 6.000000 5.665013 4.000000 8.000000 4.000000  
## 3.97302% 80.8397%  
## 4.000000 8.000000
```

Yes, it does work for cyl. From the actual data cyl has only 3 varieties (namely 4, 6 and 8 ). So when we compute quantile, we will get these 3 values irrespective of the probabilities we choose.

d(iv)

```
dat <- par(mfrow = c(1,2))  
hist(mtcars$qsec, main = "Qsec from Mtcars", xlab = "Qsec - (1/4 mile Time)", col = 'red')  
hist(samobs_fromqsec, main = "Qsec Generated from Quantile", xlab = "Qsec - (1/4 Time) - Generated", col = 'blue')
```

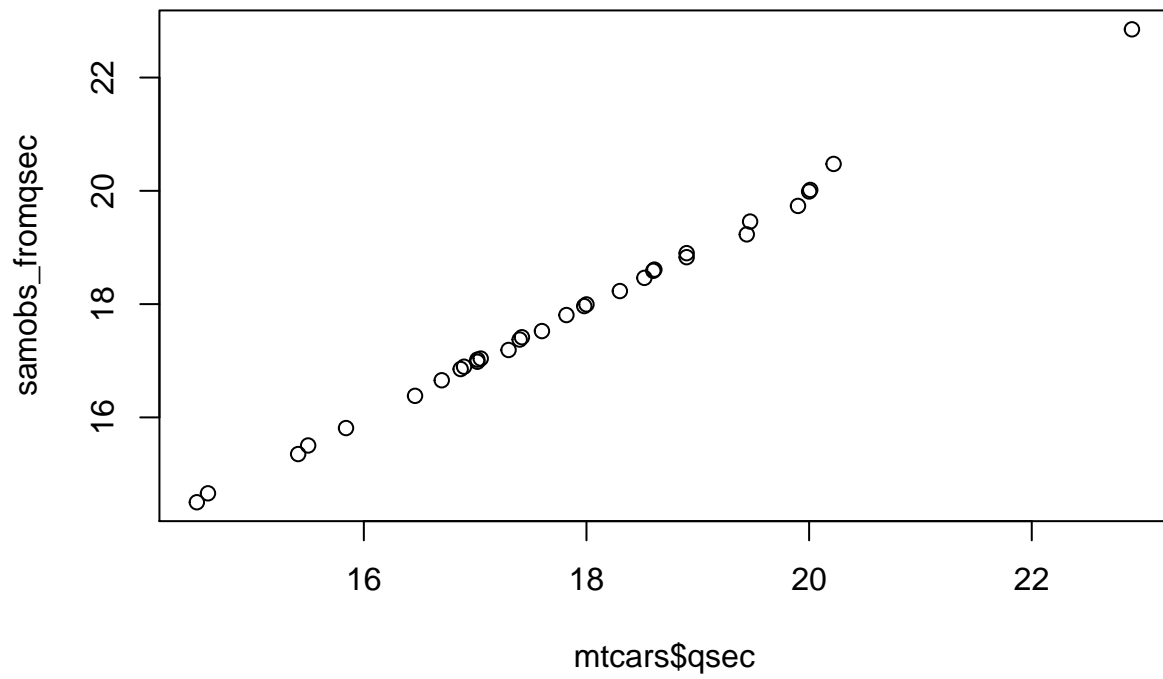


```
par(dat)
```

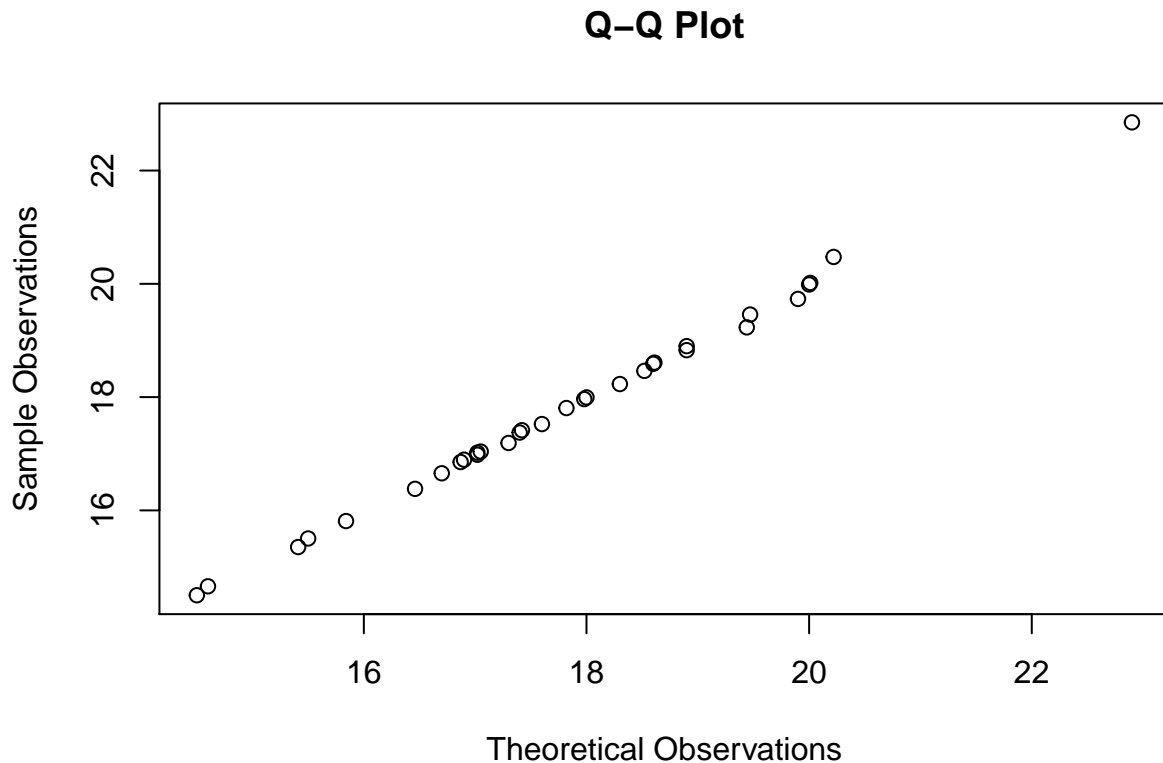
From the histograms we can infer that both the graphs follow the same distribution. Although in the left (Original Qsec data from mtcars) we can see that there is no bin at 21-22 seconds - That is, there was no car that took 21-22 seconds to attain 1/4 a mile. On the contrary on the right (Qsec generated using the quantile function) we can see that there is a bin at 21-22 since the data was generated with prob 0 and 1. On the Y-Axis (Left side Graph) we can see that the frequency goes from 0 to 10 with a step size of 2. Whereas on the Y-axis (Right side graph) we can see that the frequency goes from 0 to 250 with a step size of 50.

```
d(v)
```

```
qqplot(mtcars$qsec, samobs_fromqsec)
```



```
qqplot(mtcars$qsec, samobs_fromqsec, main = "Q-Q Plot", xlab = "Theoretical Observations", ylab = "Sample Observations")
```



Generally QQ plots a scatterplot using two quantiles against each other. From the graph we can infer that the points form a line that's roughly straight. This states that both the quantiles come from the same distribution.

d(vi)

Step 1 - Inorder to produce a bootstrap distribution for some estimator theta, Let us use quantile to produce some data. I am using cylinders from mtcars dataset. I used the quantile fucntion with number of bootstrap samples drawn as 100 with a two sided confidence interval(95%) argument.

```
quantile(mtcars$cyl, probs = runif(50), CI.type = "two.sided", nbboot=100)
```

```
## 56.00587%  4.199123%  76.1902%  83.41437%  62.53828%  83.00134%  39.83738%
##  6.723641  4.000000  8.000000  8.000000  8.000000  8.000000  6.000000
## 56.17437% 12.53718% 24.38958% 16.61068% 16.29795% 30.58102% 69.53286%
##  6.828107  4.000000  4.000000  4.000000  4.000000  4.000000  8.000000
## 49.74612% 64.31155% 12.35054% 65.31711% 91.66203% 28.39503% 79.67062%
##  6.000000  8.000000  4.000000  8.000000  8.000000  4.000000  8.000000
## 49.13785% 10.84139% 24.14521% 9.745784% 0.6232897% 61.11649% 13.30628%
##  6.000000  4.000000  4.000000  4.000000  4.000000  8.000000  4.000000
## 60.64681% 33.4013% 63.99745% 63.43665% 38.81281% 58.90254% 79.40596%
```

```
## 8.000000 4.708803 8.000000 8.000000 6.000000 8.000000 8.000000
## 90.96417% 94.81757% 27.15874% 23.83174% 20.02258% 71.44294% 88.76813%
## 8.000000 8.000000 4.000000 4.000000 4.000000 8.000000 8.000000
## 8.288973% 36.00035% 11.23074% 41.27613% 64.51539% 56.33739% 12.87077%
## 4.000000 6.000000 4.000000 6.000000 8.000000 6.929183 4.000000
## 1.997268%
## 4.000000
```

I would recommend using quantiles over bootsrtap. The main reason behind is that quantile function is computationally less expensive when compared to bootstrapping.