

Uebung 6

Computational Statistics
Sommersemester 2019
April 15, 2019
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Name:

```
set.seed(42)
```

A 1

a)

```
gv <- runif(10000,min=0,max=1)
lambda <- 3
invexp <- -(1/lambda)*log(gv)
exp <- rexp(10000,lambda)

invpois = NULL
for (i in 1:10000){
  j <- 0
  U <- runif(1,0,1)
  Y <- -(1/lambda)*log(U)
  sum <- Y

  while(sum < 1){
    U <- runif(1,min = 0,max = 1)
    Y <- -(1/lambda)*log(U)
    sum <- sum + Y
    j <- j + 1;
  }

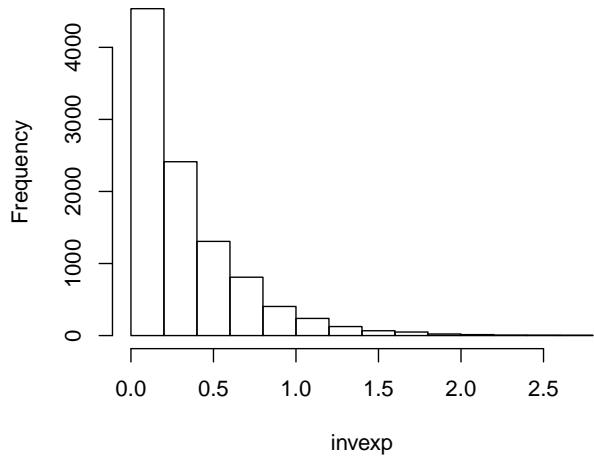
  invpois <- rbind(invpois,j)
}

pois <- rpois(10000,lambda)

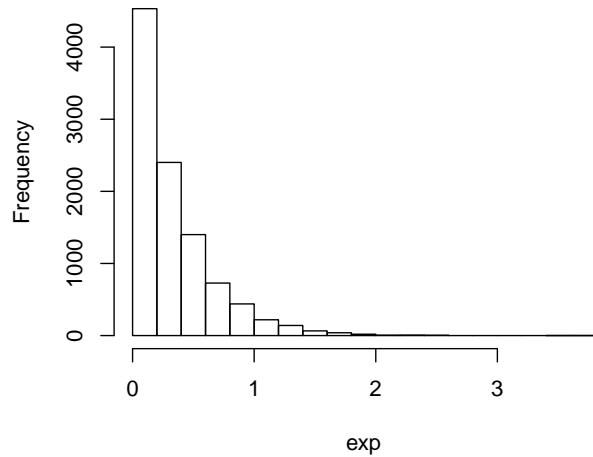
par(mfrow=c(1,2))

hist(invexp)
hist(exp)
```

Histogram of invexp

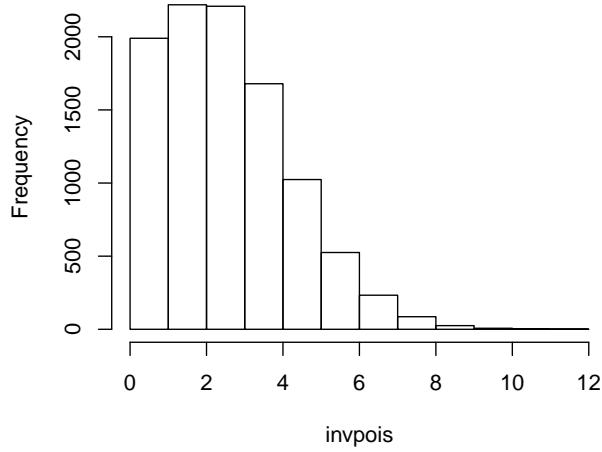


Histogram of exp

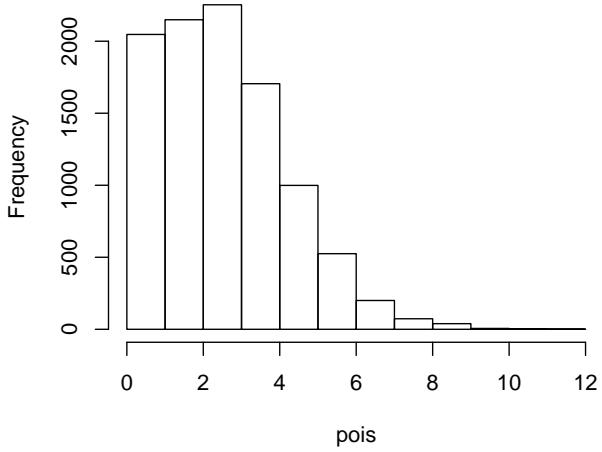


```
hist(invpois)  
hist(pois)
```

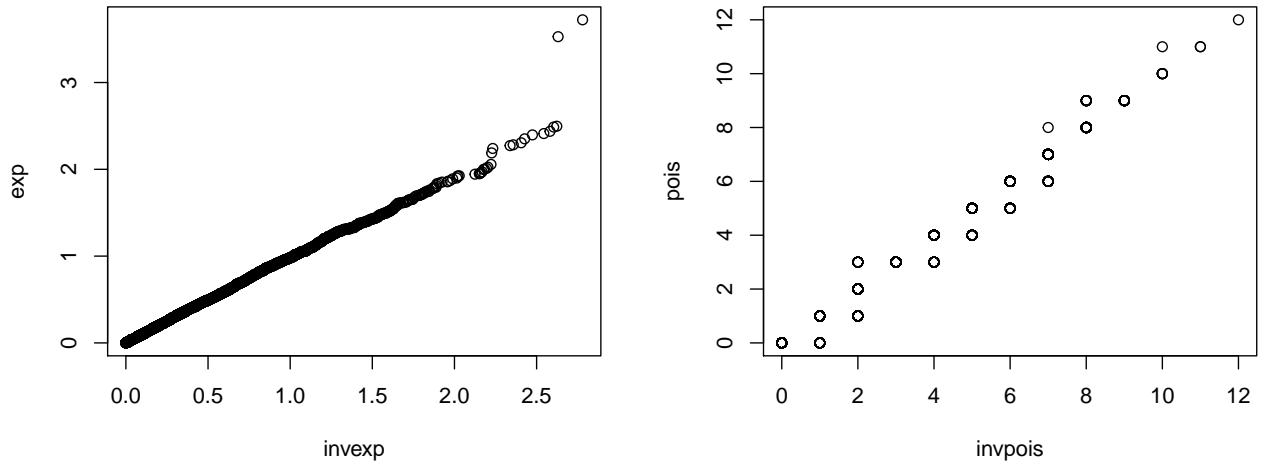
Histogram of invpois



Histogram of pois



```
qqplot(invexp,exp)  
qqplot(invpois,pois)
```



Es zeigt sich dass sich die Inversionsverfahren bei $n=10000$ relativ gut den zugrundeliegenden Verteilungen annähern.

b)

```

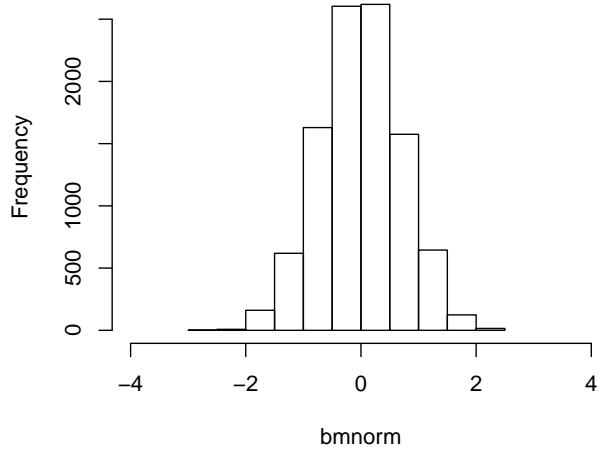
u <- runif(10000,0,1)
v <- runif(10000,0,1)

bmnorm = cos(2*pi*u)*sqrt(-log(v))
norm <- rnorm(10000)

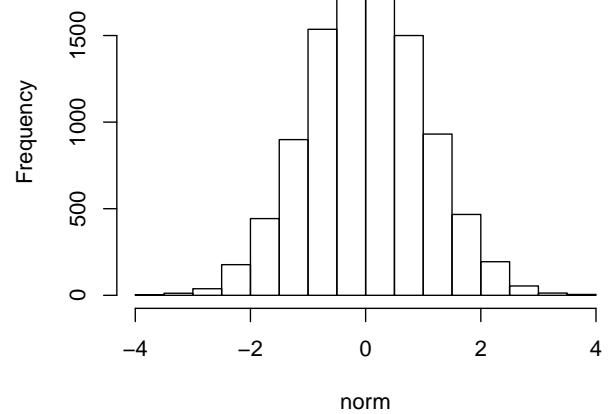
par(mfrow=c(1,2))
hist(bmnorm,xlim=c(-4,4))
hist(norm, xlim=c(-4,4))

```

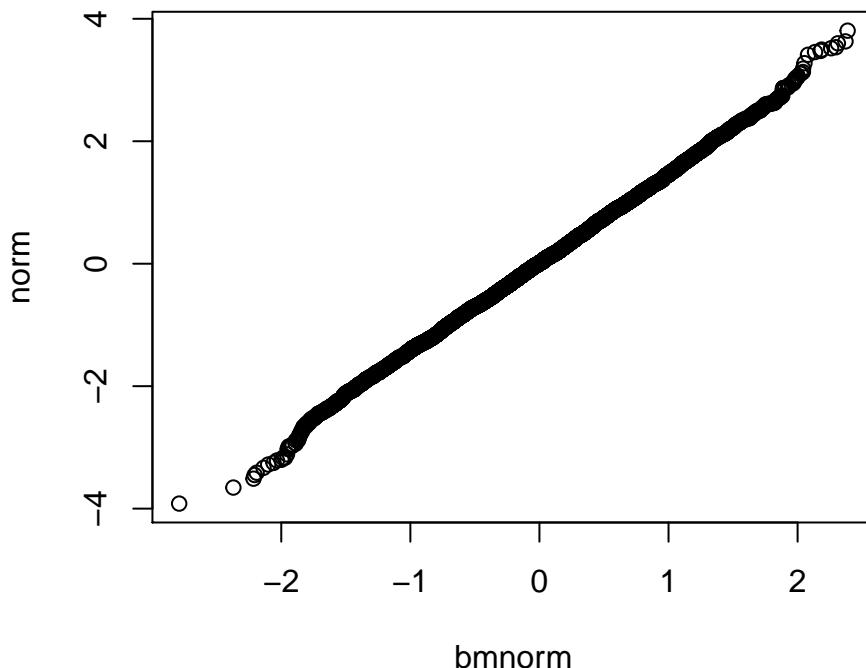
Histogram of bmnorm



Histogram of norm



```
qqplot(bmnorm,norm)
```



Es zeigt sich eine Normalverteilung der transformierten gleichverteilten Zufallszahlen durch die die Transformationsformel. Da die Standardabweichung der durch die Box-Müller Methode erzeugten Zufallszahlen 0.6985516 ist, ist das Histogramm im Vergleich zu den von R erzeugten normalverteilten Zufallszahlen "schmäler". Da rnorm standardmäßig standardnormalverteilte Zufallszahlen erzeugt, sieht man hier deutliche Unterschiede.

A 2

a)

```
randomorg <- read.csv('randomorg.csv', header=FALSE)
randomorg <- randomorg[, "V1"]
```

b)

```
midsqr = function(seed, length) {
  ergebnis = c()
  for(i in 1:length) {
    value = seed * seed
    seed = (value %% 10000) %% 10000000
    ergebnis = c(ergebnis, seed/10000000)
  }
  return(ergebnis)
}
midsqr = midsqr(222222222222, 10000)
```

c)

```
standardize <- function(x){(x-min(x))/(max(x)-min(x))}

baumannikov_zwirbler <- function(seed, n){
  if((seed %% 2) == 0){
    seed <- seed +1
  }
  start <- as.character((seed*n)^2)
  save <- c()
  for (i in 1:(n)){
    if (i == 1){
      x <- start
      k <- nchar(x)
      y <- as.numeric(substr(as.numeric(x), round(0.25*k), round(0.75*k)))
      save[i] <- y
    }
    else{
      x <- as.character((save[i-1]*n)^2)
      k <- nchar(x)
      y <- as.numeric(substr(as.numeric(x), round(0.25*k), round(0.75*k)))
      save[i] <- y
    }
  }
  return(standardize(save%%100))
}

baumannikov_zwirbler <- baumannikov_zwirbler(22222222, 10000)
```

d)

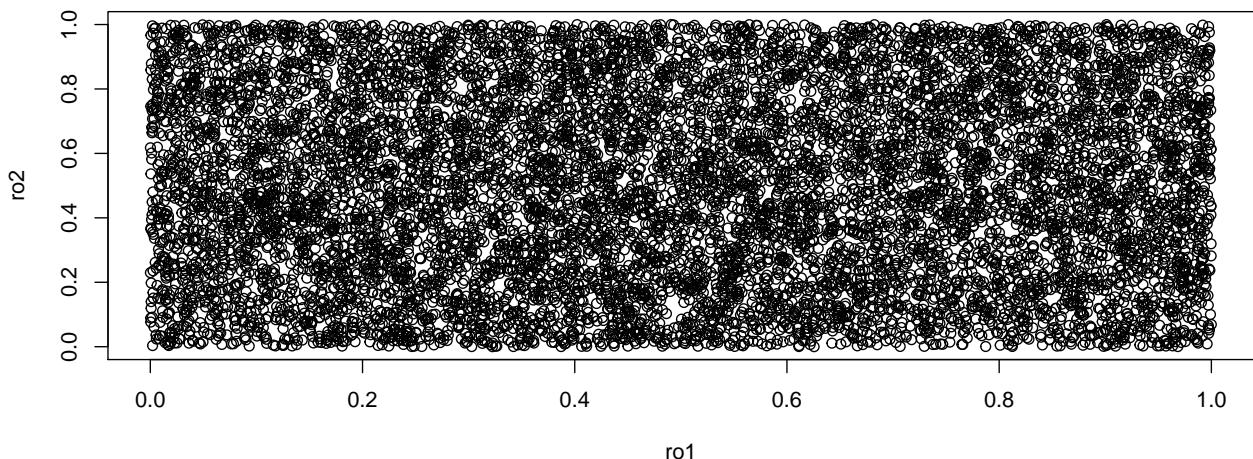
```
set.seed(22222222,kind='Mersenne-Twister')
mersenne <- runif(10000)

set.seed(22222222,kind='Super-Duper')
superduper <- runif(10000)
```

e)

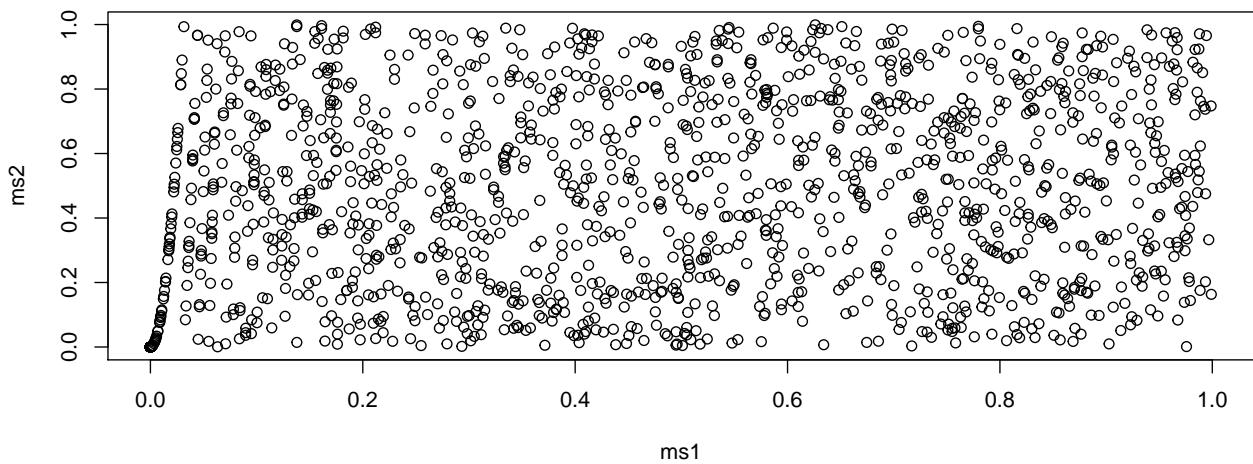
```
ro1 <- randomorg[-length(randomorg)]
ro2 <- randomorg[-1]
plot(ro1,ro2, main="RandomOrg")
```

RandomOrg



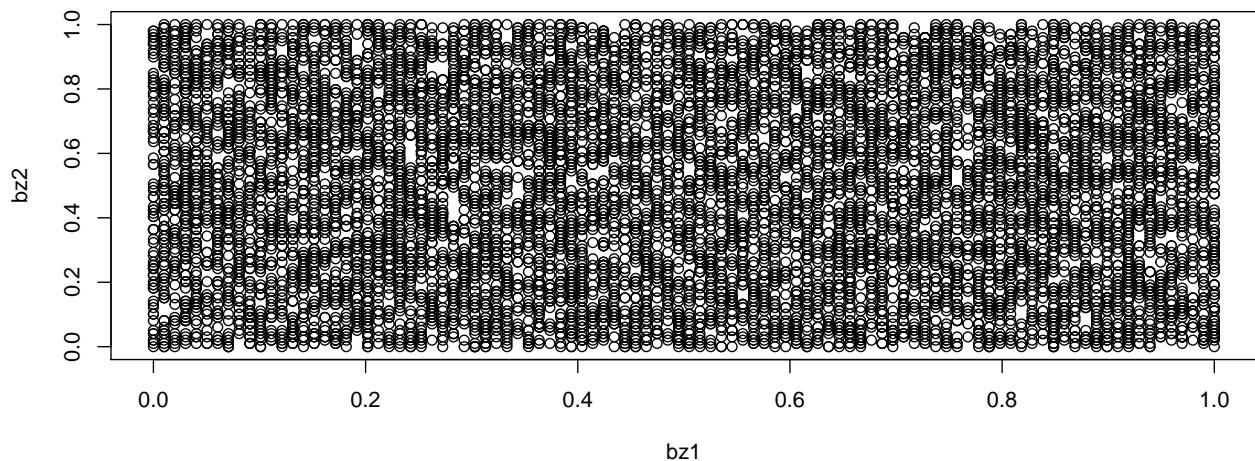
```
ms1 <- midsqr[-length(midsqr)]
ms2 <- midsqr[-1]
plot(ms1,ms2, main="Midsquare")
```

Midsquare



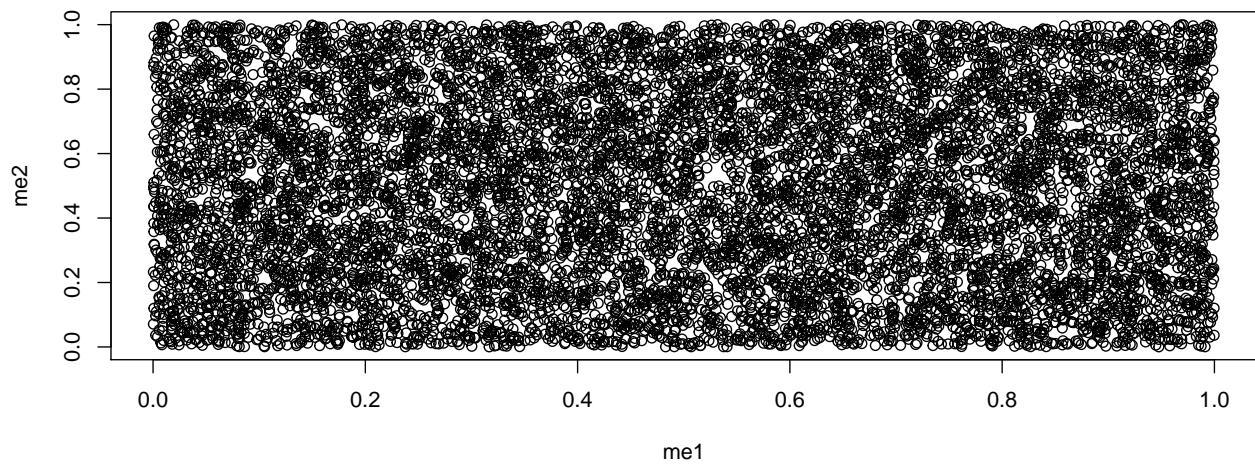
```
bz1 <- baumannikov_zwirbler[-length(baumannikov_zwirbler)]
bz2 <- baumannikov_zwirbler[-1]
plot(bz1, bz2, main="Baumannikov-Zwirbler")
```

Baumannikov–Zwirbler



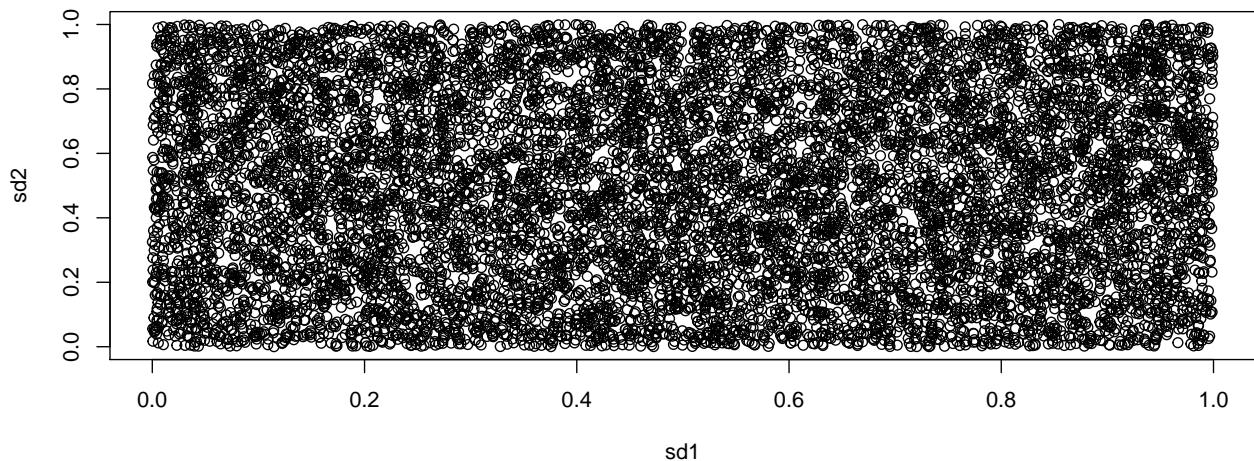
```
me1 <- mersenne[-length(mersenne)]
me2 <- mersenne[-1]
plot(me1,me2, main="Mersenne")
```

Mersenne



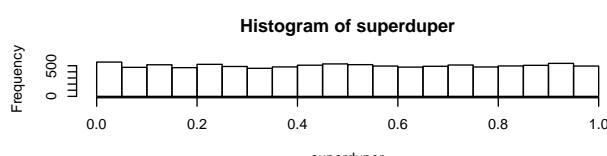
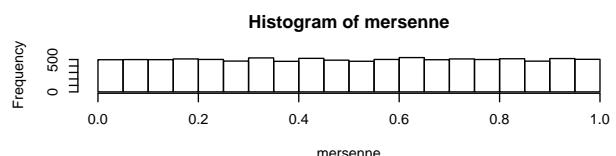
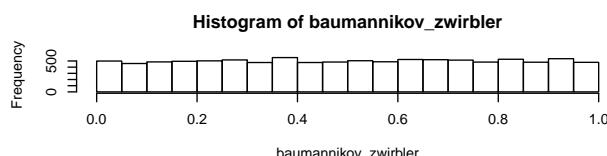
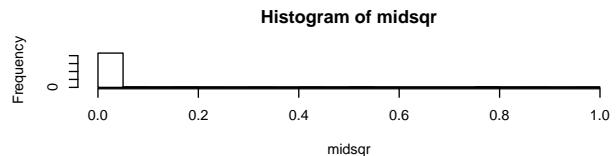
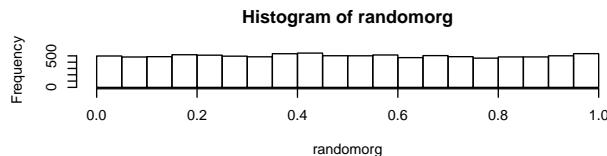
```
sd1 <- superduper[-length(superduper)]
sd2 <- superduper[-1]
plot(sd1,sd2, main="SuperDuper")
```

SuperDuper

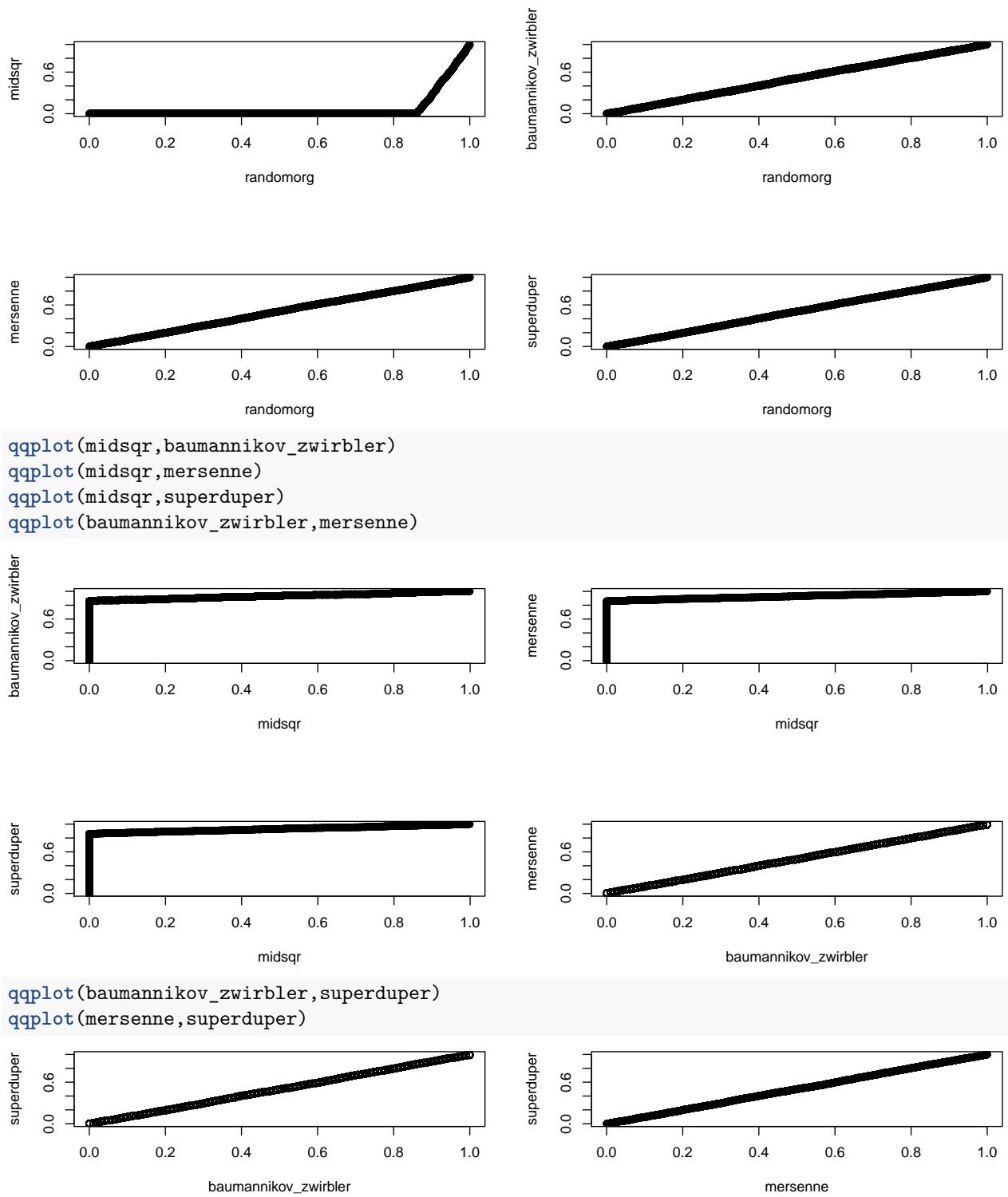


```
par(mfrow=c(3,2))
hist(randomorg)
hist(midsqr)
hist(baumannikov_zwirbler)
hist(mersenne)
hist(superduper)
```

```
par(mfrow=c(2,2))
```



```
qqplot(randomorg,midsqr)
qqplot(randomorg,baumannikov_zwirbler)
qqplot(randomorg,mersenne)
qqplot(randomorg,superduper)
```



Die Scatterplots der ZweierTupel zeigen eine zufällige Verteilung der Zufallszahlen bis auf die Methode des Mittsquadrat Verfahrens von Neumann. Bei diesem zeigt sich eine deutliche Überverteilung der Zahlen bei (0,0). Die Histogramme bestätigen dieses Bild. Auch die QQPlots zeigen die Schwäche des Mittsquadrat Verfahrens von Neumann im Vergleich zu den anderen Zufallsgeneratoren.