

# Einfaches Samplen

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## Einfaches Sampling

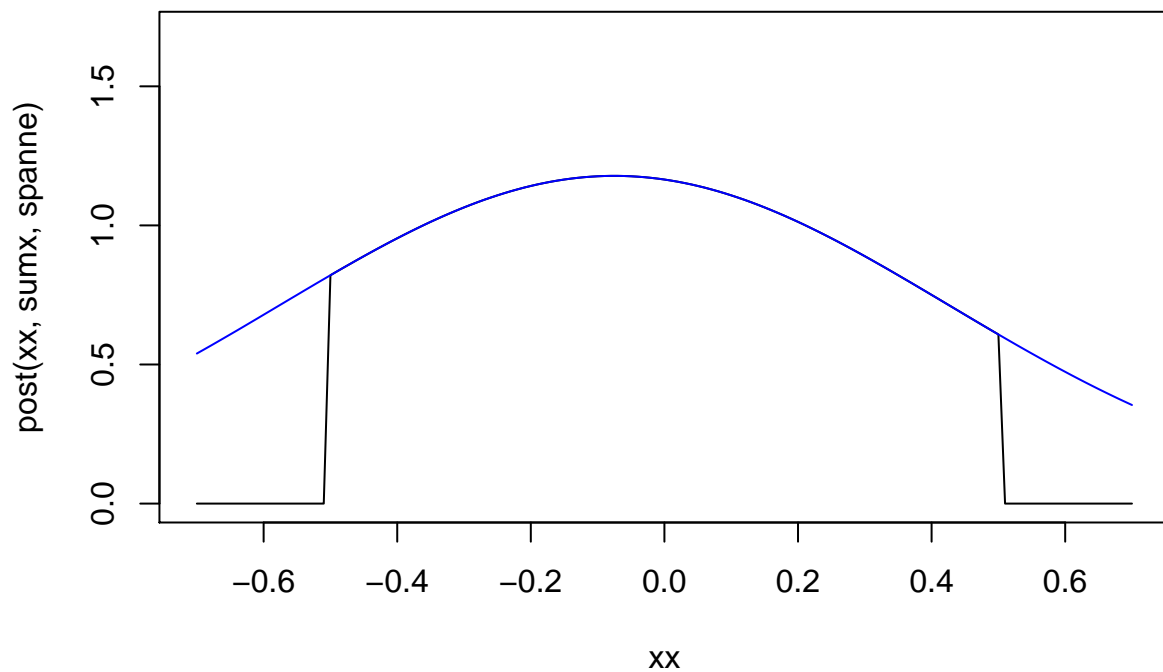
Sei  $x_i \sim N(\mu, \tau^{-1})$  mit  $\tau = 1$  bekannt. Sei  $\mu \sim N_{[-1,1]}(0, 1)$  und  $x = 1, 0.2, -1.5$ . Samplen Sie aus der Posteriori mit Acceptance-Rejection-Methode.

Posteriori:  $\mu|x \sim N_{[-1,1]}(\sum x_i/4, 1/4)$

```
x=c(1,0.2,-1.5)
n=3
sumx <- sum(x)
unten <- pnorm(-.5, sumx/4, sd=1/2)
oben <- pnorm(.5, sumx/4, sd=1/2)
spanne <- oben-unten
xx <- seq(-.7,.7, by=0.01)
post<-function(x,sumx,spanne)
{
  y<-dnorm(x, sumx/4, 1/2)
  y <- y/spanne
  y[x< -.5]<-0
  y[x>.5]<-0
  return(y)
}
plot(xx, post(xx, sumx, spanne), type="l", ylim=c(0,1.7))

factor <- post(-.5, sumx, spanne)/dnorm(-.5,sumx/4,1/2)

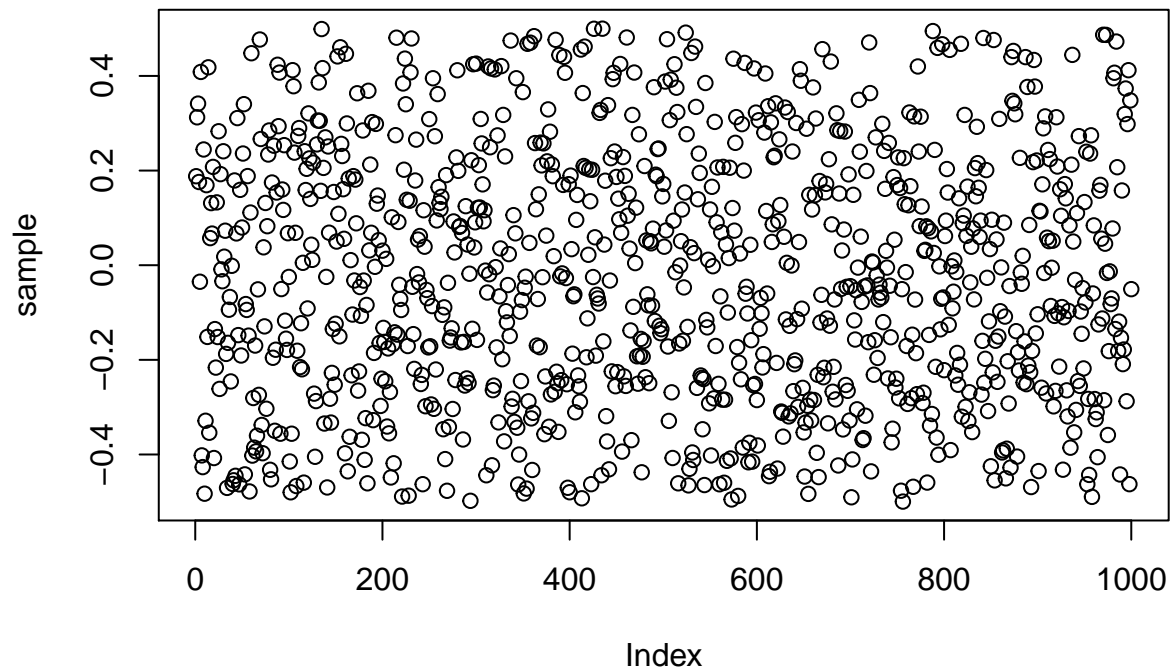
lines(xx, factor*dnorm(xx,sumx/4,1/2), col="blue")
```



```
sample <- c()
N<-1000
while(N!=0)
{
  cat(".",")
  prop <- rnorm(N, sumx/4, 1/2)
  alpha <- post(prop, sumx, spanne)/dnorm(prop, sumx/4, 1/2)/factor
  u <- runif(N)
  sample <- c(sample, prop[alpha>u])
  N <- 1000-length(sample)
}
```

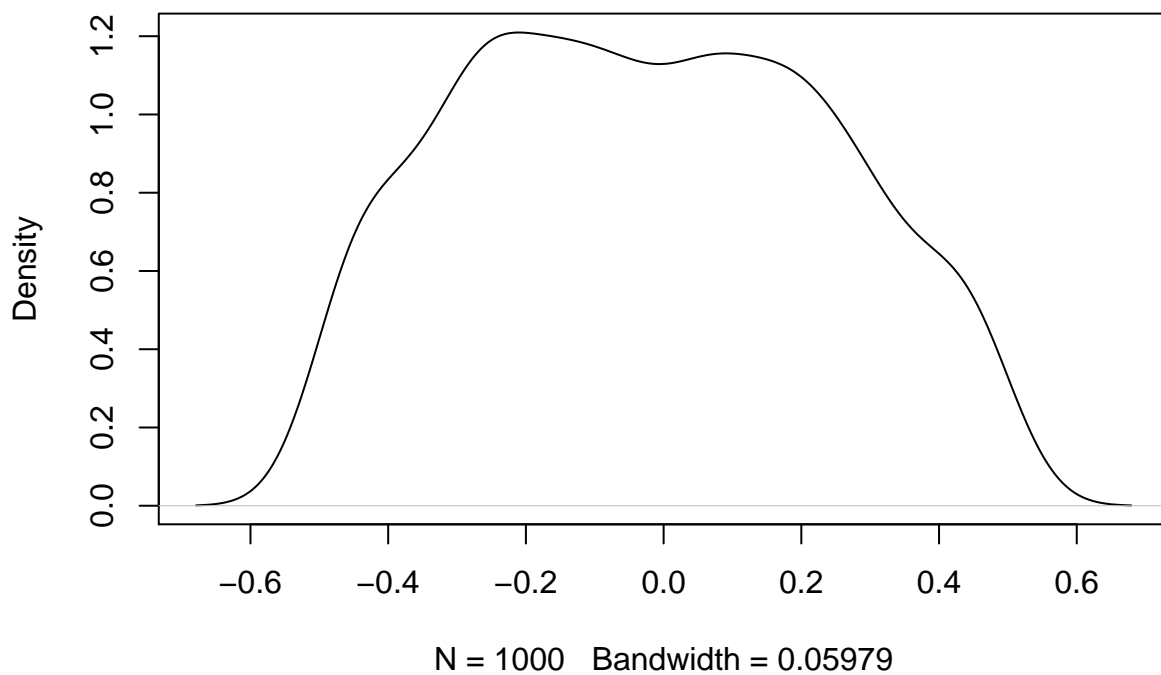
```
## .....
```

```
plot(sample)
```



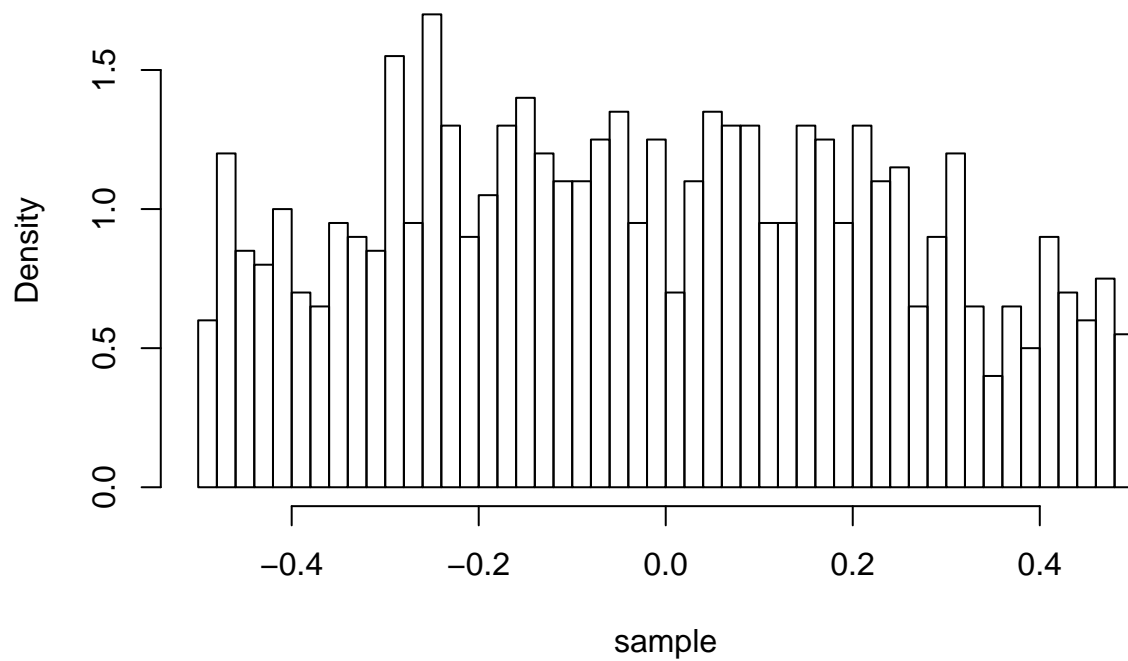
```
plot(density(sample))
```

**density.default(x = sample)**

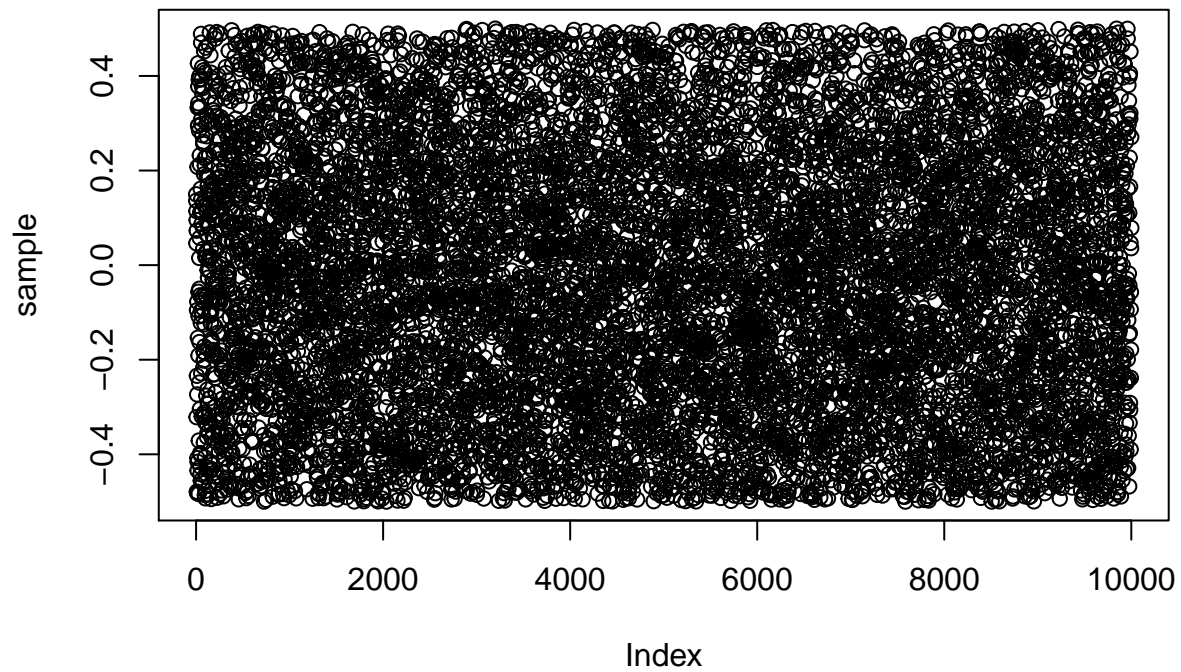


```
hist(sample,breaks = 50, freq=FALSE)
```

## Histogram of sample

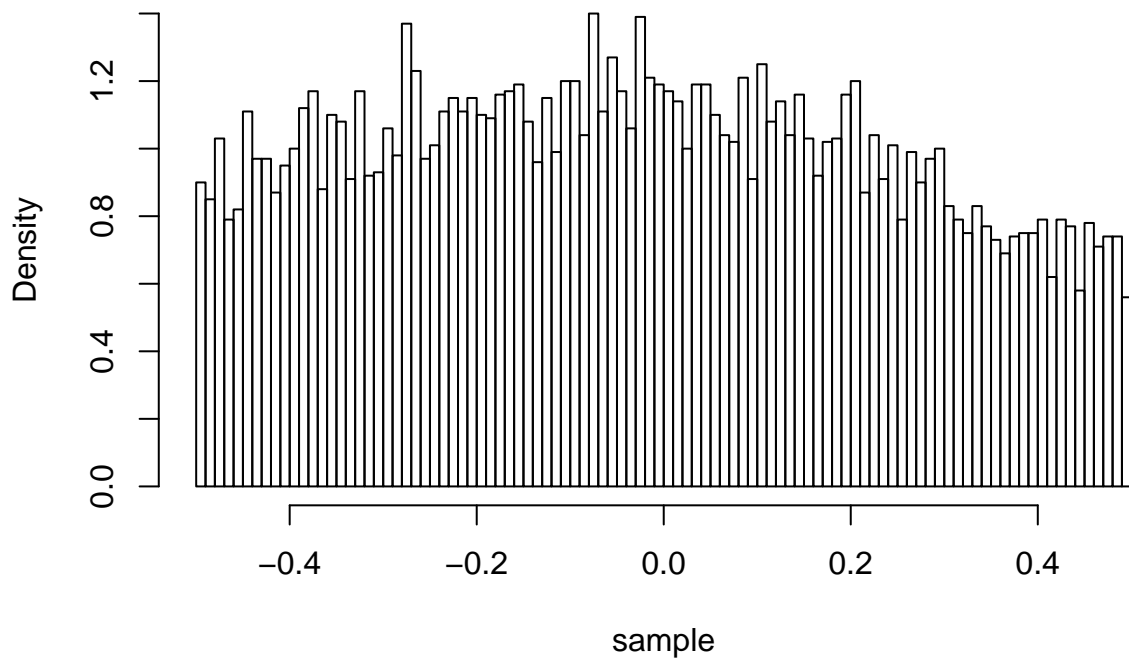


```
draw.from.post<-function(N,mean,sd,unten=-.5,oben=.5)
{
  prop <- rnorm(N, mean, sd)
  while((sum(prop<unten)+sum(prop>oben))!=0)
  {
    prop[prop<unten]<-rnorm(sum(prop<unten),mean,sd)
    prop[prop>oben]<-rnorm(sum(prop>oben),mean,sd)
  }
  return(prop)
}
sample<-draw.from.post(10000,sumx/4,1/2,-.5,.5)
plot(sample)
```



```
hist(sample, breaks = 100, freq=FALSE)
```

**Histogram of sample**



## Gibbs-Sampling

Sei nun  $\tau \sim Ga(1, 1/1000)$ . Ziehe aus der gemeinsamen Posteriori von  $\tau$  und  $\mu$  mittels Gibbs-Sampling. Zuerst mal mit  $\mu \sim N(0, 1)$ .

```

I<-500

s0<-1
a<-1
b<-0.001
n<-length(x)

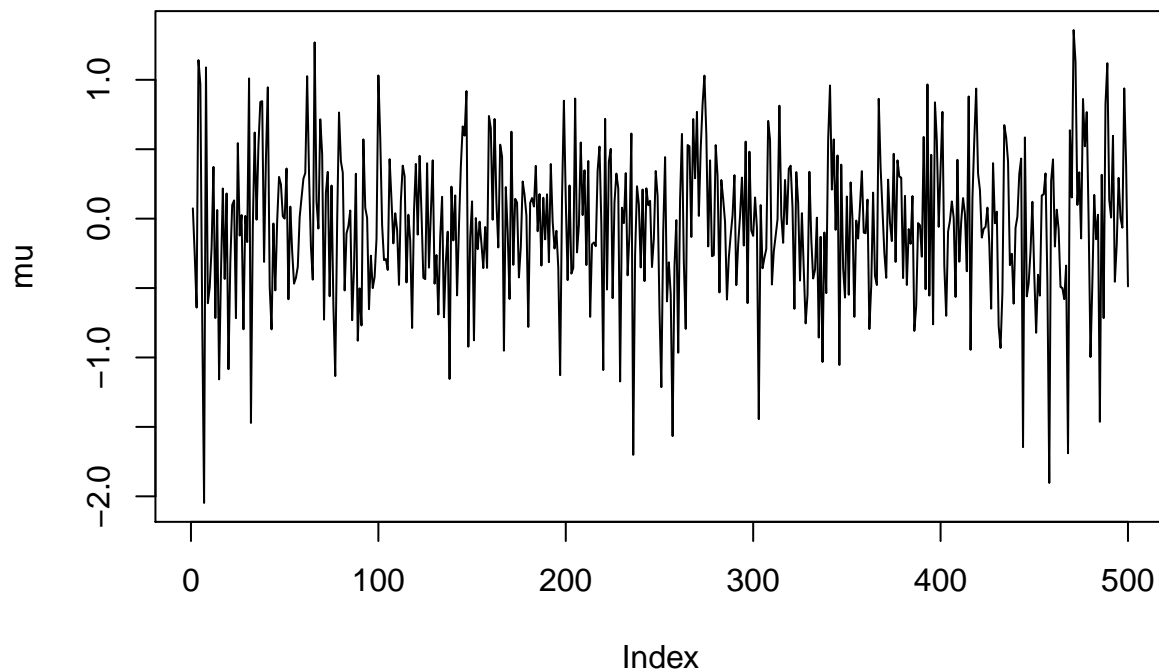
mu<-tau<-rep(NA,I+1)
tau[1]<-1000
mu[1]<-1

for (i in (1:I)+1)
{
  tau[i]<-rgamma(1,a+n/2,b+0.5*sum((x-mu[i-1])^2))
  m<-tau[i]*sumx
  s<-n*tau[i]+1/s0
  mu[i]<-rnorm(1,m/s,sqrt(1/s))
}

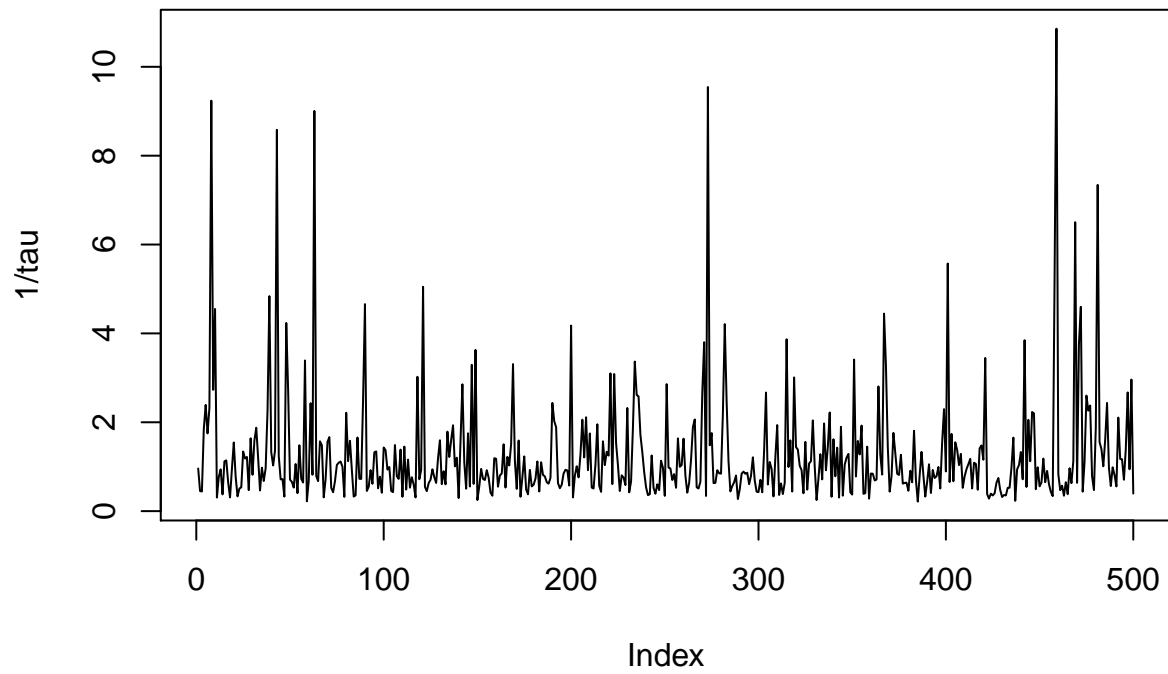
tau<-tau[-1]
mu<-mu[-1]

plot(mu,type="l")

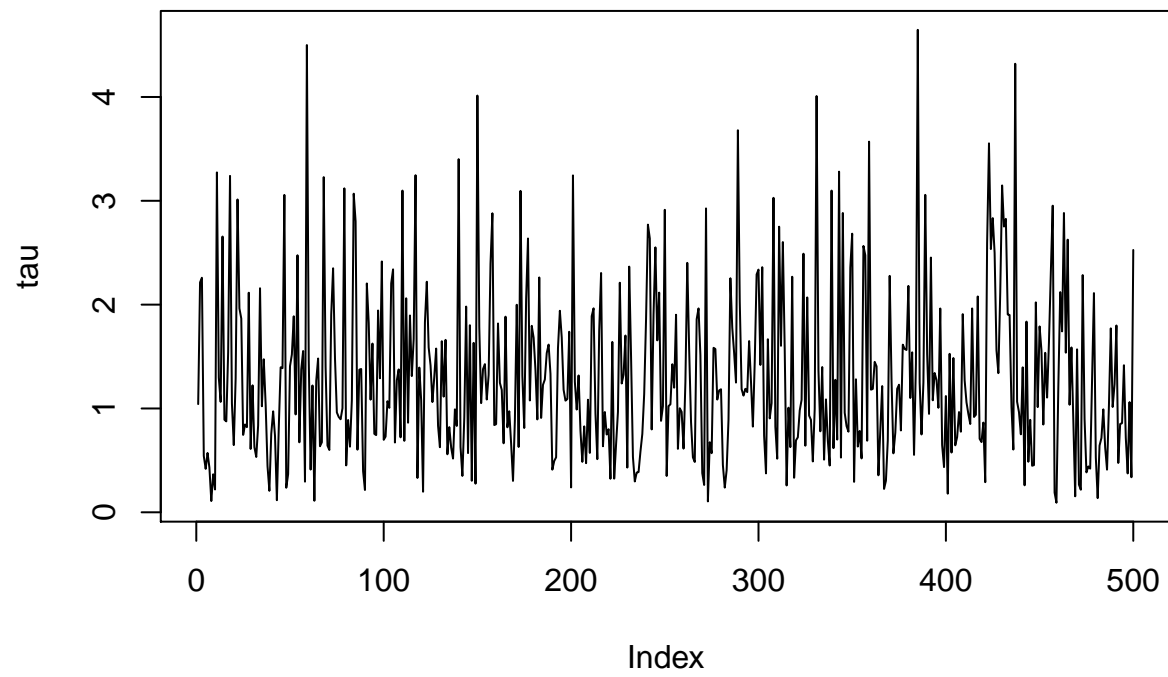
```



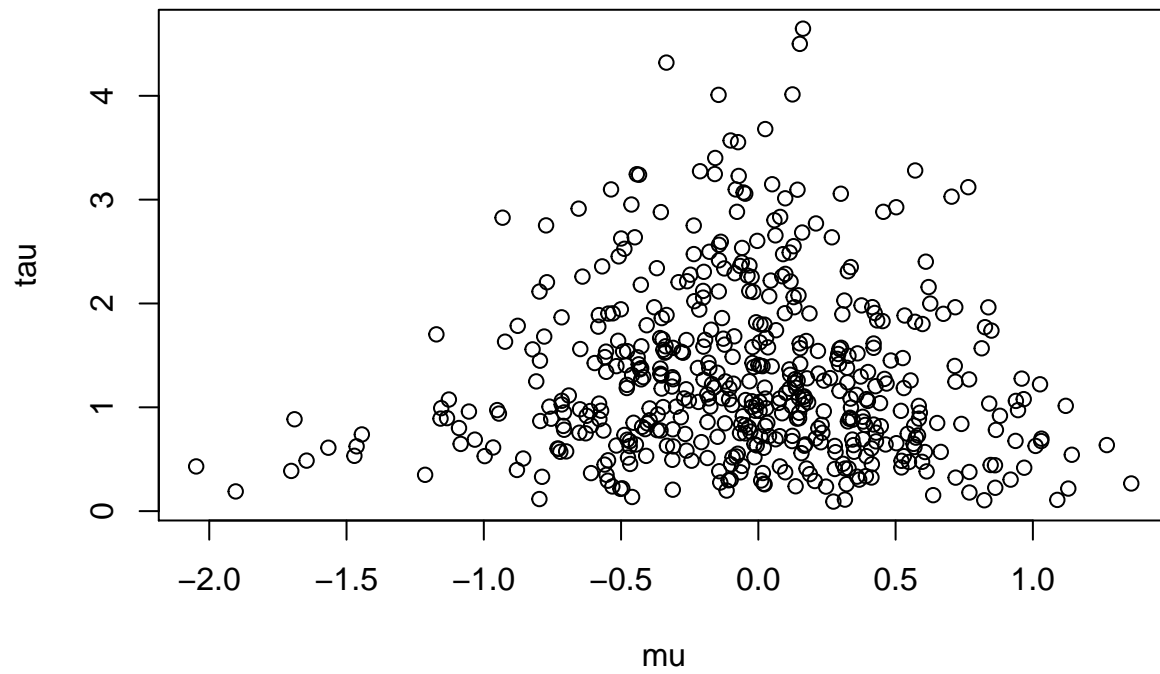
```
plot(1/tau,type="l")
```



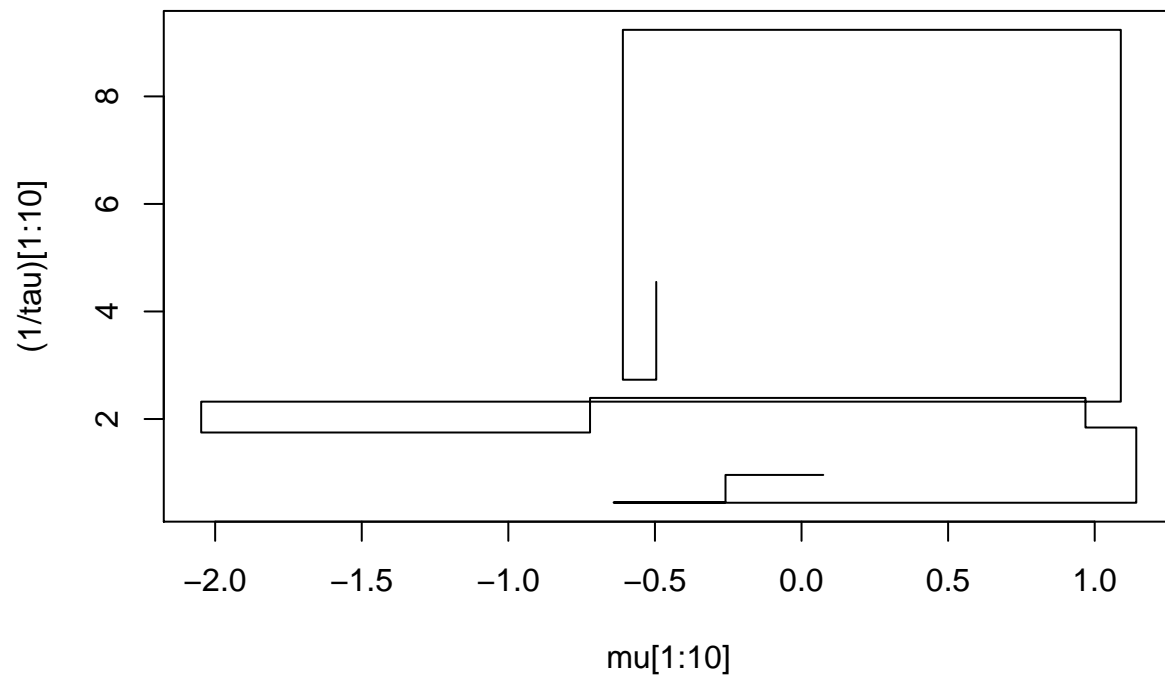
```
plot(tau,type="l")
```



```
plot(mu,tau)
```



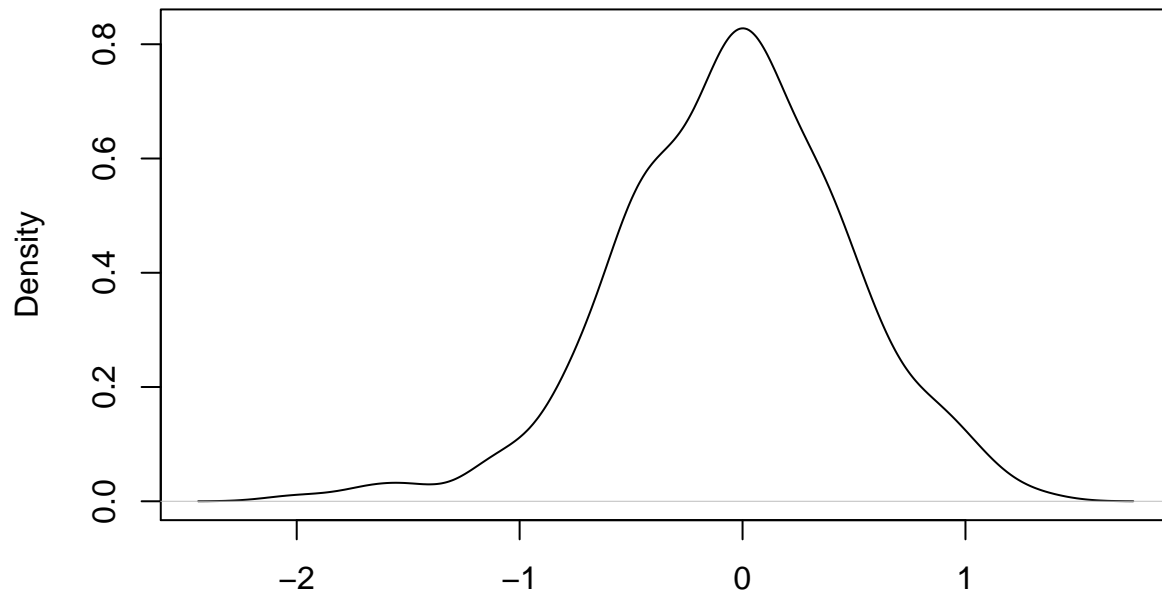
```
plot(mu[1:10], (1/tau)[1:10], type="s")
```



```
plot(density(mu))
```



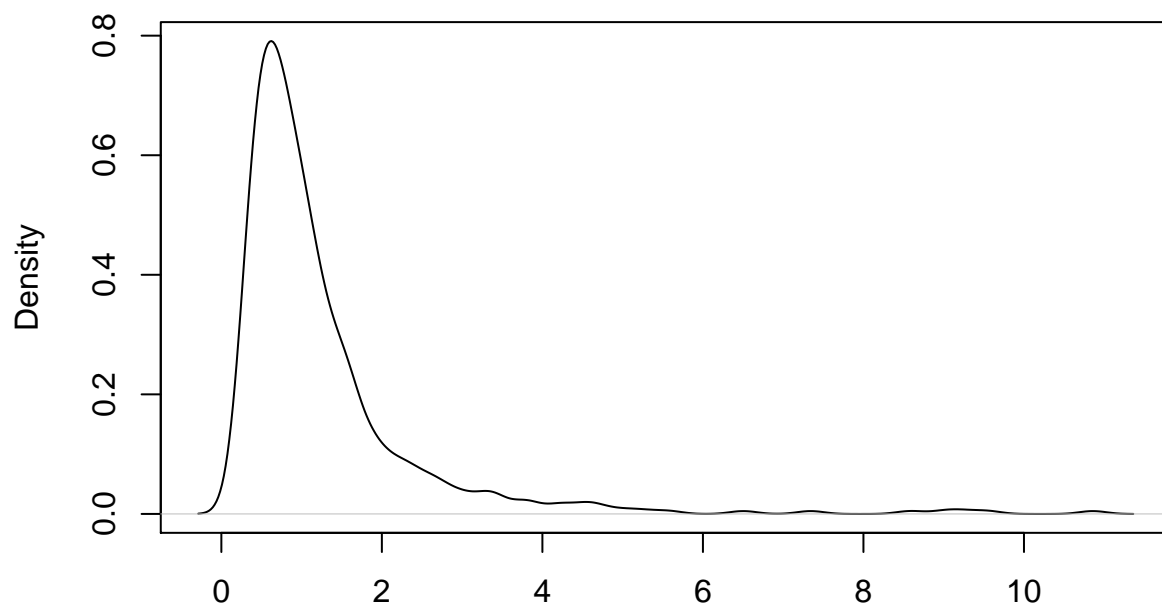
**density.default(x = mu)**



N = 500 Bandwidth = 0.1314

```
plot(density(1/tau))
```

**density.default(x = 1/tau)**



N = 500 Bandwidth = 0.168

Jetzt mit beschränkter Priori:

```
I<-500
```

```

s0<-1
a<-1
b<-0.001
n<-length(x)

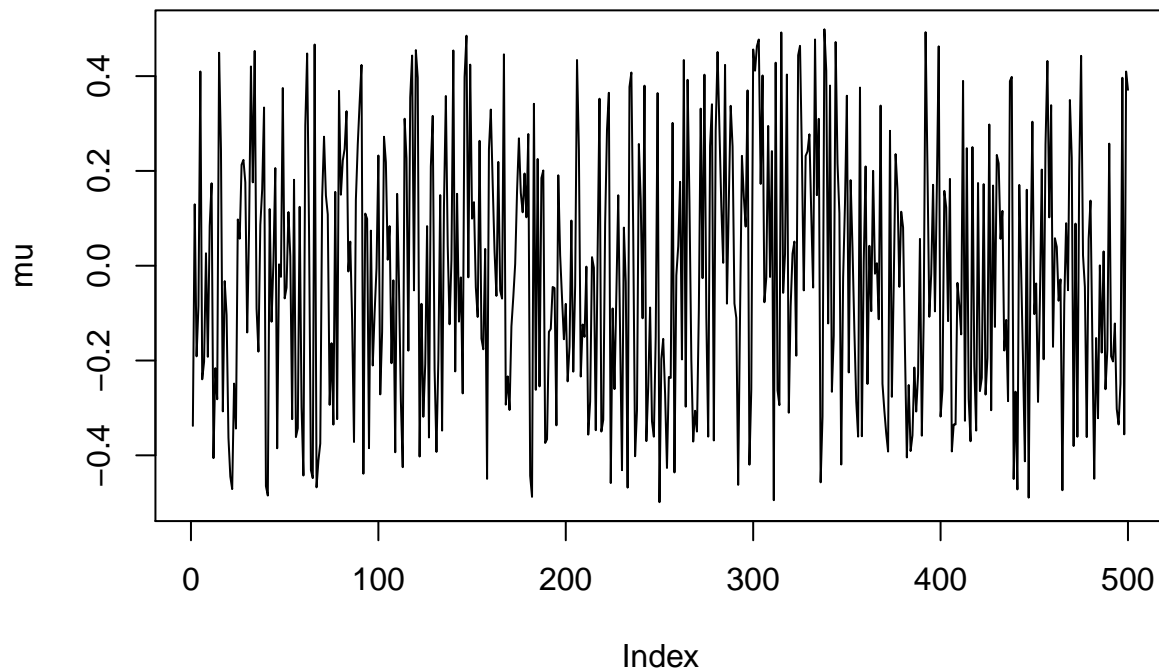
mu<-tau<-rep(NA,I+1)
tau[1]<-1000
mu[1]<-1

for (i in (1:I)+1)
{
  tau[i]<-rgamma(1,a+n/2,b+0.5*sum((x-mu[i-1])^2))
  m<-tau[i]*sumx
  s<-n*tau[i]+1/s0
  mu[i]<-draw.from.post(1,m/s,sqrt(1/s))
}

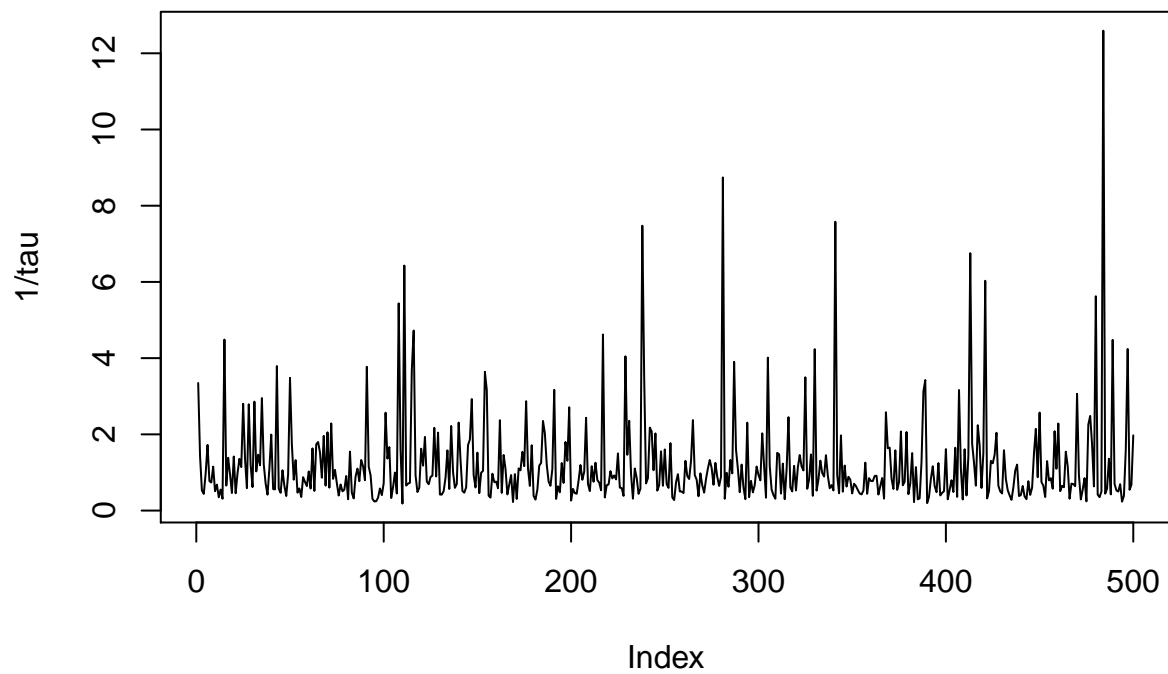
tau<-tau[-1]
mu<-mu[-1]

plot(mu,type="l")

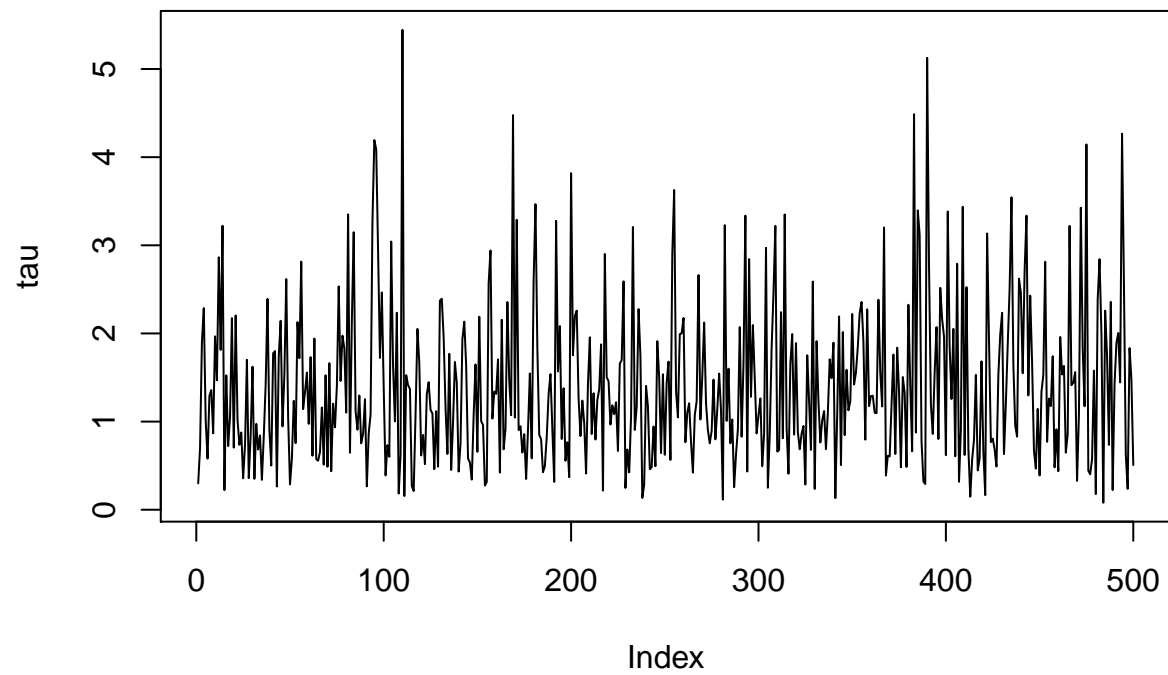
```



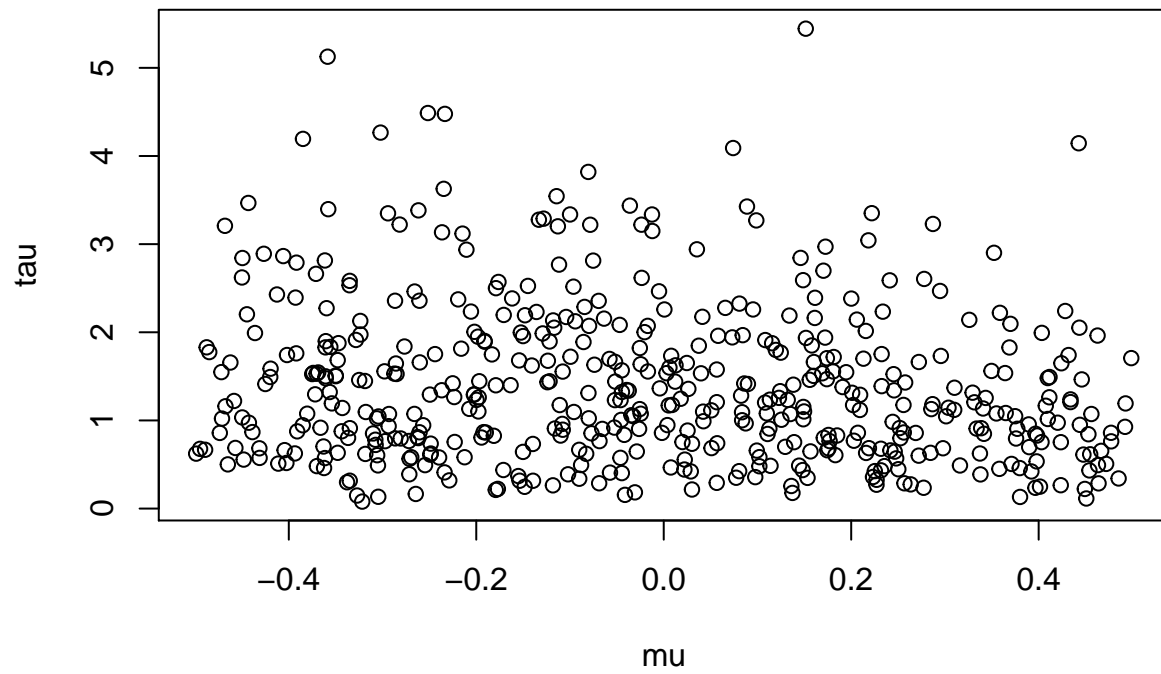
```
plot(1/tau,type="l")
```



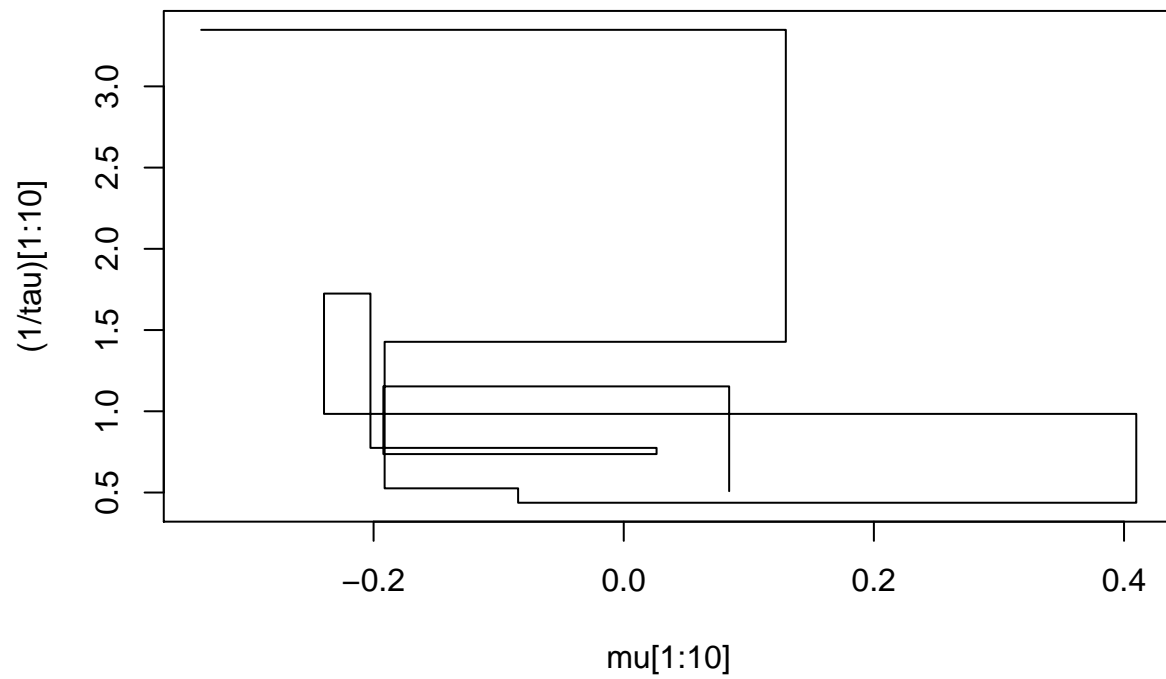
```
plot(tau,type="l")
```



```
plot(mu,tau)
```

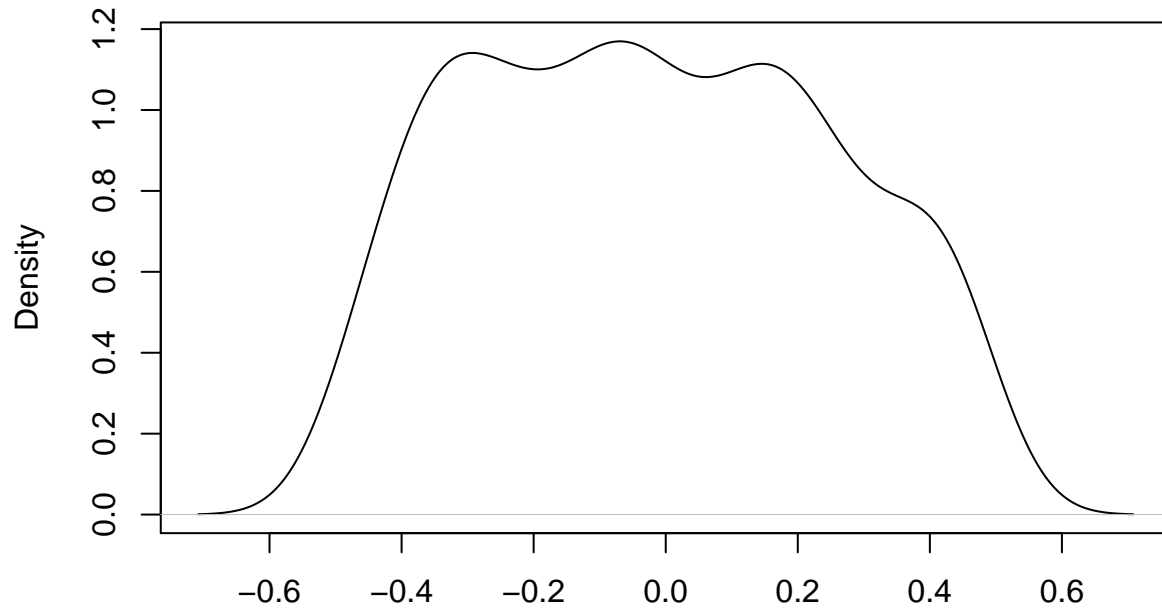


```
plot(mu[1:10], (1/tau)[1:10], type="s")
```



```
plot(density(mu))
```

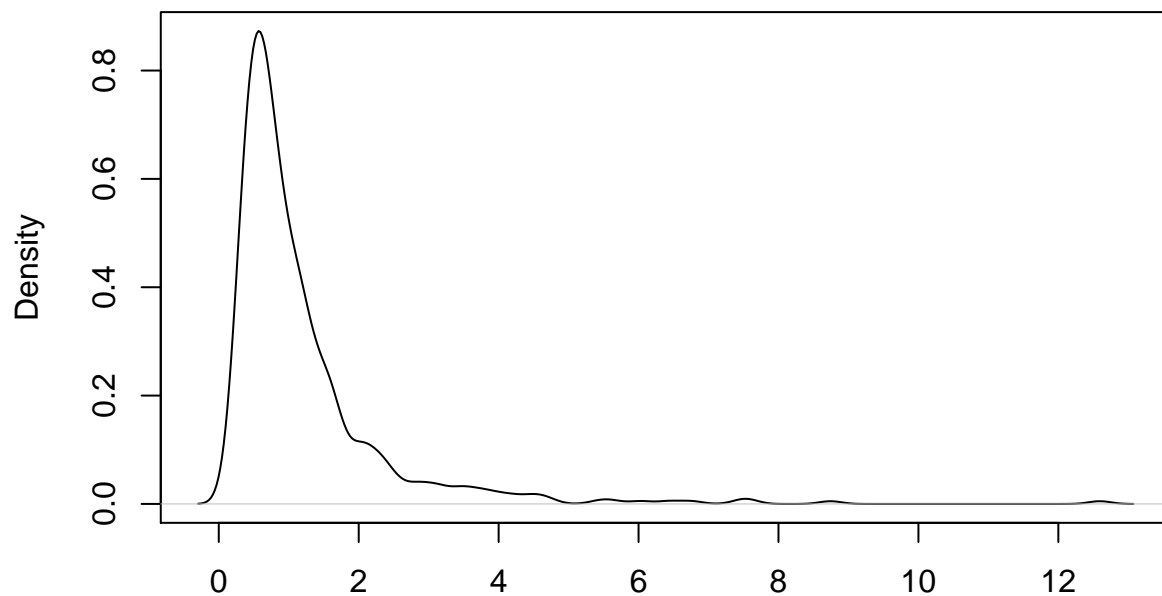
**density.default(x = mu)**



N = 500 Bandwidth = 0.06975

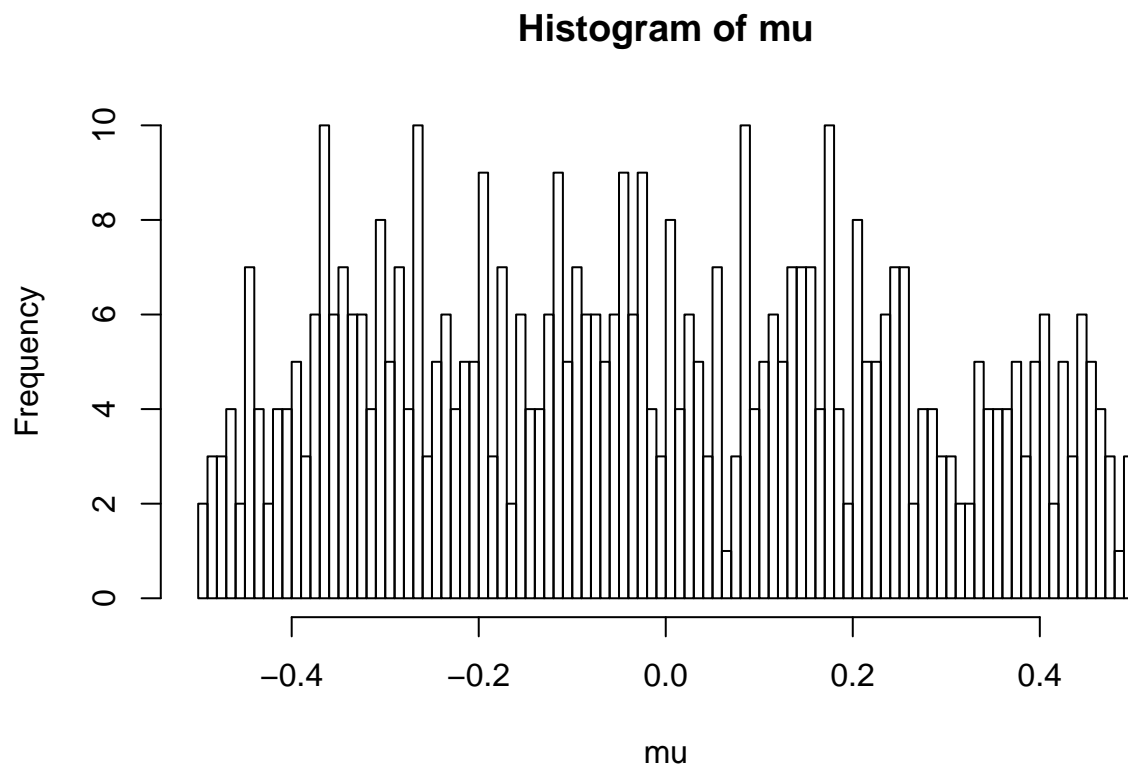
```
plot(density(1/tau))
```

**density.default(x = 1/tau)**



N = 500 Bandwidth = 0.1594

```
hist(mu,breaks=100)
```



```
print(mean(mu))
```

```
## [1] -0.01710357
```

```
print(mean(tau))
```

```
## [1] 1.394044
```

## Metropolis-Hastings-Sampling

Sei nun  $\tau \sim LN(0, 1)$ . Ziehe aus der gemeinsamen Posteriori mittels Metropolis\_Hastings.

```
I<-500
```

```
s0<-1
```

```
a<-5
```

```
b<-1
```

```
n<-length(x)
```

```
mu<-tau<-rep(NA,I+1)
```

```
tau[1]<-1 # !
```

```
mu[1]<-1 # !
```

```
log.fc.tau <- function(tau,n,sumxmu2)
```

```
{
```

```
  if (tau<0|tau>1)return(0)
```

```
  return<-n*log(tau)/2-tau*sumxmu2/2+log(dlnorm(tau))
```

```
  return(return)
```

```
}
```

```

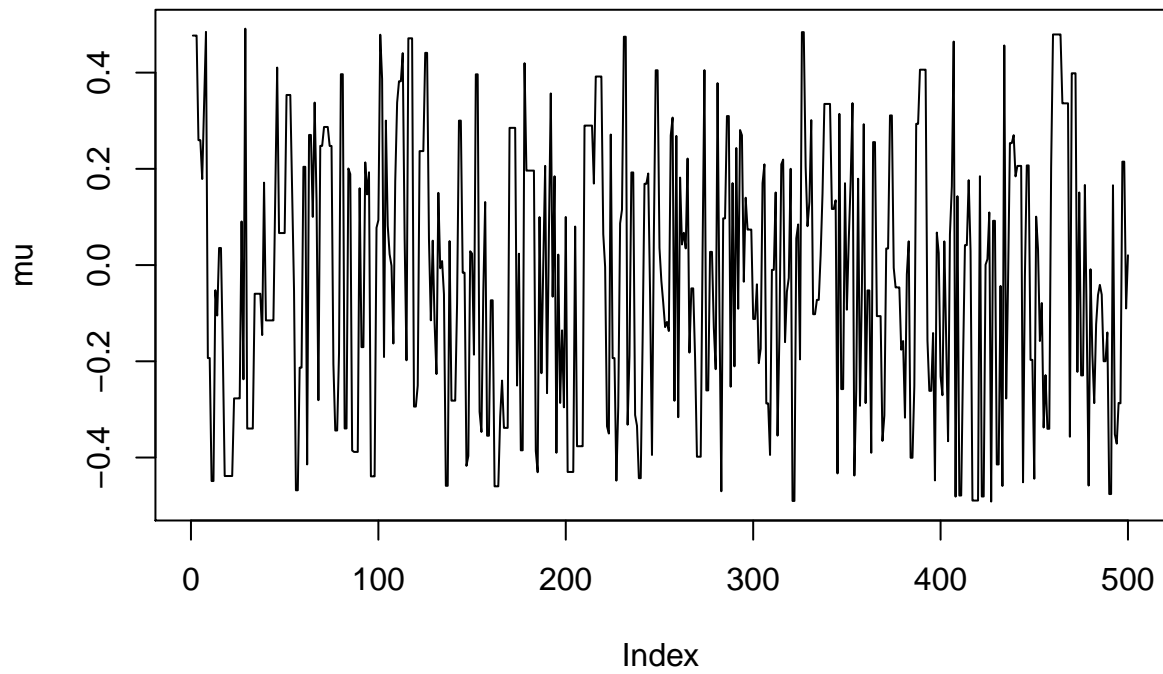
for (i in (1:I)+1)
{
  taustern <- rnorm(1,tau[i-1],.1)
  sumxmu2<-sum((x-mu[i-1])^2)
  logalpha<-log.fc.tau(taustern,n,sumxmu2)-log.fc.tau(tau[i-1],n,sumxmu2)
  alpha<-exp(logalpha)
  if (runif(1)<alpha)
  {
    tau[i]<-taustern
  }
  else
  {
    tau[i]<-tau[i-1]
  }

  m<-tau[i]*sumx
  s<-n*tau[i]+1/s0
  mustern<-rnorm(1,m/s,sqrt(1/s))
  logalpha <- log(post(mustern,sumx,spanne))-log(post(mu[i-1],sumx,spanne))
  logalpha <- logalpha+log(dnorm(mu[i-1],m/s,sqrt(1/s)))-log(dnorm(mustern,m/s,sqrt(1/s)))
  alpha<-exp(logalpha)
  if(!is.na(alpha))if(!is.na(mustern))
  {
    if (runif(1)<alpha)
    {
      mu[i]<-mustern
    }
    else
    {
      mu[i]<-mu[i-1]
    }
  }
}

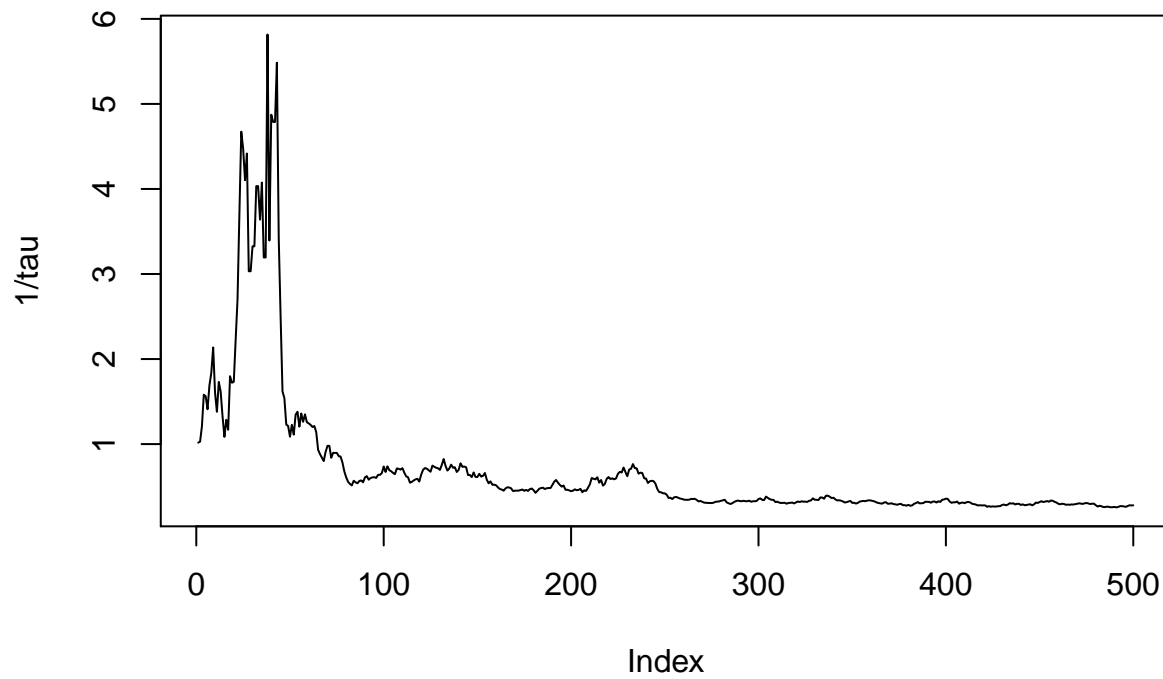
tau<-tau[-1]
mu<-mu[-1]

plot(mu,type="l")

```

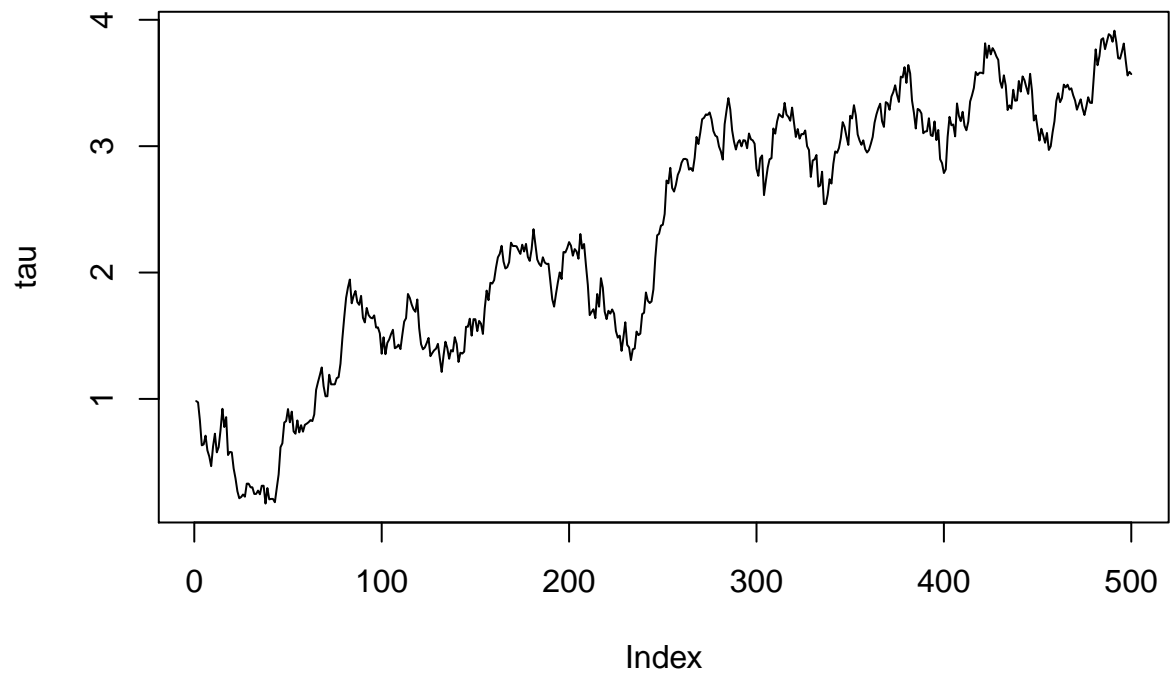


```
plot(1/tau,type="l")
```

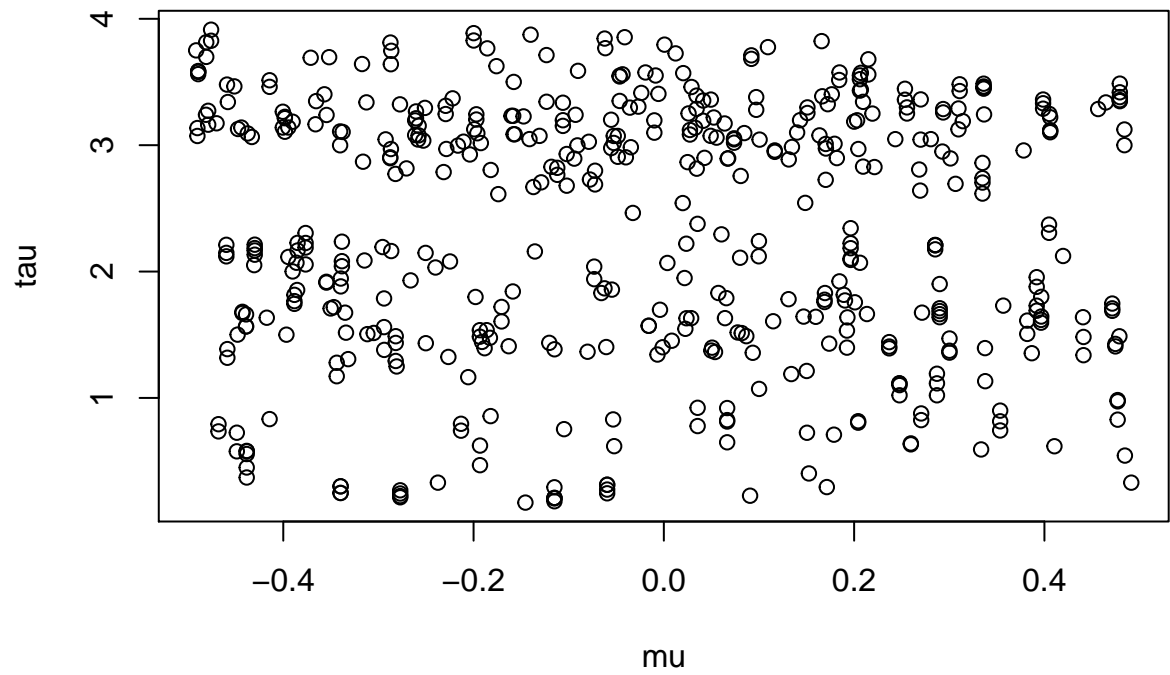


```
plot(tau,type="l")
```

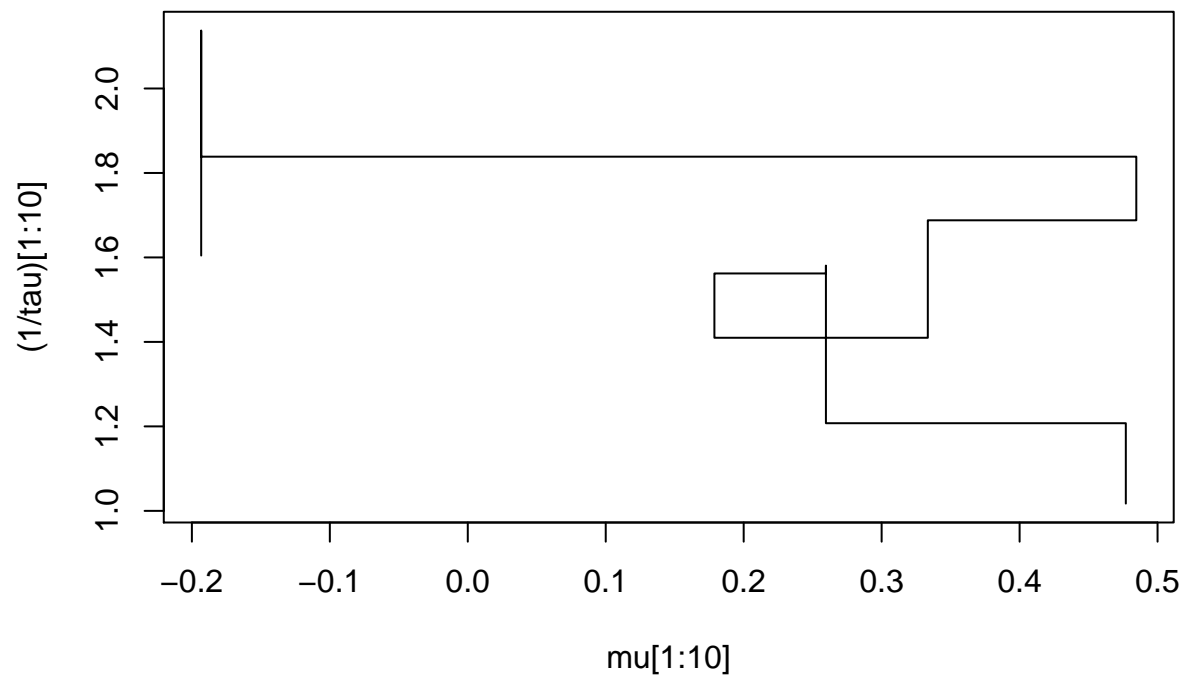




```
plot(mu,tau)
```

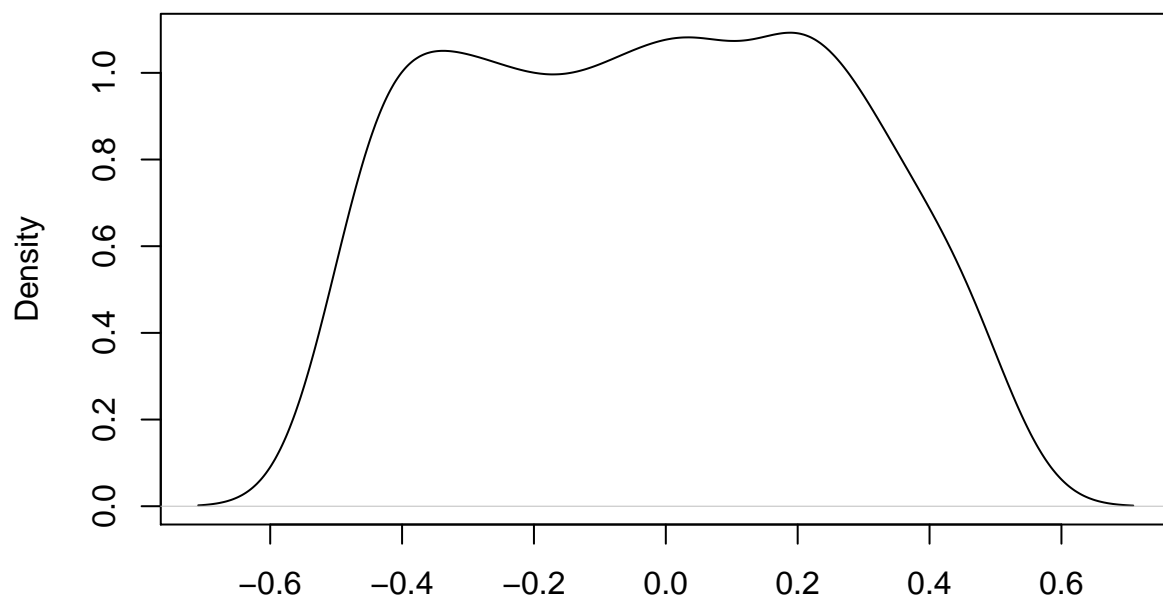


```
plot(mu[1:10],(1/tau)[1:10],type="s")
```



```
plot(density(mu))
```

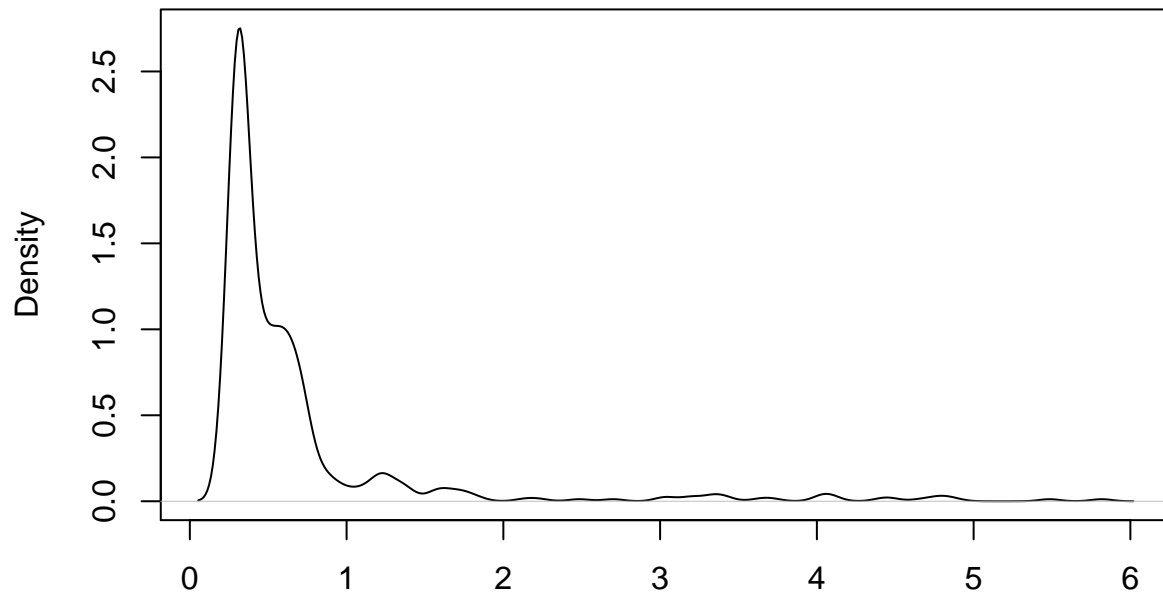
**density.default(x = mu)**



N = 500 Bandwidth = 0.07256

```
plot(density(1/tau))
```

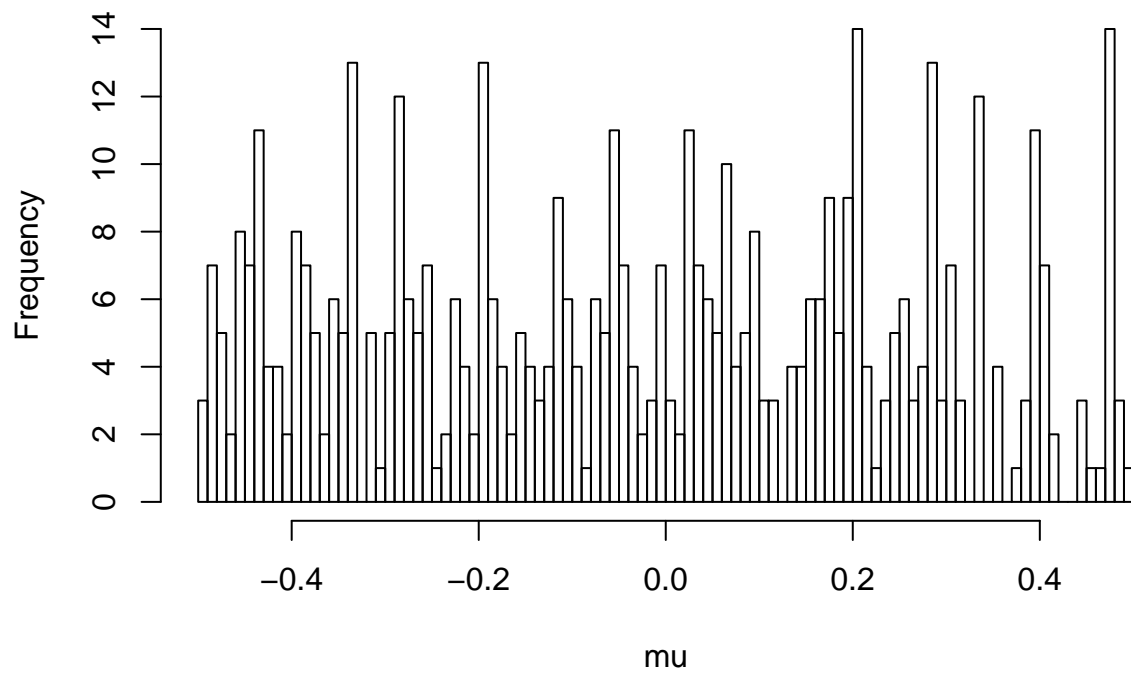
**density.default(x = 1/tau)**



N = 500 Bandwidth = 0.06743

```
hist(mu,breaks=100)
```

**Histogram of mu**



```
print(mean(mu))
```

```
## [1] -0.02504598
```

```
print(mean(tau))
```

```
## [1] 2.316955
```