

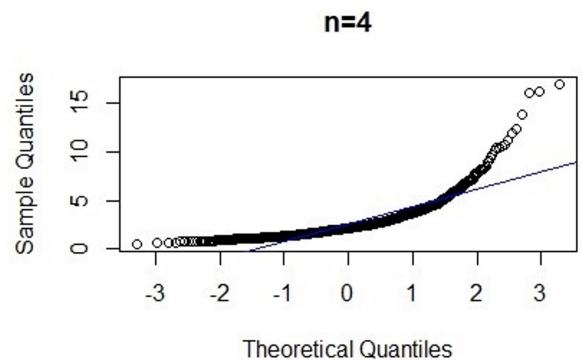
Let $X \stackrel{d}{=} \text{Exp}(\lambda)$

$$\Rightarrow E[X] = \frac{1}{\lambda}$$

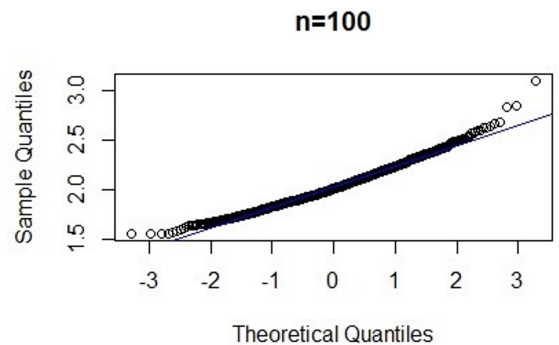
By MoM $E[X] = m_1$

$$\Rightarrow \boxed{\lambda = \frac{1}{m_1}}$$

```
> n=4
> lambda=2
> for (i in c(1:1000)) {
>   lambda_estimator=rep(0,1000)
+   lambda_estimator[i]=1/mean(rexp(n,lambda))
+ }
> qqnorm(lambda_estimator,main="n=4")
>
> abline(mean(lambda_estimator),sd(lambda_estimator),col='blue')
```



```
n=100
> lambda=2
> lambda_estimator=rep(0,1000)
> for (i in c(1:1000)) {
+   lambda_estimator[i]=1/mean(rexp(n,lambda))
+ }
> qqnorm(lambda_estimator,main="n=100")
>
> abline(mean(lambda_estimator),sd(lambda_estimator),col='blue')
```



We can observe that when we increase n then variation in the estimation of λ decreases and also the values of λ fit normal distribution more accurately.

This means that the reciprocal of mean (which is distributed normally itself due to CLT) will also resemble normal distribution when we increase n .