hw1

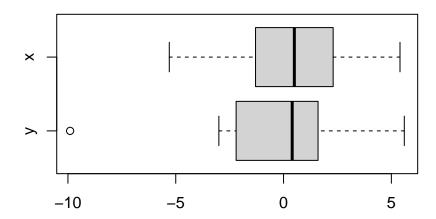
danie

2025-01-24

```
knitr::opts_chunk$set(fig.width = 5, fig.height = 3.5, fig.align = "center")
colorArea <- function(from, to, density, ..., col="blue", dens=NULL){</pre>
    y_seq <- seq(from, to, length.out=500)</pre>
    d \leftarrow c(0, density(y_seq, ...), 0)
    polygon(c(from, y_seq, to), d, col=col, density=dens)
}
1
i
Using:
x \leftarrow c(2.7, 4.0, 2.3, 5.4, -5.3, 1.8, -1.3, -2.9, 2.1, 3.9,
       -1.8, 0.4, -4.2, 0.5, -0.1, 1.5, -0.7)
y \leftarrow c(1.4, 2.5, 2.6, 5.6, -2.2, 0.4, 0.1, -3.0, 2.2, 0.9,
       -2.4, 1.6, -2.5, 0.1, -9.9, 1.1, -1.7)
Five-number summary of x:
summary(x)
                               Mean 3rd Qu.
      Min. 1st Qu. Median
                                                 Max.
## -5.3000 -1.3000 0.5000 0.4882 2.3000 5.4000
Five-number summary of y:
summary(y)
##
      Min. 1st Qu. Median
                               Mean 3rd Qu.
                                                 Max.
## -9.9000 -2.2000 0.4000 -0.1882 1.6000 5.6000
Sample variance of x:
var(x)
## [1] 8.673603
Sample variance of y:
var(y)
## [1] 11.37985
boxplot(y,x,
  main = "Distribution of xs and ys",
  names = c("y", "x"),
```

```
horizontal = TRUE
)
```

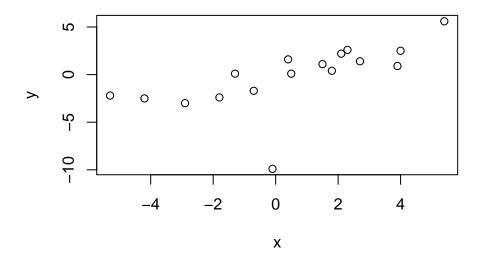
Distribution of xs and ys



The ys are skewed right while the \$xs have no skew. The outlier is the y-value -9.9 from the point (-0.1, -9.9).

ii

```
plot(x, y)
```



Correlation coefficient:

```
cor(x, y)
```

[1] 0.6289777

Which means x and y are moderately linearly correlated.

iii

Yes; (-0.1, -9.9) is an outlier.

```
x2 <- c(2.7, 4.0, 2.3, 5.4, -5.3, 1.8, -1.3, -2.9, 2.1, 3.9,

-1.8, 0.4, -4.2, 0.5, 1.5, -0.7)

y2 <- c(1.4, 2.5, 2.6, 5.6, -2.2, 0.4, 0.1, -3.0, 2.2, 0.9,

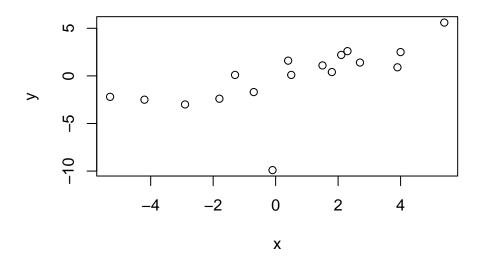
-2.4, 1.6, -2.5, 0.1, 1.1, -1.7)

cor(x2, y2)
```

[1] 0.8822511

iv

```
plot(x, y)
```



I can see the outlier (-0.1, -9.9) at the bottom-center of the graph.

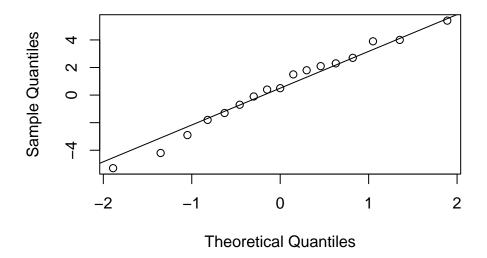
\mathbf{v}

The sample correlation coefficient in iii is much higher than the one in ii.

\mathbf{vi}

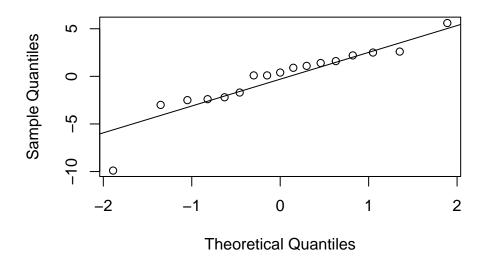
```
qqnorm(x, main="xs with outlier"); qqline(x)
```

xs with outlier



qqnorm(y, main="ys with outlier"); qqline(y)

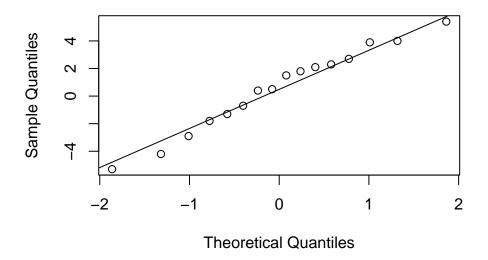
ys with outlier



The xs look much closer to normal distribution than the ys.

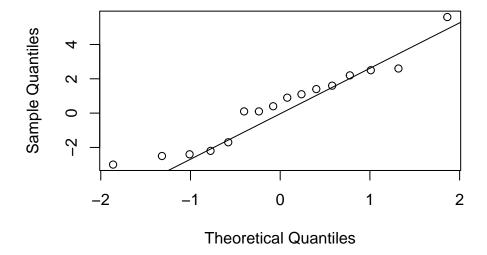
qqnorm(x2, main="xs without outlier"); qqline(x2)

xs without outlier



qqnorm(y2, main="ys without outlier"); qqline(y2)

ys without outlier



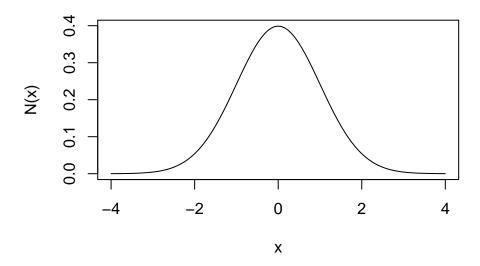
The xs still look much closer to normal distribution than the ys.

2

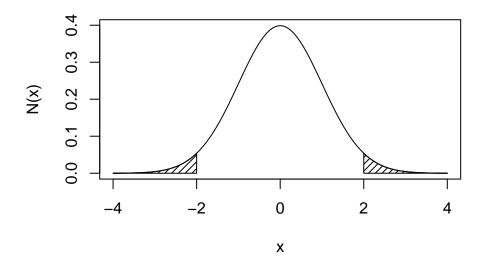
$$P(|Z| < 1) =$$
pnorm(-1) + pnorm(1, lower.tail=FALSE)

[1] 0.3173105

```
N <- function(j) dnorm(j)
curve(N, from = -4, to = 4)
colorArea(from=-4, to=-1, dnorm, col=0, dens=20) #P(Z < -1)
colorArea(from=1, to=4, dnorm, col=0, dens=20) #P(Z > 1)
```



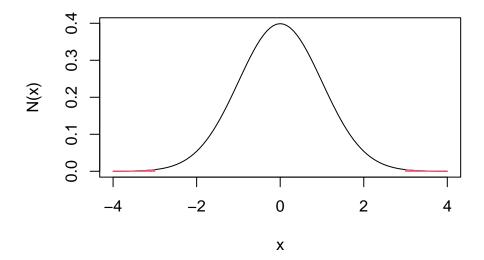
```
P(|Z| < 2) = \\ pnorm(-2) + pnorm(2, lower.tail=FALSE) \\ \#\# [1] 0.04550026 \\ curve(N, from = -4, to = 4) \\ colorArea(from=-4, to=-2, dnorm, col=1, dens=20) \#P(Z < -2) \\ colorArea(from=2, to=4, dnorm, col=1, dens=20) \#P(Z > 2) \\ \label{eq:pnorm}
```



```
P(|Z| < 3) = pnorm(-3) + pnorm(3, lower.tail=FALSE)
```

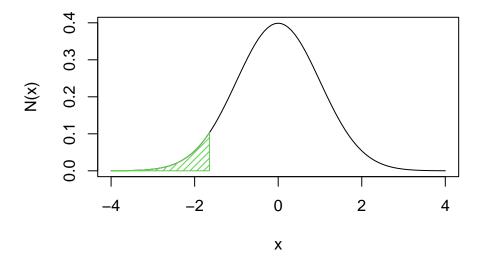
```
## [1] 0.002699796
```

```
curve(N, from = -4, to = 4) colorArea(from=-4, to=-3, dnorm, col=2, dens=20) \#P(Z < -3) colorArea(from=3, to=4, dnorm, col=2, dens=20) \#P(Z > 3)
```

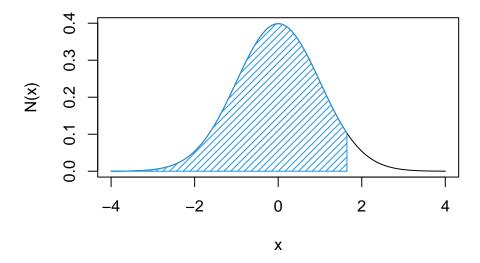


$$P(Z \le z_{0.1/2}) =$$

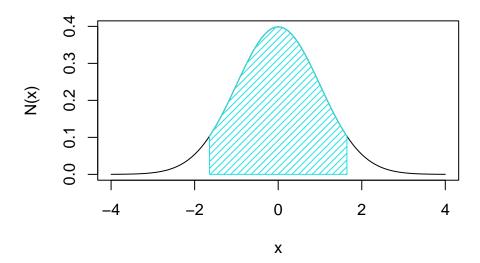
```
pnorm(qnorm(0.1/2))
## [1] 0.05
curve(N, from = -4, to = 4)
colorArea(from = -4, to = qnorm(0.1/2), dnorm, col=3, dens=20)
```



```
P(Z \le z_{1-0.1/2}) = \\ pnorm(qnorm(1-0.1/2)) \\ \#\# [1] 0.95 \\ curve(N, from = -4, to = 4) \\ colorArea(from = -4, to = qnorm(1 - 0.1/2), dnorm, col=4, dens=20)
```



```
\begin{split} &P(z_{0.1/2} \leq Z \leq z_{1-0.1/2}) = \\ &\text{pnorm}(\text{qnorm}(0.1/2)) + \text{pnorm}(\text{qnorm}(1-0.1/2), \text{lower.tail=FALSE}) \\ &\text{## [1] 0.1} \\ &\text{curve}(\mathbb{N}, \text{ from = -4, to = 4}) \\ &\text{colorArea}(\text{from = qnorm}(0.1/2), \text{ to = qnorm}(1 - 0.1/2), \text{ dnorm, col=5, dens=20}) \end{split}
```



3

We know F' is the inverse of cdf $F(x) = P(X \le x)$, which means $P(X \le F'(x)) = F(F'(x)) = x$. This means:

- $P(X \le F^{-1}(\alpha/2)) = \alpha/2$.
- This probability will decrease with a decrease in α .

 $P(X > F^{-1}(1 \alpha/2)) = 1 P(X \le F^{-1}(1 \alpha/2)) = 1 1 \alpha/2 = -\alpha/2$.

 This probability will increase with a decrease in α .
- $P(F^{-1}(\alpha/2) \le X \le F^{-1}(1-\alpha/2)) = P(X \le F^{-1}(1-\alpha/2)) P(X \le F^{-1}(\alpha/2)) = 1 \alpha/2 \alpha/2 = 1 \alpha$. This probability will increase with a decrease in α .