

BST225

HW5

1a)

Calculate sample size using all these values, change values for one factor at one time and keep other factors the same; and make a sample size analysis table; then summarize how difference between μ_1 and μ_0 , σ and power affect sample size

```
for (power in c(0.95,0.9,0.85,0.8))
{
  for (sigma_1 in c(2.2,2.0,1.8))
  {
    for (sigma_2 in c(1.8))
    {
      for (mu1_min_mu2 in c(1.5,1.0,0.5))
      {
        n <- ((sigma_1^2 + sigma_2^2)/(mu1_min_mu2)^2)*(qnorm(0.05/2)+qnorm(1-power))^2
        print (sprintf("Power: %s Sigma1: %s Sigma2: %s Mu1-Mu2: %s Sample size: %s", power, sigma_1, sigma_2, mu1_min_mu2, n))
      }
    }
  }
}
```

```
## [1] "Power: 0.95 Sigma1: 2.2 Sigma2: 1.8 Mu1-Mu2: 1.5 Sample size: 47"
## [1] "Power: 0.95 Sigma1: 2.2 Sigma2: 1.8 Mu1-Mu2: 1 Sample size: 105"
## [1] "Power: 0.95 Sigma1: 2.2 Sigma2: 1.8 Mu1-Mu2: 0.5 Sample size: 420"
## [1] "Power: 0.95 Sigma1: 2 Sigma2: 1.8 Mu1-Mu2: 1.5 Sample size: 42"
## [1] "Power: 0.95 Sigma1: 2 Sigma2: 1.8 Mu1-Mu2: 1 Sample size: 94"
## [1] "Power: 0.95 Sigma1: 2 Sigma2: 1.8 Mu1-Mu2: 0.5 Sample size: 376"
## [1] "Power: 0.95 Sigma1: 1.8 Sigma2: 1.8 Mu1-Mu2: 1.5 Sample size: 37"
## [1] "Power: 0.95 Sigma1: 1.8 Sigma2: 1.8 Mu1-Mu2: 1 Sample size: 84"
## [1] "Power: 0.95 Sigma1: 1.8 Sigma2: 1.8 Mu1-Mu2: 0.5 Sample size: 337"
## [1] "Power: 0.9 Sigma1: 2.2 Sigma2: 1.8 Mu1-Mu2: 1.5 Sample size: 38"
## [1] "Power: 0.9 Sigma1: 2.2 Sigma2: 1.8 Mu1-Mu2: 1 Sample size: 85"
## [1] "Power: 0.9 Sigma1: 2.2 Sigma2: 1.8 Mu1-Mu2: 0.5 Sample size: 340"
## [1] "Power: 0.9 Sigma1: 2 Sigma2: 1.8 Mu1-Mu2: 1.5 Sample size: 34"
## [1] "Power: 0.9 Sigma1: 2 Sigma2: 1.8 Mu1-Mu2: 1 Sample size: 76"
## [1] "Power: 0.9 Sigma1: 2 Sigma2: 1.8 Mu1-Mu2: 0.5 Sample size: 304"
## [1] "Power: 0.9 Sigma1: 1.8 Sigma2: 1.8 Mu1-Mu2: 1.5 Sample size: 30"
## [1] "Power: 0.9 Sigma1: 1.8 Sigma2: 1.8 Mu1-Mu2: 1 Sample size: 68"
## [1] "Power: 0.9 Sigma1: 1.8 Sigma2: 1.8 Mu1-Mu2: 0.5 Sample size: 272"
## [1] "Power: 0.85 Sigma1: 2.2 Sigma2: 1.8 Mu1-Mu2: 1.5 Sample size: 32"
## [1] "Power: 0.85 Sigma1: 2.2 Sigma2: 1.8 Mu1-Mu2: 1 Sample size: 73"
```

```
## [1] "Power: 0.85 Sigma1: 2.2 Sigma2: 1.8 Mu1-Mu2: 0.5 Sample size: 290"
## [1] "Power: 0.85 Sigma1: 2 Sigma2: 1.8 Mu1-Mu2: 1.5 Sample size: 29"
## [1] "Power: 0.85 Sigma1: 2 Sigma2: 1.8 Mu1-Mu2: 1 Sample size: 65"
## [1] "Power: 0.85 Sigma1: 2 Sigma2: 1.8 Mu1-Mu2: 0.5 Sample size: 260"
## [1] "Power: 0.85 Sigma1: 1.8 Sigma2: 1.8 Mu1-Mu2: 1.5 Sample size: 26"
## [1] "Power: 0.85 Sigma1: 1.8 Sigma2: 1.8 Mu1-Mu2: 1 Sample size: 58"
## [1] "Power: 0.85 Sigma1: 1.8 Sigma2: 1.8 Mu1-Mu2: 0.5 Sample size: 233"
## [1] "Power: 0.8 Sigma1: 2.2 Sigma2: 1.8 Mu1-Mu2: 1.5 Sample size: 28"
## [1] "Power: 0.8 Sigma1: 2.2 Sigma2: 1.8 Mu1-Mu2: 1 Sample size: 63"
## [1] "Power: 0.8 Sigma1: 2.2 Sigma2: 1.8 Mu1-Mu2: 0.5 Sample size: 254"
## [1] "Power: 0.8 Sigma1: 2 Sigma2: 1.8 Mu1-Mu2: 1.5 Sample size: 25"
## [1] "Power: 0.8 Sigma1: 2 Sigma2: 1.8 Mu1-Mu2: 1 Sample size: 57"
## [1] "Power: 0.8 Sigma1: 2 Sigma2: 1.8 Mu1-Mu2: 0.5 Sample size: 227"
## [1] "Power: 0.8 Sigma1: 1.8 Sigma2: 1.8 Mu1-Mu2: 1.5 Sample size: 23"
## [1] "Power: 0.8 Sigma1: 1.8 Sigma2: 1.8 Mu1-Mu2: 1 Sample size: 51"
## [1] "Power: 0.8 Sigma1: 1.8 Sigma2: 1.8 Mu1-Mu2: 0.5 Sample size: 203"
```

- As difference between μ_1 and μ_2 decrease the sample size needs to increase.
- As the pooled variance decreases then the necessary sample size will also decrease.
- As the power decreases the sample size decreases.

1b) Is this what is wanted?Z

300 and 700 as n_1 and n_2 Calculate power using parameters values and make a power analysis table and summarize the effect of the parameters.

For n_2 , For n_1 (same result as n_2)

```
n1 <- 300
n2 <- 700
k <- n1/n2

for (sigma_2 in c(2.2,2.0,1.8))
{
  for (mu1_min_mu2 in c(1.5,1.0,0.5))
  {
    q_norm_power <- sqrt(n2/((sigma_2^2*(1+1/k))/(mu1_min_mu2)^2)) + qnorm(0.05/2)
    print(sprintf("Power: %s Sigma1: %s Sigma2: %s Mu1-Mu2: %s ", pnorm(q_norm_power), sigma_1, si
  )
}
```

```
## [1] "Power: 0.999999999999999 Sigma1: 1.8 Sigma2: 2.2 Mu1-Mu2: 1.5 "
## [1] "Power: 0.999998145225234 Sigma1: 1.8 Sigma2: 2.2 Mu1-Mu2: 1 "
## [1] "Power: 0.908821151138328 Sigma1: 1.8 Sigma2: 2.2 Mu1-Mu2: 0.5 "
## [1] "Power: 1 Sigma1: 1.8 Sigma2: 2 Mu1-Mu2: 1.5 "
## [1] "Power: 0.999999937395856 Sigma1: 1.8 Sigma2: 2 Mu1-Mu2: 1 "
## [1] "Power: 0.95183179092774 Sigma1: 1.8 Sigma2: 2 Mu1-Mu2: 0.5 "
## [1] "Power: 1 Sigma1: 1.8 Sigma2: 1.8 Mu1-Mu2: 1.5 "
## [1] "Power: 0.999999999438264 Sigma1: 1.8 Sigma2: 1.8 Mu1-Mu2: 1 "
## [1] "Power: 0.980558287587443 Sigma1: 1.8 Sigma2: 1.8 Mu1-Mu2: 0.5 "
```

- As the sample size increase so does the power
- As the difference between the two groups increases then the power increases
- As the pooled variance between the two groups increases the the power decreases

2)

Test non-inferiority with two different margins and study how the values of margins affect the sample size.

```
for (power in c(0.9,0.8))
{
  for (sigma_1 in c(2.2))
  {
    for (sigma_2 in c(1.8))
    {
      for (mu1_min_mu2 in c(1.0,0.5))
      {
        for (margin in c(1.1,0.4))
        {
          n <- ((sigma_1^2 + sigma_2^2)/(mu1_min_mu2-margin)^2)*(qnorm(0.05/2)+qnorm(1-power))^2
          print (sprintf("Power: %s  Sigma1: %s  Sigma2: %s  Mu1-Mu2: %s  Sample size: %s  Margin: %s",
            power, sigma_1, sigma_2, mu1_min_mu2, n, margin))
        }
      }
    }
  }
}
```

```
## [1] "Power: 0.9  Sigma1: 2.2  Sigma2: 1.8  Mu1-Mu2: 1  Sample size: 8490  Margin: 1.1"
## [1] "Power: 0.9  Sigma1: 2.2  Sigma2: 1.8  Mu1-Mu2: 1  Sample size: 236  Margin: 0.4"
## [1] "Power: 0.9  Sigma1: 2.2  Sigma2: 1.8  Mu1-Mu2: 0.5  Sample size: 236  Margin: 1.1"
## [1] "Power: 0.9  Sigma1: 2.2  Sigma2: 1.8  Mu1-Mu2: 0.5  Sample size: 8490  Margin: 0.4"
## [1] "Power: 0.8  Sigma1: 2.2  Sigma2: 1.8  Mu1-Mu2: 1  Sample size: 6342  Margin: 1.1"
## [1] "Power: 0.8  Sigma1: 2.2  Sigma2: 1.8  Mu1-Mu2: 1  Sample size: 176  Margin: 0.4"
## [1] "Power: 0.8  Sigma1: 2.2  Sigma2: 1.8  Mu1-Mu2: 0.5  Sample size: 176  Margin: 1.1"
## [1] "Power: 0.8  Sigma1: 2.2  Sigma2: 1.8  Mu1-Mu2: 0.5  Sample size: 6342  Margin: 0.4"
```

- As the difference between mu1-mu2 and margin decreases then the necessary sample size to show inferiority increases.

Test equivalence with two different margins and study how the values of margins affect the sample size.

```
for (power in c(0.9,0.8))
{
  for (sigma_1 in c(2.2))
  {
    for (sigma_2 in c(1.8))
    {
      for (mu1_min_mu2 in c(1.0,0.5))
      {
        for (margin in c(1.1,0.4))
        {
          n <- ((sigma_1^2 + sigma_2^2)/(mu1_min_mu2-margin)^2)*(qnorm(0.05/2)+qnorm((1-power)/2))^2
          print (sprintf("Power: %s  Sigma1: %s  Sigma2: %s  Mu1-Mu2: %s  Sample size: %s  Margin: %s",
            power, sigma_1, sigma_2, mu1_min_mu2, n, margin))
        }
      }
    }
  }
}
```

```

    }
  }
}
}
}

```

```

## [1] "Power: 0.9 Sigma1: 2.2 Sigma2: 1.8 Mu1-Mu2: 1 Sample size: 10500 Margin: 1.1"
## [1] "Power: 0.9 Sigma1: 2.2 Sigma2: 1.8 Mu1-Mu2: 1 Sample size: 292 Margin: 0.4"
## [1] "Power: 0.9 Sigma1: 2.2 Sigma2: 1.8 Mu1-Mu2: 0.5 Sample size: 292 Margin: 1.1"
## [1] "Power: 0.9 Sigma1: 2.2 Sigma2: 1.8 Mu1-Mu2: 0.5 Sample size: 10500 Margin: 0.4"
## [1] "Power: 0.8 Sigma1: 2.2 Sigma2: 1.8 Mu1-Mu2: 1 Sample size: 8490 Margin: 1.1"
## [1] "Power: 0.8 Sigma1: 2.2 Sigma2: 1.8 Mu1-Mu2: 1 Sample size: 236 Margin: 0.4"
## [1] "Power: 0.8 Sigma1: 2.2 Sigma2: 1.8 Mu1-Mu2: 0.5 Sample size: 236 Margin: 1.1"
## [1] "Power: 0.8 Sigma1: 2.2 Sigma2: 1.8 Mu1-Mu2: 0.5 Sample size: 8490 Margin: 0.4"

```

- As the difference between $\text{abs}(\mu_1 - \mu_2)$ and margin decreases then the necessary sample size to show equivalence increases.

HW6

1a)

Calculate sample size using all these values, change values for one factor at one time and keep other factors the same; and make a sample size analysis table; then summarize how difference between p_1 and p_2 , sigma and power affect sample size

```

for (power in c(0.95,0.9,0.85,0.8))
{
  for (p_1 in c(0.85))
  {
    for (p_2 in c(0.8,0.7,0.6))
    {
      n <- ((p_1*(1-p_1) + p_2*(1-p_2))/(p_1-p_2)^2)*(qnorm(0.05/2)+qnorm(1-power))^2
      print(sprintf("Power: %s P1: %s P2: %s Sample size: %s", power, p_1, p_2, round(n)))
    }
  }
}

```

```

## [1] "Power: 0.95 P1: 0.85 P2: 0.8 Sample size: 1494"
## [1] "Power: 0.95 P1: 0.85 P2: 0.7 Sample size: 195"
## [1] "Power: 0.95 P1: 0.85 P2: 0.6 Sample size: 76"
## [1] "Power: 0.9 P1: 0.85 P2: 0.8 Sample size: 1208"
## [1] "Power: 0.9 P1: 0.85 P2: 0.7 Sample size: 158"
## [1] "Power: 0.9 P1: 0.85 P2: 0.6 Sample size: 62"
## [1] "Power: 0.85 P1: 0.85 P2: 0.8 Sample size: 1033"
## [1] "Power: 0.85 P1: 0.85 P2: 0.7 Sample size: 135"
## [1] "Power: 0.85 P1: 0.85 P2: 0.6 Sample size: 53"
## [1] "Power: 0.8 P1: 0.85 P2: 0.8 Sample size: 903"
## [1] "Power: 0.8 P1: 0.85 P2: 0.7 Sample size: 118"
## [1] "Power: 0.8 P1: 0.85 P2: 0.6 Sample size: 46"

```

- As power decreases so does the necessary sample size
- As the difference between p_1 and p_2 increases the sample size decreases

1b)

Sample size = 90,170 for each arm, calculate power using the above parameter values, change values for one factor at one time and keep other factors the same. Make a power analysis table and summarize how the difference between p_1 and p_2 , and sample size affect power.

```
for (n_1 in c(80,90))
{
  for (n_2 in c(170,180,190))
  {
    for (p_1 in c(0.85))
    {
      for (p_2 in c(0.8,0.7,0.6))
      {
        k <- n_1/n_2
        q_norm_power <- sqrt(n/(((p_1*(1-p_1))/k + p_2*(1-p_2))/(p_1-p_2)^2))) + qnorm(0.05/2)
        print(sprintf("Power: %s P1: %s P2: %s N1: %s N2: %s", pnorm(q_norm_power), p_1, p_2, n_1, n_2))
      }
    }
  }
}
```

```
## [1] "Power: 0.0745765264369497 P1: 0.85 P2: 0.8 N1: 80 N2: 170"
## [1] "Power: 0.311866717973062 P1: 0.85 P2: 0.7 N1: 80 N2: 170"
## [1] "Power: 0.661312545718699 P1: 0.85 P2: 0.6 N1: 80 N2: 170"
## [1] "Power: 0.0732730136666671 P1: 0.85 P2: 0.8 N1: 80 N2: 180"
## [1] "Power: 0.303512782702634 P1: 0.85 P2: 0.7 N1: 80 N2: 180"
## [1] "Power: 0.64796637980382 P1: 0.85 P2: 0.6 N1: 80 N2: 180"
## [1] "Power: 0.0720535859616882 P1: 0.85 P2: 0.8 N1: 80 N2: 190"
## [1] "Power: 0.295645745736641 P1: 0.85 P2: 0.7 N1: 80 N2: 190"
## [1] "Power: 0.635039488684074 P1: 0.85 P2: 0.6 N1: 80 N2: 190"
## [1] "Power: 0.0773027916025928 P1: 0.85 P2: 0.8 N1: 90 N2: 170"
## [1] "Power: 0.329136219441028 P1: 0.85 P2: 0.7 N1: 90 N2: 170"
## [1] "Power: 0.68768459899859 P1: 0.85 P2: 0.6 N1: 90 N2: 170"
## [1] "Power: 0.0759735908914191 P1: 0.85 P2: 0.8 N1: 90 N2: 180"
## [1] "Power: 0.320752018700229 P1: 0.85 P2: 0.7 N1: 90 N2: 180"
## [1] "Power: 0.675083832812154 P1: 0.85 P2: 0.6 N1: 90 N2: 180"
## [1] "Power: 0.0747269493569641 P1: 0.85 P2: 0.8 N1: 90 N2: 190"
## [1] "Power: 0.312826871764853 P1: 0.85 P2: 0.7 N1: 90 N2: 190"
## [1] "Power: 0.662821610429767 P1: 0.85 P2: 0.6 N1: 90 N2: 190"
```

- As the difference between p_1 and p_2 increases then so does the power
- As the sample size increases for n_1 then the power increases.
- As the sample size increase for n_2 then the power increases

HW7

Trial to compare a new therapy with a routine bath care in terms of the time to infection. - Hazard ratio (HR) of 1.5, 2.5, ($\beta = \log(\text{HR})$), respectively, for routine bathing care/test therapy is considered of clinical importance. - 60%, 80% of patients' infection may be observed, respectively (d) - $n = n_1 = n_2$ ($p_1 = p_2 = 0.5$), equal size treatment groups - Significance level: $\alpha = 0.05$

1A)

If Power = 0.8, $b = 1 - 0.8 = 0.2$, calculate sample size for above various HRs and d 's.

```
for (d in c(0.6, 0.8))
{
  for (hr in c(1.5, 2.0, 2.5))
  {
    n <- (qnorm(0.05/2) + qnorm(1-0.8))^2/(log(hr)^2*0.5*0.5*d)
    print(sprintf("HR: %s Patiends observed(d): %s Sample size: %s", hr, d, n))
  }
}
```

```
## [1] "HR: 1.5 Patiends observed(d): 0.6 Sample size: 318.280066973315"
## [1] "HR: 2 Patiends observed(d): 0.6 Sample size: 108.909432098118"
## [1] "HR: 2.5 Patiends observed(d): 0.6 Sample size: 62.3232120253598"
## [1] "HR: 1.5 Patiends observed(d): 0.8 Sample size: 238.710050229987"
## [1] "HR: 2 Patiends observed(d): 0.8 Sample size: 81.6820740735886"
## [1] "HR: 2.5 Patiends observed(d): 0.8 Sample size: 46.7424090190198"
```

1B)

If $n_1 = n_2 = 20, 50$, respectively, calculate power for above various HRs and d 's.

```
for (n in c(20,50))
{
  for (d in c(0.6, 0.8))
  {
    for (hr in c(1.5, 2.5))
    {
      q_norm_power <- - sqrt((n*(log(hr)^2*0.5*0.5*d))) - (qnorm(0.05/2))
      print(sprintf("HR: %s Patiends observed(d): %s Sample size: %s Power: %s", hr, d, n, pnorm(q_norm_power)))
    }
  }
}
```

```
## [1] "HR: 1.5 Patiends observed(d): 0.6 Sample size: 20 Power: 0.895745850962353"
## [1] "HR: 2.5 Patiends observed(d): 0.6 Sample size: 20 Power: 0.64538926453217"
## [1] "HR: 1.5 Patiends observed(d): 0.8 Sample size: 20 Power: 0.874728971655608"
## [1] "HR: 2.5 Patiends observed(d): 0.8 Sample size: 20 Power: 0.550681175037879"
## [1] "HR: 1.5 Patiends observed(d): 0.6 Sample size: 50 Power: 0.80221291111564"
## [1] "HR: 2.5 Patiends observed(d): 0.6 Sample size: 50 Power: 0.291364961871538"
## [1] "HR: 1.5 Patiends observed(d): 0.8 Sample size: 50 Power: 0.751041464227333"
## [1] "HR: 2.5 Patiends observed(d): 0.8 Sample size: 50 Power: 0.174224561748361"
```