

Frequentist meta-analyses

Candice

2024-06-07

Frequentist Meta-Analysis

The included studies report five outcome measures, listed in order of increasing severity: infection, symptomatic infection, severe infection, hospitalisation, and death. For each of these outcomes, we conduct separate classical (frequentist) meta-analyses. Both fixed-effect and random-effect models are applied to each outcome measure independently. These analyses do not account for potential correlations between the different outcomes. The risk ratio (RR) is used as the relative effect measure, comparing intervention 1 to intervention 2.

1. Infection

```
# Library relevant packages for meta-analysis  
library(meta)
```

```
## Loading required package: metadat
```

```
## Loading 'meta' package (version 7.0-0).  
## Type 'help(meta)' for a brief overview.  
## Readers of 'Meta-Analysis with R (Use R!)' should install  
## older version of 'meta' package: https://tinyurl.com/dt4y5drs
```

```
# Read in the infection data-set  
# with the first row containing variable names  
infec.data <- read.csv("../raw_data/infection_data.csv", header = TRUE)  
head(infec.data)
```

```
##   author_id      age interv_id1 interv_id2      N1      n1      N2      n2  
## 1         1 65-85 years          1          2    1155     41   13613    243  
## 2         2  50+ years          1          2     420     47     478    116  
## 3         3  50+ years          1          2 158993    17 236693     68  
## 4         4  50+ years          1          2  79456   297 180790    474  
## 5         5  60+ years          1          2  57604 3089  48706 4059  
## 6         6  50+ years          1          2    466     6    1196     22
```

```
# Fixed Effect & Random Effect Meta-Analysis Models  
# Using risk ratio (RR) as the measure  
infec.model <- metabin(event.e = n1, n.e = N1,
```

```

event.c = n2, n.c = N2,
data = infec.data, sm = "RR", method = "I",
studlab = author_id, label.e = "Intervention 1", label.c = "Intervention 2")
# Summary of the meta-analysis model
summary(infec.model)

```

```

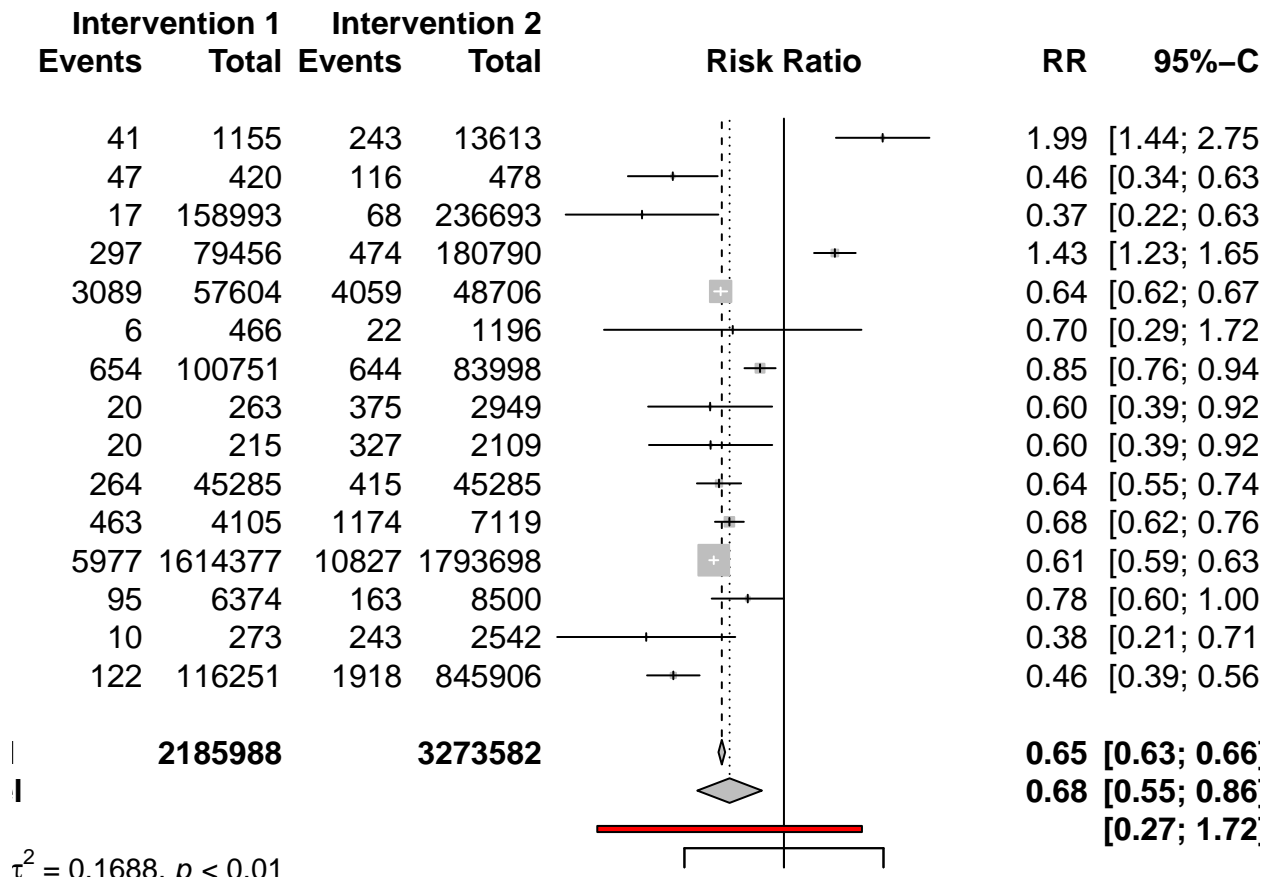
##          RR          95%-CI %W(common) %W(random)
## 1  1.9886 [1.4362; 2.7534]      0.5      6.7
## 2  0.4611 [0.3374; 0.6303]      0.6      6.8
## 3  0.3722 [0.2188; 0.6332]      0.2      5.4
## 4  1.4257 [1.2335; 1.6478]      2.6      7.6
## 5  0.6435 [0.6150; 0.6732]     26.4      7.8
## 6  0.7000 [0.2856; 1.7153]      0.1      3.5
## 7  0.8467 [0.7597; 0.9436]      4.6      7.7
## 8  0.5980 [0.3883; 0.9209]      0.3      6.1
## 9  0.6000 [0.3906; 0.9215]      0.3      6.1
## 10 0.6361 [0.5455; 0.7419]      2.3      7.5
## 11 0.6839 [0.6186; 0.7562]      5.3      7.7
## 12 0.6134 [0.5943; 0.6330]     54.3      7.8
## 13 0.7772 [0.6048; 0.9989]      0.9      7.1
## 14 0.3832 [0.2061; 0.7123]      0.1      4.9
## 15 0.4628 [0.3855; 0.5557]      1.6      7.4
##
## Number of studies: k = 15
## Number of observations: o = 5459570 (o.e = 2185988, o.c = 3273582)
## Number of events: e = 32190
##
##          RR          95%-CI      z  p-value
## Common effect model 0.6487 [0.6338; 0.6640] -36.52 < 0.0001
## Random effects model 0.6849 [0.5469; 0.8577] -3.30  0.0010
##
## Quantifying heterogeneity:
## tau^2 = 0.1688 [0.0779; 0.4644]; tau = 0.4109 [0.2790; 0.6814]
## I^2 = 93.7% [91.2%; 95.5%]; H = 3.99 [3.36; 4.73]
##
## Test of heterogeneity:
##      Q d.f.  p-value
## 222.61  14 < 0.0001
##
## Details on meta-analytical method:
## - Inverse variance method
## - Restricted maximum-likelihood estimator for tau^2
## - Q-Profile method for confidence interval of tau^2 and tau

```

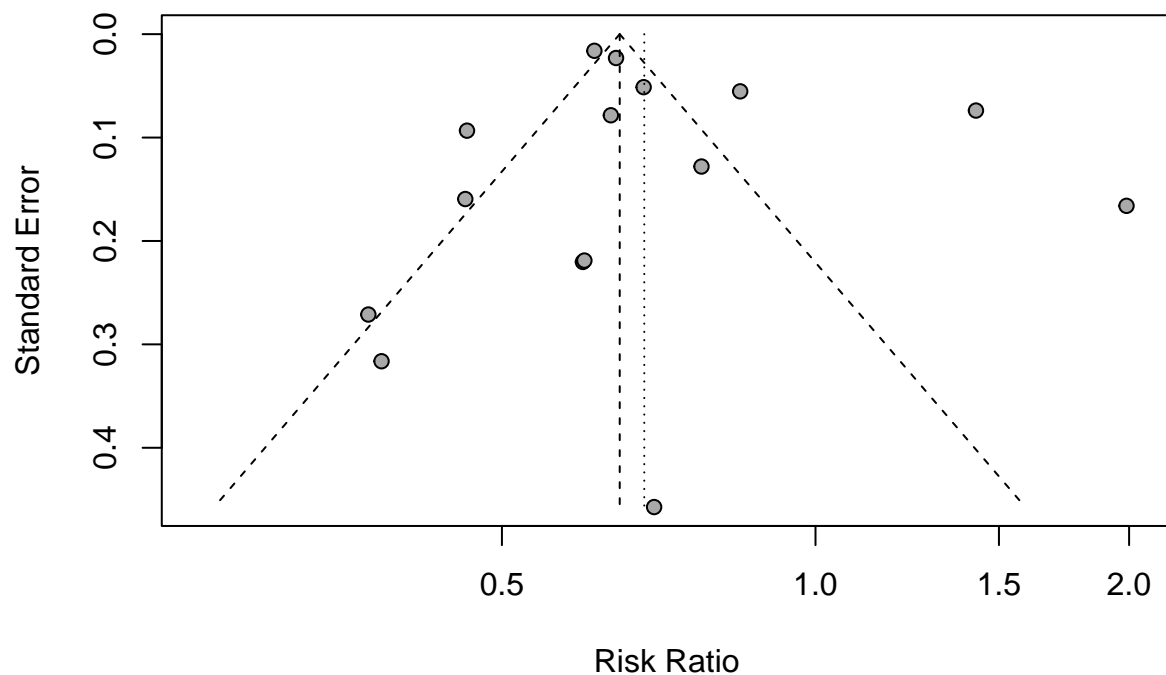
```

# Forest Plot
# Showing relative risk of intervention 1 relative to intervention 2
# with corresponding 95% confidence intervals
forest(infec.model, studlab = author_id, prediction = TRUE)

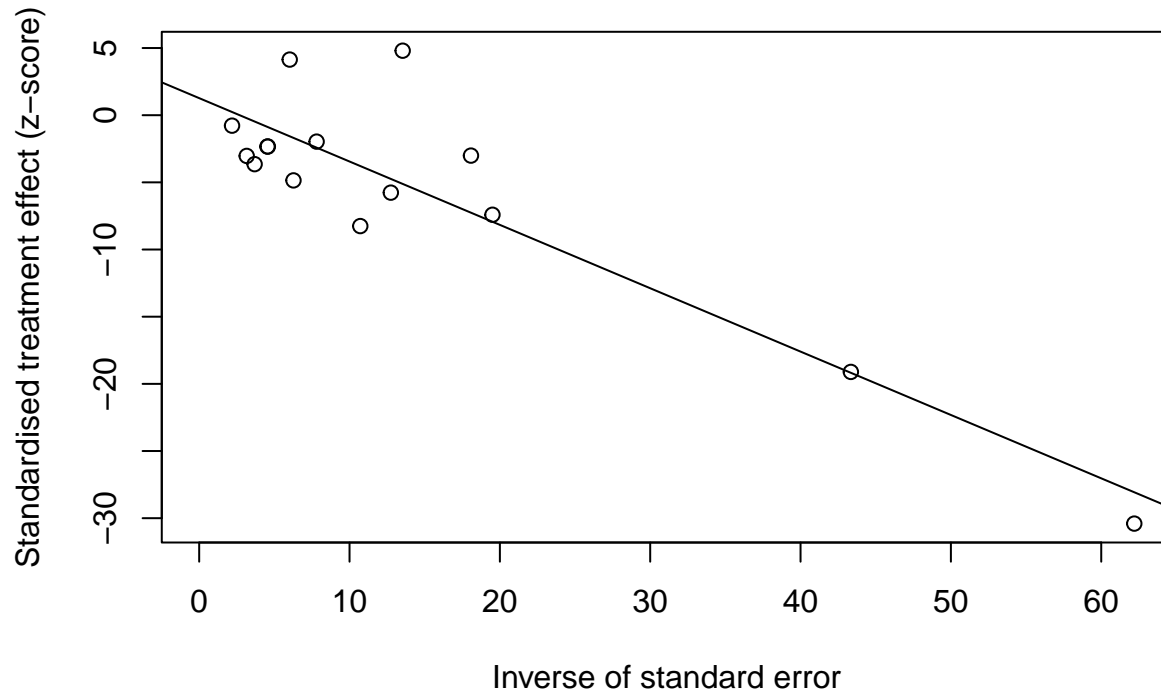
```



```
# Assessing and accounting for small-study effects
# Funnel Plot
funnel(infec.model)
```



```
# Test for funnel plot asymmetry using the linear regression method
metabias(infec.model, method.bias = "linreg", plotit = TRUE)
```



```
## Linear regression test of funnel plot asymmetry
##
## Test result: t = 0.91, df = 13, p-value = 0.3782
## Bias estimate: 1.2701 (SE = 1.3921)
##
## Details:
## - multiplicative residual heterogeneity variance (tau^2 = 16.0932)
## - predictor: standard error
## - weight: inverse variance
## - reference: Egger et al. (1997), BMJ
```

Based on the meta-analysis of 15 studies with a total of 5,459,570 vaccinations and 32,190 infection events, the overall relative risk (RR) for intervention 1 compared to intervention 2 was found to be 0.6487 (95% CI: 0.6338 to 0.6640) using the fixed-effect model, and 0.6849 (95% CI: 0.5469 to 0.8577) using the random-effects model. Both models indicate a statistically significant reduction in risk with intervention 1 ($p < 0.0001$ for the fixed-effect model and $p = 0.0010$ for the random-effects model). The high I^2 value of 93.7% suggests substantial heterogeneity among the included studies. The test for funnel plot asymmetry (Egger's test) did not indicate significant publication bias ($p = 0.3782$). These results suggest that intervention 1 is associated with a lower risk of infection compared to intervention 2.

2. Symptomatic Infection

```
# Read in the symptomatic infection data-set
# with the first row containing variable names
sym.infec.data <- read.csv("../raw_data/symptomatic_infection_data.csv", header = TRUE)
head(sym.infec.data)
```

```
##   author_id      age interv_id1 interv_id2      N1  n1      N2  n2
## 1         1 50+ years          1          2 158993  17 236693  68
## 2         2 60+ years          1          2   1831 474   2139 719
## 3         3 65+ years          1          2 100751 654   83998 644
## 4         4 60+ years          1          2    263  20    2949 375
## 5         5 55+ years          1          2   4105 463   7119 1174
```

```
# Fixed Effect & Random Effect Meta-Analysis Models
```

```
sym.infec.model <- metabin(event.e = n1, n.e = N1,
                           event.c = n2, n.c = N2,
                           data = sym.infec.data, sm = "RR", method = "I",
                           studlab = author_id, label.e = "Intervention 1", label.c = "Intervention 2")
```

```
# Summary of the meta-analysis model
```

```
summary(sym.infec.model)
```

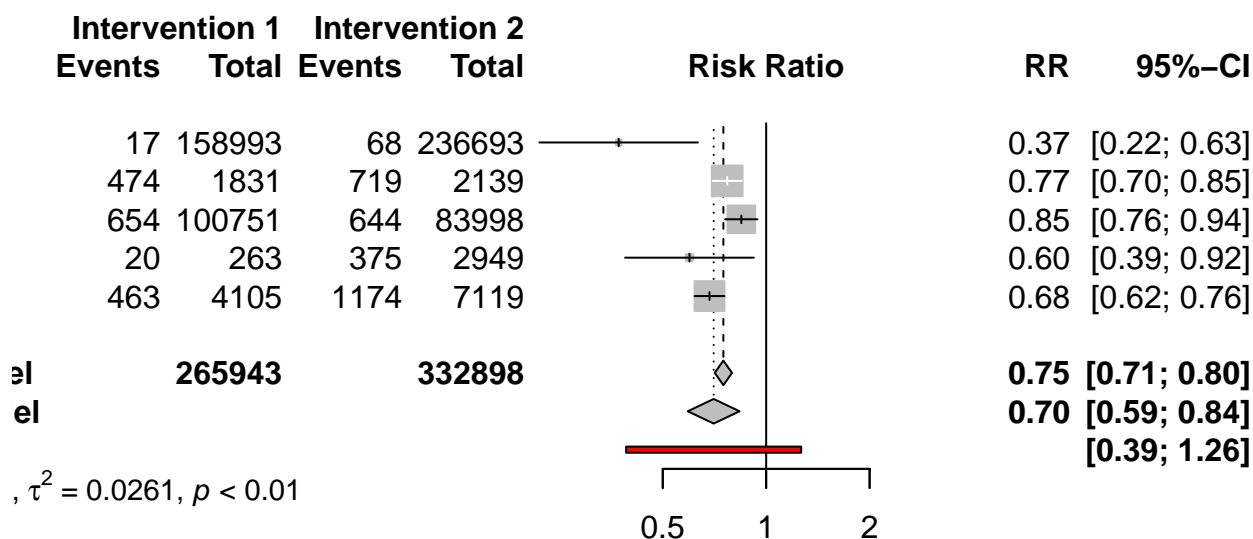
```
##           RR           95%-CI %W(common) %W(random)
## 1 0.3722 [0.2188; 0.6332]      1.2        7.9
## 2 0.7701 [0.6984; 0.8492]     35.2       27.4
## 3 0.8467 [0.7597; 0.9436]     28.6       26.9
## 4 0.5980 [0.3883; 0.9209]      1.8       10.5
## 5 0.6839 [0.6186; 0.7562]     33.3       27.3
##
## Number of studies: k = 5
## Number of observations: o = 598841 (o.e = 265943, o.c = 332898)
## Number of events: e = 4608
##
##           RR           95%-CI      z  p-value
## Common effect model 0.7506 [0.7084; 0.7954] -9.70 < 0.0001
## Random effects model 0.7033 [0.5912; 0.8367] -3.97 < 0.0001
##
## Quantifying heterogeneity:
## tau^2 = 0.0261 [0.0033; 0.8051]; tau = 0.1617 [0.0578; 0.8973]
## I^2 = 75.1% [38.6%; 89.9%]; H = 2.00 [1.28; 3.14]
##
## Test of heterogeneity:
##      Q d.f. p-value
## 16.06   4  0.0029
##
## Details on meta-analytical method:
## - Inverse variance method
## - Restricted maximum-likelihood estimator for tau^2
## - Q-Profile method for confidence interval of tau^2 and tau
```

```
# Forest Plot
```

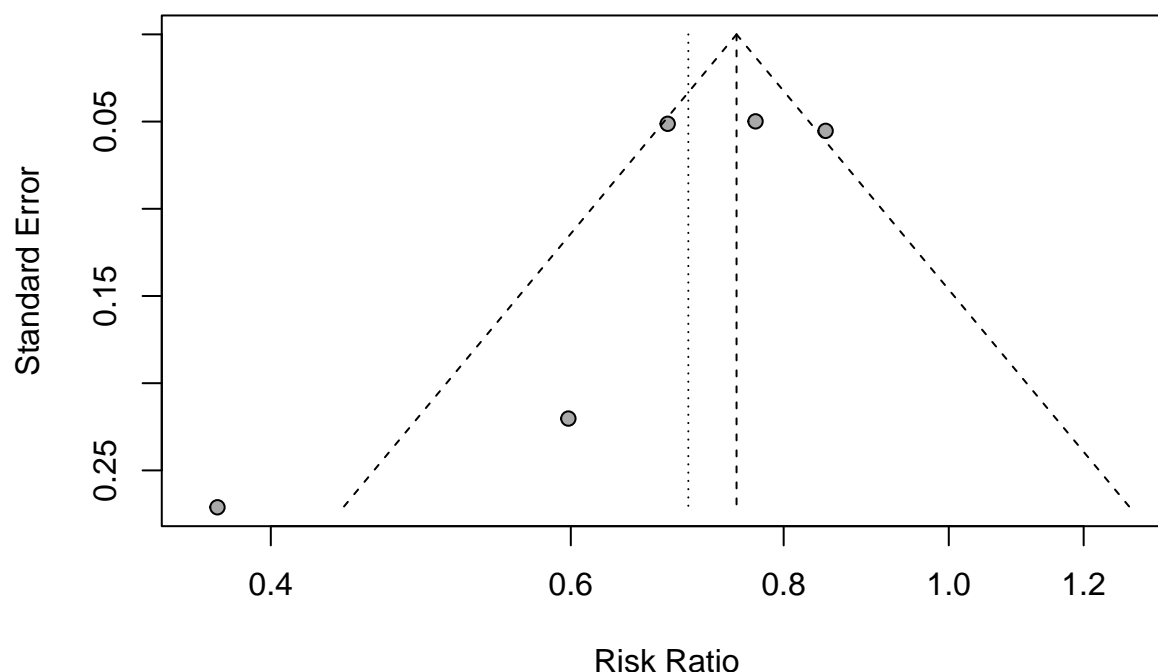
```
# Showing relative risk of intervention 1 relative to intervention 2
```

```
# with corresponding 95% confidence intervals
```

```
forest(sym.infec.model, studlab = author_id, prediction = TRUE)
```



```
# Assessing and accounting for small-study effects
# Funnel Plot
funnel(sym.infec.model)
```



The meta-analysis of five studies on symptomatic infection, comprising a total of 598,841 observations and 4,608 events, shows a statistically significant reduction in risk for intervention 1 compared to intervention 2. The fixed-effect model indicates a relative risk (RR) of 0.7506 (95% CI: 0.7084 to 0.7954, $p < 0.0001$), while the random-effects model shows a slightly lower RR of 0.7033 (95% CI: 0.5912 to 0.8367, $p < 0.0001$).

These results suggest that intervention 1 is associated with a lower risk of symptomatic infection compared to intervention 2. However, there is substantial heterogeneity among the studies, as indicated by an I^2 value of 75.1%, which means that a significant proportion of the variability in effect estimates is due to heterogeneity rather than chance. The test for heterogeneity ($Q = 16.06$, $df = 4$, $p = 0.0029$) also confirms the presence of significant heterogeneity.

Overall, intervention 1 shows a consistent benefit in reducing symptomatic infection across the studies, but the substantial heterogeneity should be considered in interpreting these results.

3. Severe Infection

```
# Read in the severe infection data-set
# with the first row containing variable names
sev.infec.data <- read.csv("../raw_data/severe_infection_data.csv", header = TRUE)
head(sev.infec.data)
```

```
##   author_id      age interv_id1 interv_id2      N1  n1      N2  n2
## 1         1 50+ years          1          2 158993  2  236693  21
## 2         2 60+ years          1          2   1518 161   1638 218
## 3         3 65+ years          1          2 100751 14   83998  22
## 4         4 55+ years          1          2 402102 183  453015 296
## 5         5 50+ years          1          2 1614377 646 1793698 1195
## 6         6 55+ years          1          2  116251 24   845906 273
```

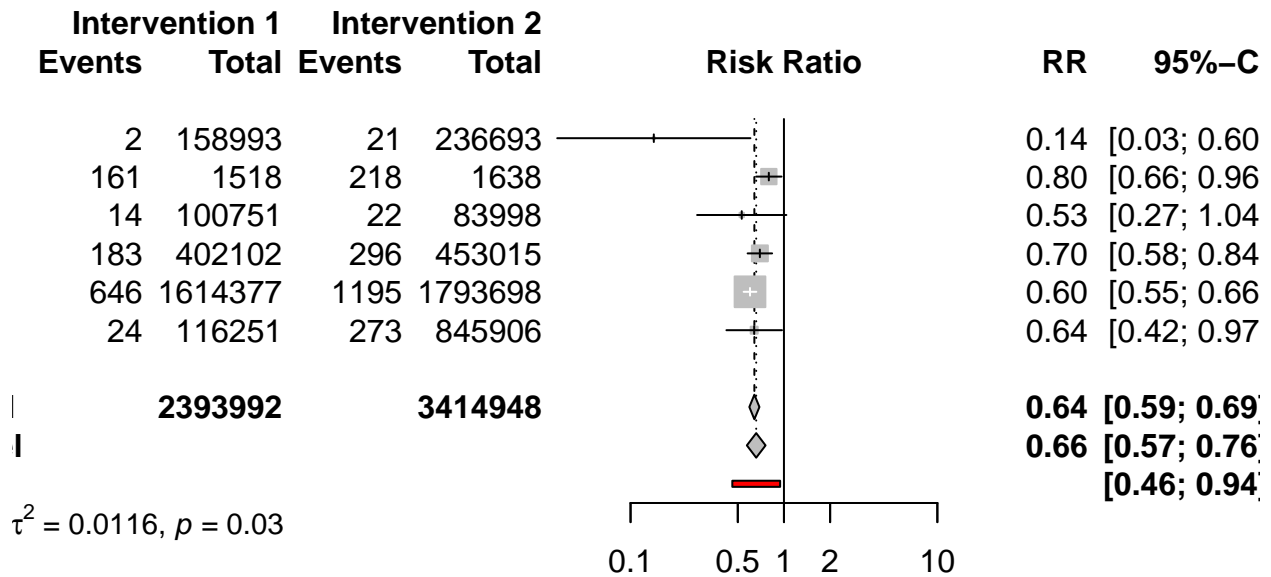
```
# Fixed Effect & Random Effect Meta-Analysis Models
sev.infec.model <- metabin(event.e = n1, n.e = N1,
                           event.c = n2, n.c = N2,
                           data = sev.infec.data, sm = "RR", method = "I",
                           studlab = author_id, label.e = "Intervention 1", label.c = "Intervention 2")
summary(sev.infec.model)
```

```
##           RR           95%-CI %W(common) %W(random)
## 1 0.1418 [0.0332; 0.6047]      0.3         0.9
## 2 0.7969 [0.6581; 0.9649]     15.7        24.3
## 3 0.5305 [0.2715; 1.0368]      1.3         4.0
## 4 0.6965 [0.5793; 0.8374]     16.9        25.1
## 5 0.6006 [0.5458; 0.6609]     62.6        36.7
## 6 0.6397 [0.4215; 0.9709]      3.3         9.0
##
## Number of studies: k = 6
## Number of observations: o = 5808940 (o.e = 2393992, o.c = 3414948)
## Number of events: e = 3055
##
##           RR           95%-CI      z  p-value
## Common effect model 0.6415 [0.5947; 0.6920] -11.49 < 0.0001
## Random effects model 0.6594 [0.5731; 0.7587]  -5.82 < 0.0001
##
## Quantifying heterogeneity:
## tau^2 = 0.0116 [0.0000; 1.9887]; tau = 0.1076 [0.0000; 1.4102]
## I^2 = 58.3% [0.0%; 83.1%]; H = 1.55 [1.00; 2.43]
##
## Test of heterogeneity:
##      Q d.f. p-value
## 11.99   5 0.0349
##
## Details on meta-analytical method:
## - Inverse variance method
```

```
## - Restricted maximum-likelihood estimator for tau^2
## - Q-Profile method for confidence interval of tau^2 and tau
```

```
# Forest Plot
```

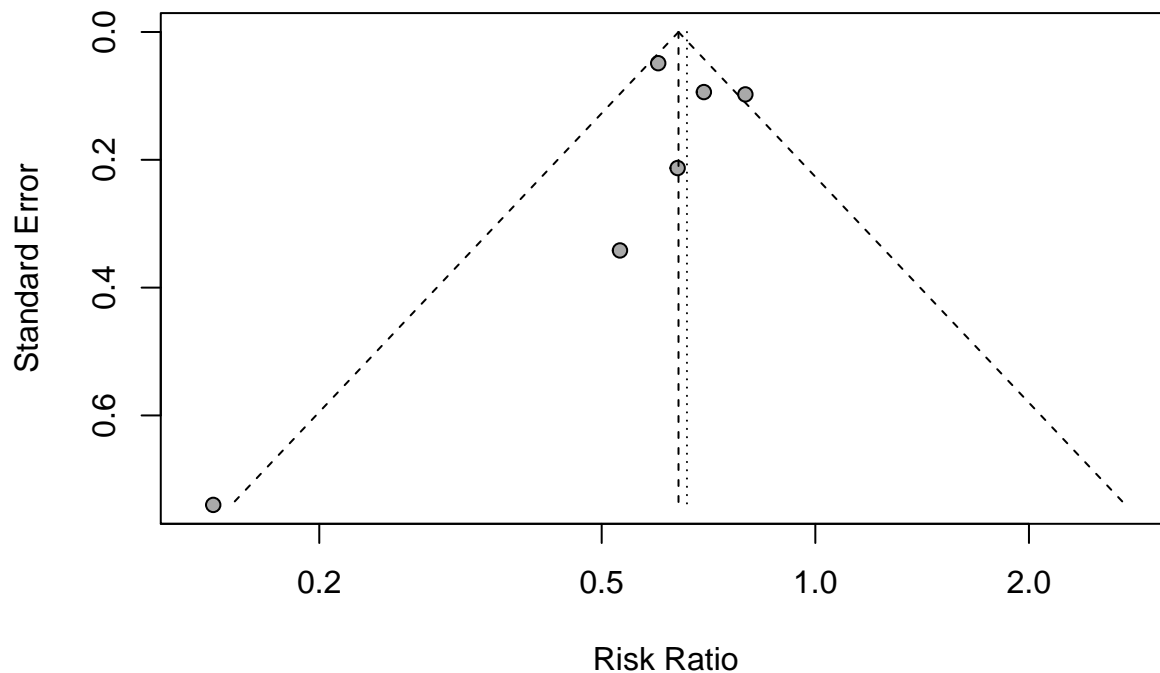
```
forest(sev.infec.model, studlab = author_id, prediction = TRUE)
```



```
# Assessing and accounting for small-study effects
```

```
# Funnel Plot
```

```
funnel(sev.infec.model)
```



The meta-analysis of six studies on severe infection, comprising a total of 5,808,940 observations and 3,055 events, indicates that intervention 1 is associated with a statistically significant reduction in the risk of severe infection compared to intervention 2. The fixed-effect model shows a relative risk (RR) of 0.6415 (95% CI:

0.5947 to 0.6920, $p < 0.0001$), while the random-effects model indicates an RR of 0.6594 (95% CI: 0.5731 to 0.7587, $p < 0.0001$).

These results suggest that intervention 1 significantly reduces the risk of severe infection. However, there is moderate heterogeneity among the included studies, as reflected by an I^2 value of 58.3%, indicating that approximately 58.3% of the variability in effect estimates is due to heterogeneity rather than chance. The heterogeneity test ($Q = 11.99$, $df = 5$, $p = 0.0349$) also supports the presence of significant heterogeneity.

4. Hospitalisation

```
# Read in the hospitalisation data-set
hos.data <- read.csv("../raw_data/hospitalisation_data.csv", header = TRUE)
head(hos.data)
```

##	author_id	age	interv_id1	interv_id2	N1	n1	N2	n2
## 1	1	50+ years	1	2	158993	2	236693	21
## 2	2	65+ years	1	2	45285	41	45285	69
## 3	3	55+ years	1	2	402102	183	453015	296
## 4	4	50+ years	1	2	1614377	646	1793698	1195

```
# Fixed Effect & Random Effect Meta-Analysis Models
# Using risk ratio (RR) as the measure
hos.model <- metabin(event.e = n1, n.e = N1,
                     event.c = n2, n.c = N2,
                     data = hos.data, sm = "RR", method = "I",
                     studlab = author_id, label.e = "Intervention 1", label.c = "Intervention 2")
summary(hos.model)
```

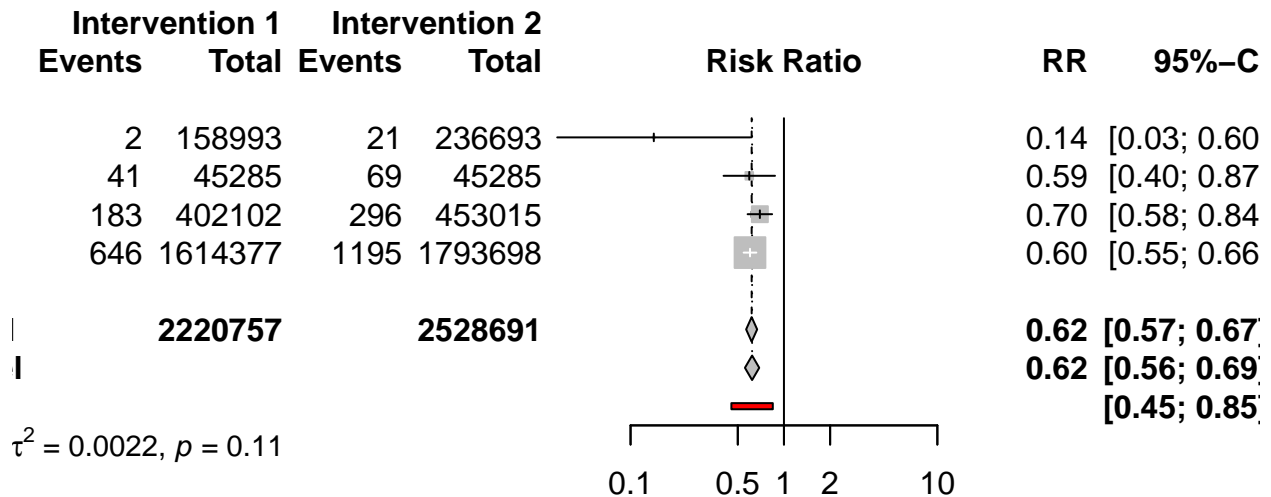
##	RR	95%-CI	%W(common)	%W(random)
## 1	0.1418	[0.0332; 0.6047]	0.3	0.5
## 2	0.5942	[0.4038; 0.8744]	4.6	7.3
## 3	0.6965	[0.5793; 0.8374]	20.2	27.1
## 4	0.6006	[0.5458; 0.6609]	74.9	65.1

```
##
## Number of studies: k = 4
## Number of observations: o = 4749448 (o.e = 2220757, o.c = 2528691)
## Number of events: e = 2453
##
##
## RR          95%-CI      z  p-value
## Common effect model 0.6157 [0.5667; 0.6688] -11.48 < 0.0001
## Random effects model 0.6198 [0.5568; 0.6901] -8.74 < 0.0001
##
## Quantifying heterogeneity:
## tau^2 = 0.0022 [0.0000; 7.3734]; tau = 0.0472 [0.0000; 2.7154]
## I^2 = 49.6% [0.0%; 83.3%]; H = 1.41 [1.00; 2.45]
##
## Test of heterogeneity:
## Q d.f. p-value
## 5.95 3 0.1141
##
## Details on meta-analytical method:
## - Inverse variance method
```

```
## - Restricted maximum-likelihood estimator for tau^2
## - Q-Profile method for confidence interval of tau^2 and tau
```

```
# Forest Plot
```

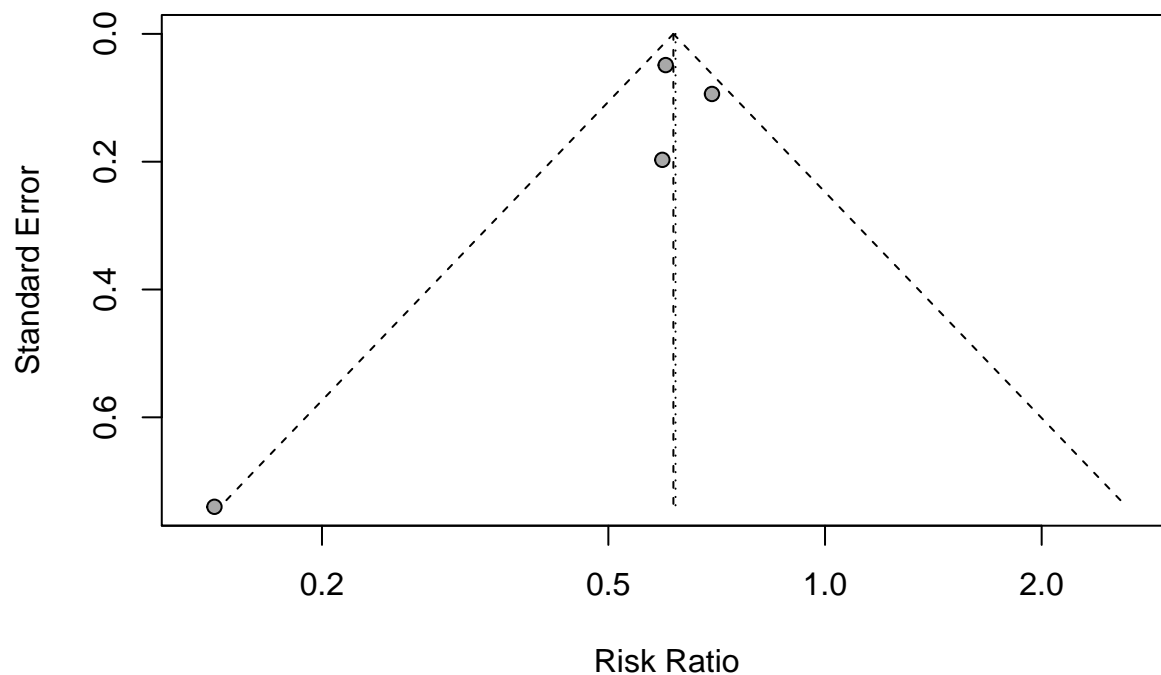
```
forest(hos.model, studlab = author_id, prediction = TRUE)
```



```
# Assessing and accounting for small-study effects
```

```
# Funnel Plot
```

```
funnel(hos.model)
```



The meta-analysis of four studies on hospitalisation, comprising a total of 4,749,448 observations and 2,453 events, shows a statistically significant reduction in the risk of hospitalisation for intervention 1 compared to intervention 2. The fixed-effect model indicates a relative risk (RR) of 0.6157 (95% CI: 0.5667 to 0.6688, $p < 0.0001$), while the random-effects model shows an RR of 0.6198 (95% CI: 0.5568 to 0.6901, $p < 0.0001$).

These findings suggest that intervention 1 significantly reduces the risk of hospitalisation. The I^2 value of 49.6% indicates moderate heterogeneity, meaning that nearly half of the variability in effect estimates is due

to heterogeneity rather than chance. However, the test for heterogeneity ($Q = 5.95$, $df = 3$, $p = 0.1141$) is not statistically significant, suggesting that the observed heterogeneity might not be substantial.

5. Death

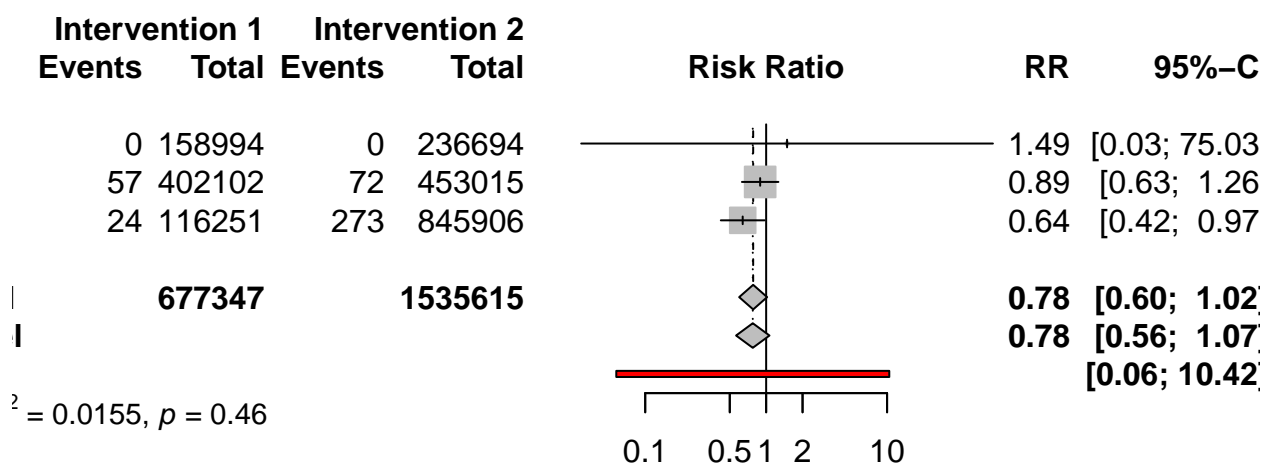
```
# Read in the death data-set
death.data <- read.csv("../raw_data/death_data.csv", header = TRUE)
head(death.data)

##   author_id      age interv_id1 interv_id2      N1      n1      N2      n2
## 1         1 50+ years          1          2 158994  0.5 236694   0.5
## 2         2 55+ years          1          2 402102 57.0 453015  72.0
## 3         3 55+ years          1          2 116251 24.0 845906 273.0

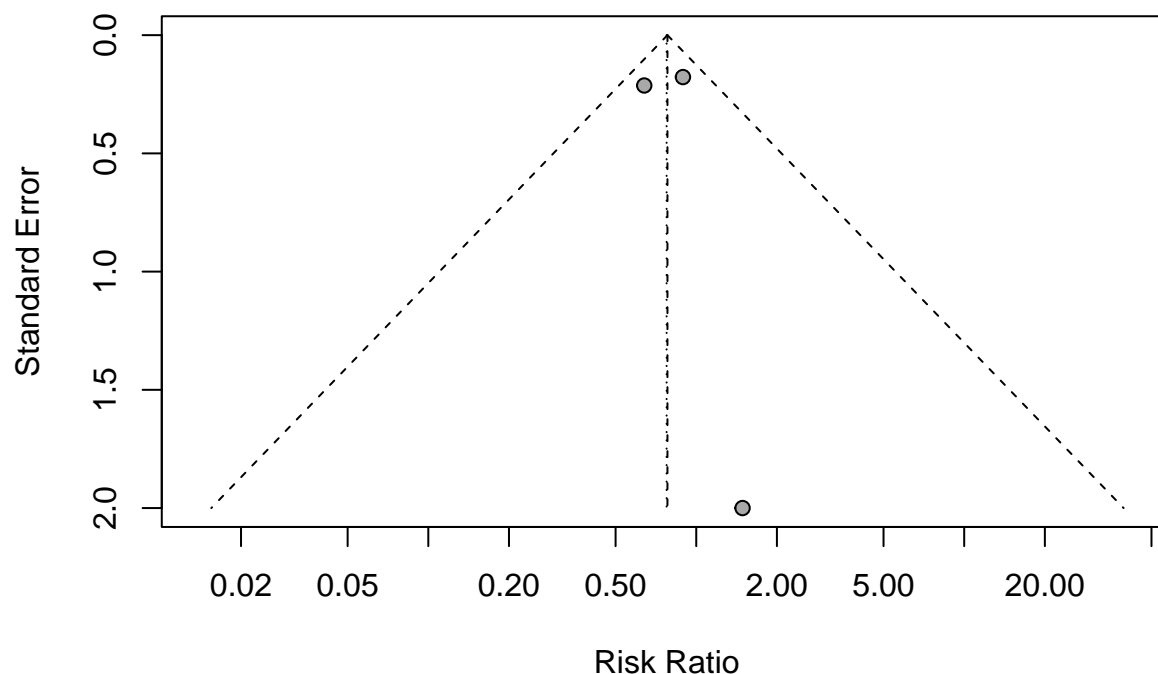
# Fixed Effect & Random Effect Meta-Analysis Models
# Using risk ratio (RR) as the measure
death.model <- metabin(event.e = n1, n.e = N1,
                       event.c = n2, n.c = N2,
                       data = death.data, sm = "RR", method = "I",
                       studlab = author_id, label.e = "Intervention 1", label.c = "Intervention 2")
summary(death.model)

##           RR           95%-CI %W(common) %W(random)
## 1 1.4887 [0.0295; 75.0252]         0.5         0.7
## 2 0.8919 [0.6301; 1.2625]        58.8        56.1
## 3 0.6397 [0.4215; 0.9709]        40.8        43.3
##
## Number of studies: k = 3
## Number of observations: o = 2212962 (o.e = 677347, o.c = 1535615)
## Number of events: e = 427
##
##           RR           95%-CI      z p-value
## Common effect model 0.7807 [0.5982; 1.0191] -1.82 0.0686
## Random effects model 0.7750 [0.5639; 1.0652] -1.57 0.1162
##
## Quantifying heterogeneity:
## tau^2 = 0.0155 [0.0000; 5.0358]; tau = 0.1246 [0.0000; 2.2441]
## I^2 = 0.0% [0.0%; 89.6%]; H = 1.00 [1.00; 3.10]
##
## Test of heterogeneity:
##      Q d.f. p-value
## 1.54   2 0.4621
##
## Details on meta-analytical method:
## - Inverse variance method
## - Restricted maximum-likelihood estimator for tau^2
## - Q-Profile method for confidence interval of tau^2 and tau

# Forest Plot
forest(death.model, studlab = author_id, prediction = TRUE)
```



```
# Assessing and accounting for small-study effects
# Funnel Plot
funnel(death.model)
```



The meta-analysis of three studies on death, comprising a total of 2,212,962 observations and 427 events, suggests that intervention 1 may reduce the risk of death compared to intervention 2, but the results are not statistically significant. The fixed-effect model shows a relative risk (RR) of 0.7807 (95% CI: 0.5982 to 1.0191, $p = 0.0686$), and the random-effects model indicates an RR of 0.7750 (95% CI: 0.5639 to 1.0652, $p = 0.1162$).

These results suggest a potential reduction in the risk of death with intervention 1, but the evidence is not strong enough to conclude a significant effect, as indicated by the p -values above 0.05 in both models. The I^2 value of 0.0% indicates no observed heterogeneity among the included studies, suggesting that the variability in effect estimates is due to chance rather than differences between studies. This is further supported by the non-significant test for heterogeneity ($Q = 1.54, df = 2, p = 0.4621$).

Conclusion

According to the meta-analyses conducted on the five outcomes separately, intervention 1 consistently demonstrates a significant benefit over intervention 2 in reducing the risks of infection, symptomatic infection, severe infection, and hospitalisation. However, the data on mortality does not show a statistically significant difference between the two interventions, indicating that while intervention 1 may have a positive impact, further research is needed to establish a definitive conclusion regarding its effect on reducing deaths.

In summary, intervention 1 appears to be more effective than intervention 2. The heterogeneity observed in some analyses suggests that further studies could help clarify the extent of these benefits and provide more robust conclusions.