

LDA_IGA-NTX

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§ 1

Load required packages:

```
#if (Sys.info()["sysname"] != "Darwin"){  
#  default_theme <- theme_minimal() +  
#  theme(text = element_text(family = "Decima WE", size = 15)) +  
#  theme(panel.grid.major = element_line(color = "grey", size = 0.3)) +  
#  theme(axis.line = element_line(color = "black", size = 0.4))  
#  
#} else {  
#  default_theme <- theme_minimal()  
#}  
default_theme <- theme_minimal()  
two_scale_fill <- scale_fill_manual(values=c("#69b3a2", "#404080"))
```

Read data using read_data.R scrip:

```
source("read_data.R")
```

§ 2 EDA

```
follow_up <- years(10)
data_iga$follow_up_truncated <- pmin(data_iga$`T-dls`, data_iga$`T-date` + follow_up)

data_iga$mismatch_sum <- as.numeric(as.character(data_iga$`mm-A`)) + as.numeric(as.character(data_iga$`T-dls` - data_iga$`T-date`))

data_iga_pos <- data_iga[`biopsy proven recurrence (0=no, 1=yes)` == 1]
data_iga_neg <- data_iga[`biopsy proven recurrence (0=no, 1=yes)` == 0]
```

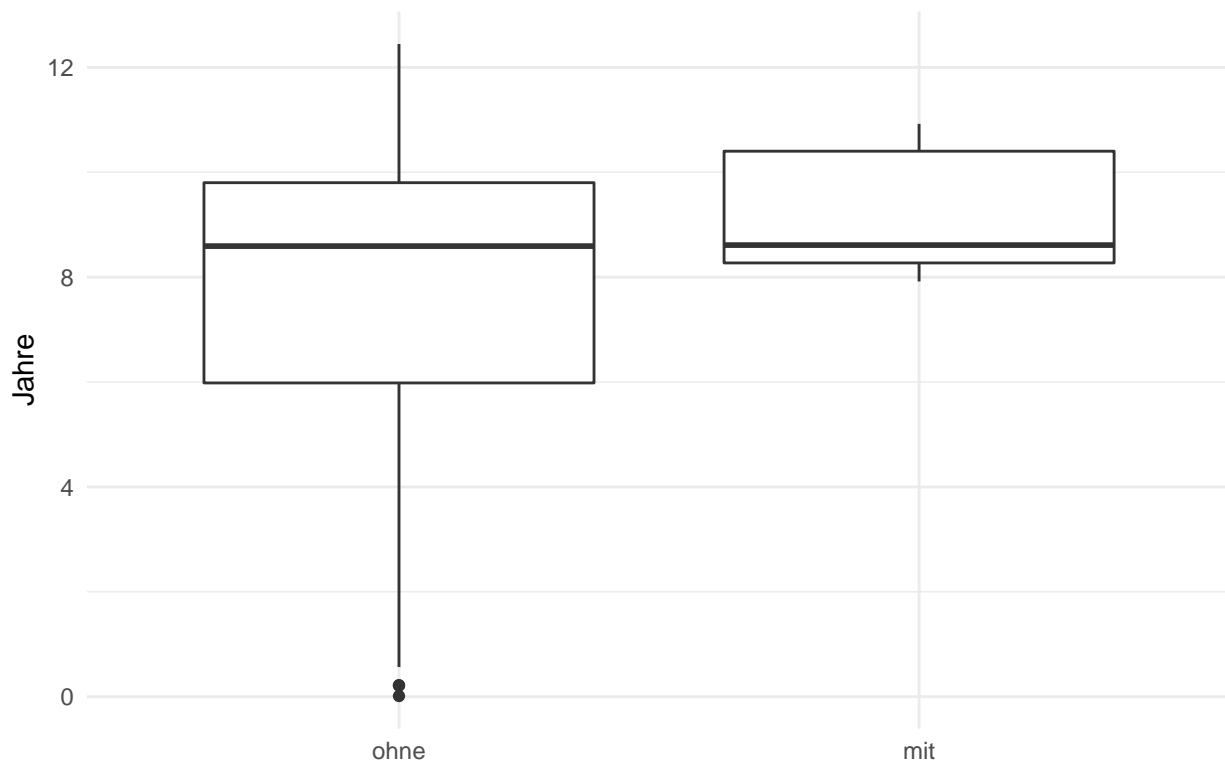
§ 2.1 IGA

```
# follow-up mean / median
tbl_iga_follow_up_mean <- data.frame(
  iga_all = mean((interval(data_iga$`T-date`, data_iga$`T-dls`) / years(1))),
  iga_pos = mean((interval(data_iga_pos$`T-date`, data_iga_pos$`T-dls`) / years(1))),
  iga_neg = mean((interval(data_iga_neg$`T-date`, data_iga_neg$`T-dls`) / years(1)))
)

tbl_iga_follow_up_median <- data.frame(
  iga_all = median((interval(data_iga$`T-date`, data_iga$`T-dls`) / years(1))),
  iga_pos = median((interval(data_iga_pos$`T-date`, data_iga_pos$`T-dls`) / years(1))),
  iga_neg = median((interval(data_iga_neg$`T-date`, data_iga_neg$`T-dls`) / years(1)))
)

## ToDo: All IGA
ggplot() +
  geom_boxplot(aes(x = `biopsy proven recurrence (0=no, 1=yes)`, y = interval(data_iga$`T-date`, data_iga$`T-dls`))) +
  default_theme +
  ggtitle("Boxplot: Follow Up Period") +
  ylab("Jahre") +
  xlab("") +
  scale_x_discrete(labels = c("ohne", "mit"))
```

Boxplot: Follow Up Period



```
# follow_up mean/median (truncated)
tbl_iga_follow_up_mean_truncated <- data.frame(
  iga_all = mean((interval(data_iga`T-date`, data_iga$follow_up_truncated) / years(1))),
  iga_pos = mean((interval(data_iga_pos`T-date`, data_iga_pos$follow_up_truncated) / years(1))),
  iga_neg = mean((interval(data_iga_neg`T-date`, data_iga_neg$follow_up_truncated) / years(1)))
)

tbl_iga_follow_up_median_truncated <- data.frame(
  iga_all = median((interval(data_iga`T-date`, data_iga$follow_up_truncated) / years(1))),
  iga_pos = median((interval(data_iga_pos`T-date`, data_iga_pos$follow_up_truncated) / years(1))),
  iga_neg = median((interval(data_iga_neg`T-date`, data_iga_neg$follow_up_truncated) / years(1)))
)

# patient death within follow up
tbl_iga_pat_death <- data.frame(
  iga_all = nrow(data_iga[(`T-dls` <= (`T-date` + follow_up)) & `Pat death (0=alive, 1= dead)` == 1]),
  iga_pos = nrow(data_iga_pos[(`T-dls` <= (`T-date` + follow_up)) & `Pat death (0=alive, 1= dead)` == 1]),
  iga_neg = nrow(data_iga_neg[(`T-dls` <= (`T-date` + follow_up)) & `Pat death (0=alive, 1= dead)` == 1])
)

# patient drop out within follow up
tbl_iga_pat_drop <- data.frame(
  iga_all = nrow(data_iga[(`T-dls` <= (`T-date` + follow_up)) & `Pat death (0=alive, 1= dead)` == 0]),
  iga_pos = nrow(data_iga_pos[(`T-dls` <= (`T-date` + follow_up)) & `Pat death (0=alive, 1= dead)` == 0]),
  iga_neg = nrow(data_iga_neg[(`T-dls` <= (`T-date` + follow_up)) & `Pat death (0=alive, 1= dead)` == 0])
)

# patients with graft loss
tbl_iga_graft_loss <- data.frame(
  iga_all = nrow(data_iga[`graft loss (0=functionial, 1=loss)` == 1]),
```

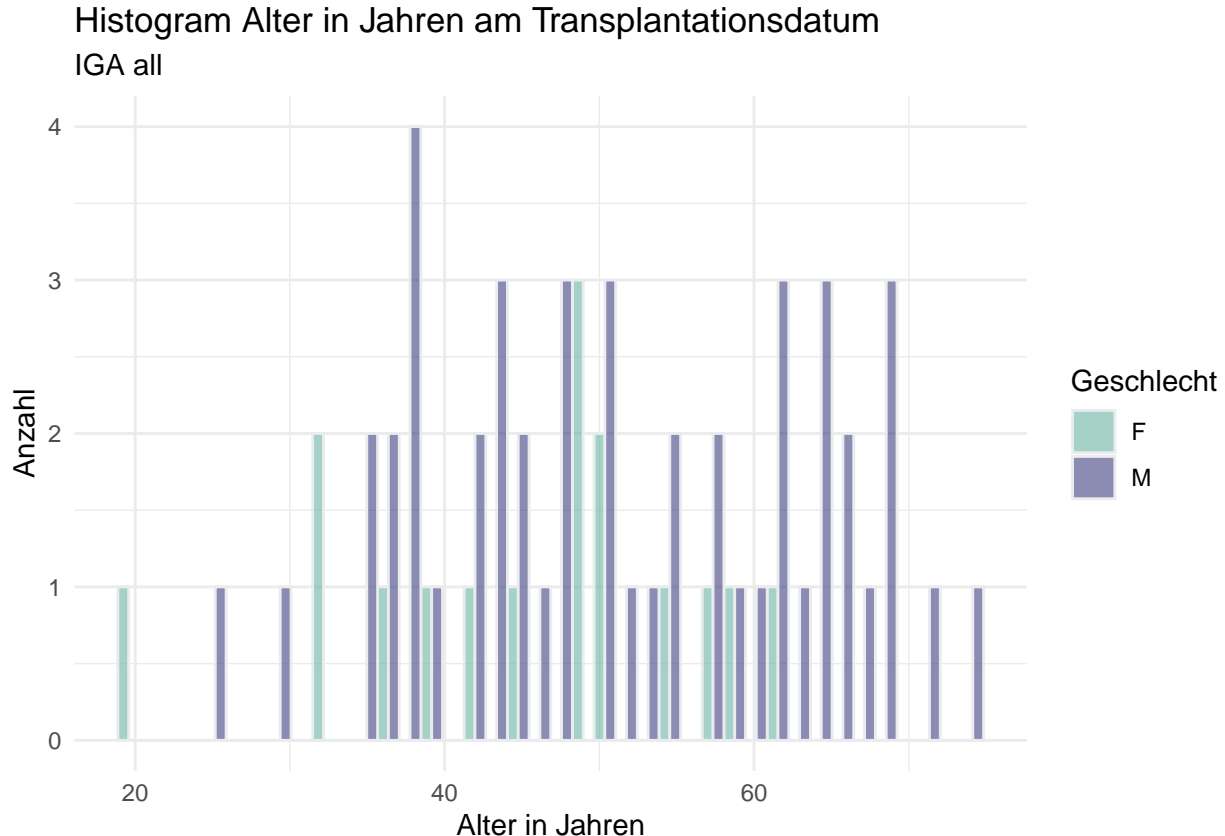
```

iga_pos = nrow(data_iga_pos[`graft loss` (0=functional, 1=loss) == 1]),
iga_neg = nrow(data_iga_neg[`graft loss` (0=functional, 1=loss) == 1])
)

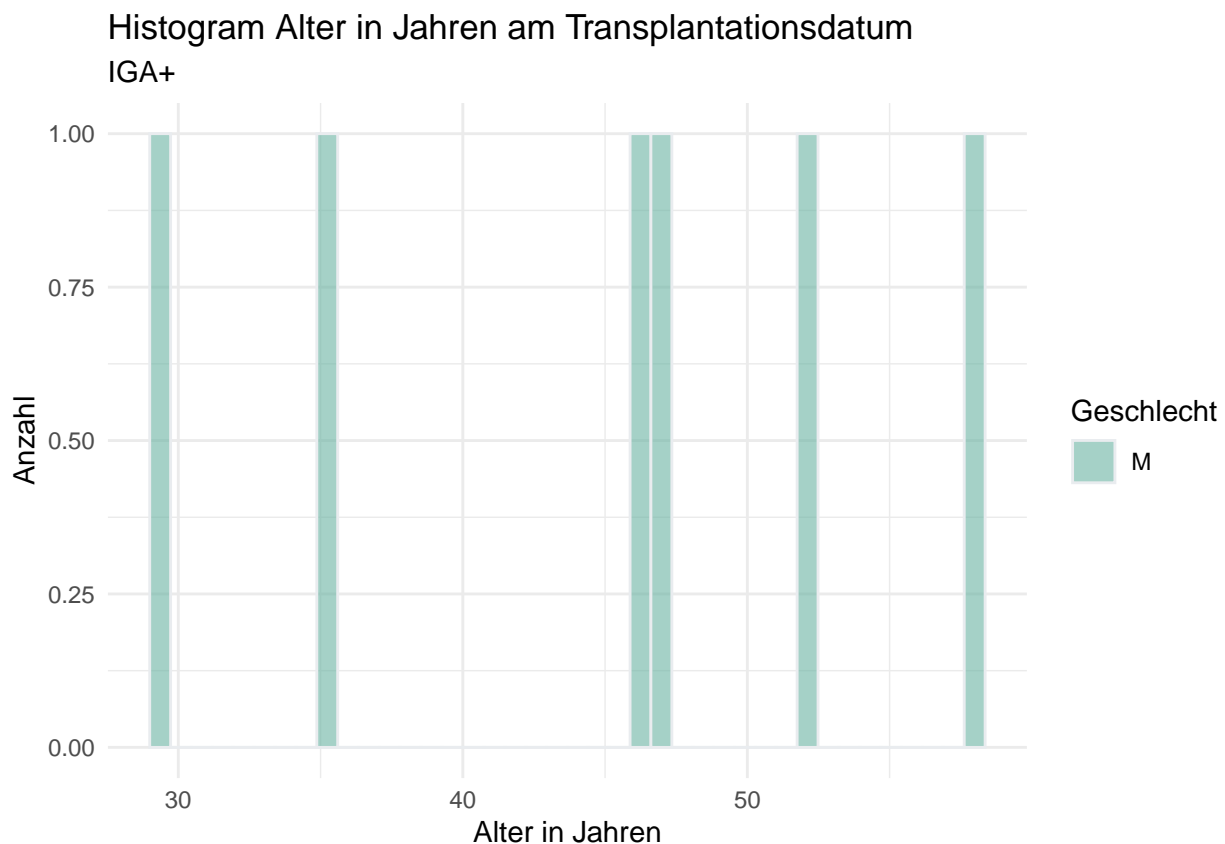
# patients with graft loss within follow up period
tbl_iga_graft_loss_follow_up <- data.frame(
  iga_all = nrow(data_iga[`graft loss date` < `T-date` + follow_up]),
  iga_pos = nrow(data_iga_pos[`graft loss date` < `T-date` + follow_up]),
  iga_neg = nrow(data_iga_neg[`graft loss date` < `T-date` + follow_up])
)

# ToDo: median IQR
# age patients (yrs.)
ggplot(data = data_iga) +
  geom_histogram(mapping = aes(x = interval(`Date of birth`, `T-date`) / years(1),
                                fill = `R-sex`),
                color="#e9ecef",
                alpha=0.6,
                bins = 40,
                position = "dodge") +
  two_scale_fill +
  ylab("Anzahl") +
  xlab("Alter in Jahren") +
  ggtitle("Histogram Alter in Jahren am Transplantationsdatum",
          subtitle = "IGA all") +
  labs(fill = "Geschlecht") +
  default_theme

```

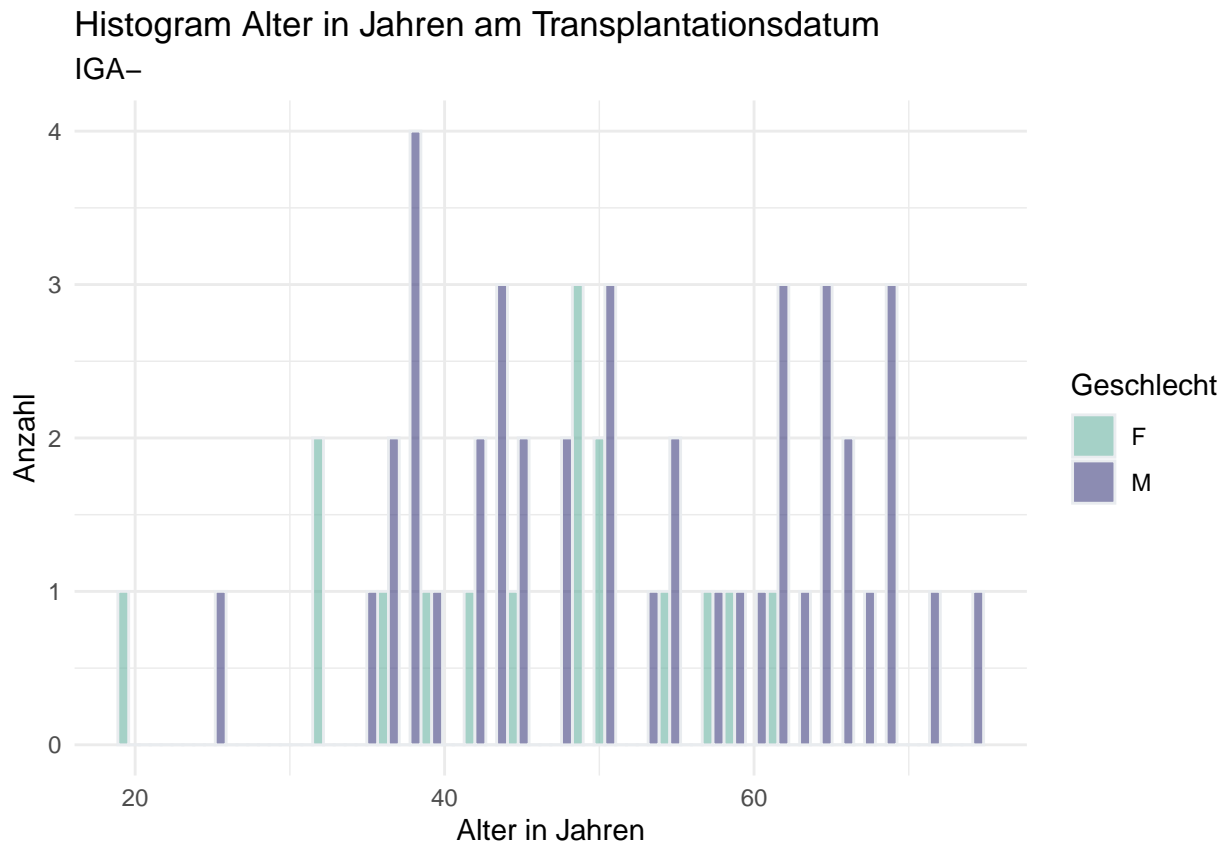


```
ggplot(data = data_iga_pos) +
  geom_histogram(mapping = aes(x = interval(`Date of birth`, `T-date`) / years(1),
                                     fill = `R-sex`),
                color="#e9ecef",
                alpha=0.6,
                bins = 40,
                position = "dodge") +
  two_scale_fill +
  ylab("Anzahl") +
  xlab("Alter in Jahren") +
  ggtitle("Histogram Alter in Jahren am Transplantationsdatum",
          subtitle = "IGA+") +
  labs(fill = "Geschlecht") +
  default_theme
```



```
ggplot(data = data_iga_neg) +
  geom_histogram(mapping = aes(x = interval(`Date of birth`, `T-date`) / years(1),
                                     fill = `R-sex`),
                color="#e9ecef",
                alpha=0.6,
                bins = 40,
                position = "dodge") +
  two_scale_fill +
  ylab("Anzahl") +
  xlab("Alter in Jahren") +
  ggtitle("Histogram Alter in Jahren am Transplantationsdatum",
          subtitle = "IGA-") +
  labs(fill = "Geschlecht") +
  default_theme
```

```
labs(fill = "Geschlecht") +
default_theme
```



```
## IQR
tbl_iga_iqr_age <- data.frame(
  iga_all = IQR(data_iga$R_age_Tdate),
  iga_po = IQR(data_iga_pos$R_age_Tdate),
  iga_neg = IQR(data_iga_neg$R_age_Tdate)
)
```

```
## boxpot

p1 <- ggplot(data = data_iga) +
  geom_boxplot(aes(x = R_age_Tdate)) +
  coord_flip() +
  ggtitle("IGA (all)") +
  default_theme

p2 <- ggplot(data = data_iga_pos) +
  geom_boxplot(aes(x = R_age_Tdate)) +
  coord_flip() +
  ggtitle("IGA r+") +
  default_theme

p3 <- ggplot(data = data_iga_neg) +
  geom_boxplot(aes(x = R_age_Tdate)) +
  coord_flip() +
```

```

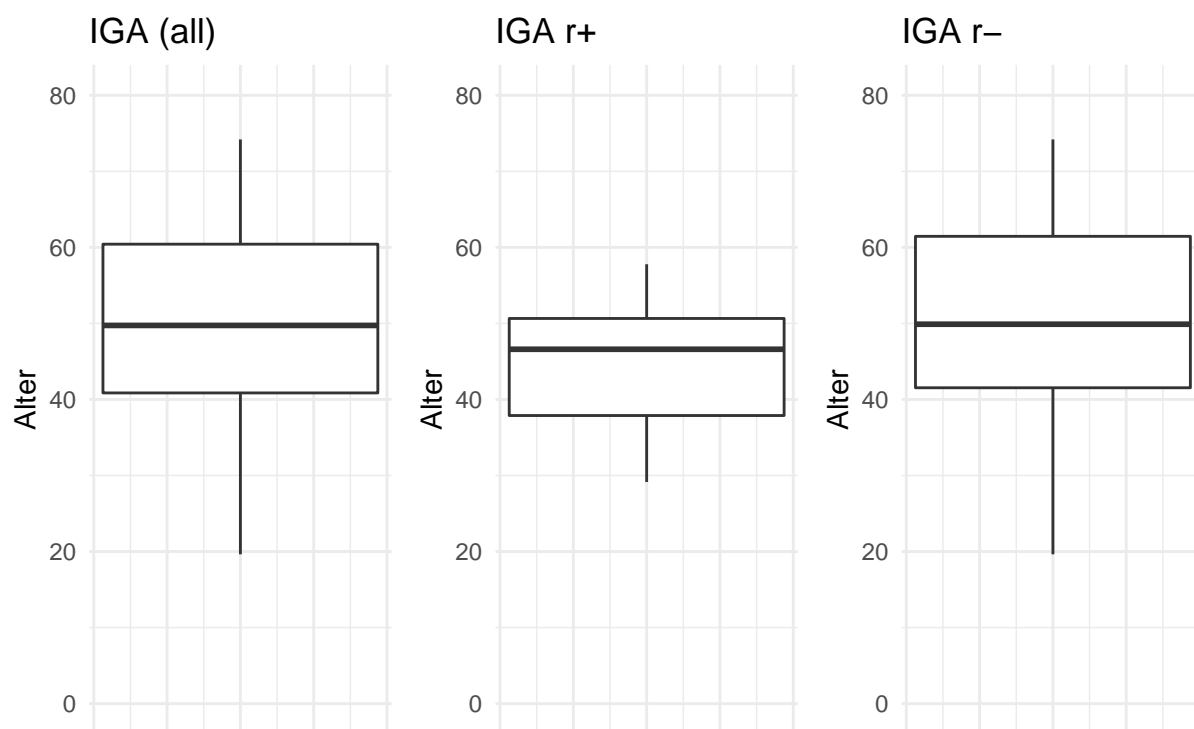
ggtitle("IGA r-") +
  default_theme

p1$labels$x <- p2$labels$x <- p3$labels$x <- ""
p1$labels$y <- p2$labels$y <- p3$labels$y <- ""

patch <- p1 | p2 | p3
patch <- patch + plot_annotation(
  title = "Boxplot Altersverteilung"
)
patch & xlim(0, 80) & theme(axis.ticks.x = element_blank(),
  axis.text.x = element_blank()) & xlab("Alter")

```

Boxplot Altersverteilung



```

# sex
tbl_iga_sex_abs <- data.frame(
  iga_all = summary(data_iga$`R-sex`),
  iga_pos = summary(data_iga_pos$`R-sex`),
  iga_neg = summary(data_iga_neg$`R-sex`)
)

p1 <- data.frame(group = rownames(tbl_iga_sex_abs),
  value = tbl_iga_sex_abs$iga_all) %>%
  ggplot(aes(x = "", y = value, fill = group)) +
  geom_bar(width = 1, stat = "identity") +
  coord_polar("y", start = 0) +
  guides(fill=guide_legend(title = "Geschlecht")) +
  theme_void() +
  two_scale_fill +
  ggtitle("iga all")

```

```

p2 <- data.frame(group = rownames(tbl_iga_sex_abs),
                 value = tbl_iga_sex_abs$iga_pos) %>%
  ggplot(aes(x = "", y = value, fill = group)) +
  geom_bar(width = 1, stat = "identity") +
  coord_polar("y", start = 0) +
  guides(fill=guide_legend(title = "Geschlecht")) +
  theme_void() +
  two_scale_fill +
  ggtitle("iga +")

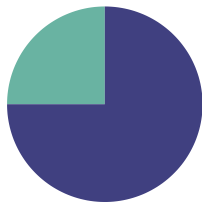
p3 <- data.frame(group = rownames(tbl_iga_sex_abs),
                 value = tbl_iga_sex_abs$iga_neg) %>%
  ggplot(aes(x = "", y = value, fill = group)) +
  geom_bar(width = 1, stat = "identity") +
  coord_polar("y", start = 0) +
  guides(fill=guide_legend(title = "Geschlecht")) +
  theme_void() +
  two_scale_fill +
  ggtitle("iga -")

patch <- p1 | p2 | p3
patch + plot_annotation(title = "Kreisdiagramme Geschlecht (absolut)")

```

Kreisdiagramme Geschlecht (absolut)

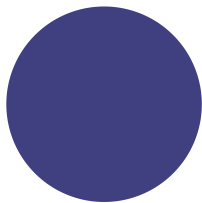
iga all



Geschlecht



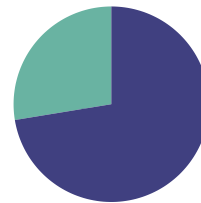
iga +



Geschlecht



iga -



Geschlecht



```

# sex percent
tbl_iga_sex_percent <- data.frame(
  iga_all = round(summary(data_iga$`R-sex`) / nrow(data_iga), 2),
  iga_pos = round(summary(data_iga_pos$`R-sex`) / nrow(data_iga_pos), 2),
  iga_neg = round(summary(data_iga_neg$`R-sex`) / nrow(data_iga_neg), 2)
)

```

```

# BMI (mean.)
ggplot(data = data_iga) +
  geom_histogram(mapping = aes(`D-weight` * (`D-height`)^2,
                              fill = `D-sex`),
                alpha = 0.6) +
  two_scale_fill +
  # median BMI total
  geom_vline(aes(xintercept = median(`D-weight` * (`D-height`)^2)),
            size = 1.0, color = "red", linetype = "dashed") +
  # median BMI F
  geom_vline(aes(xintercept = median(`D-weight` * (`D-height`)^2),
                data = data_iga[`R-sex` == "F"], linetype = "dashed") +
  # median BMI M

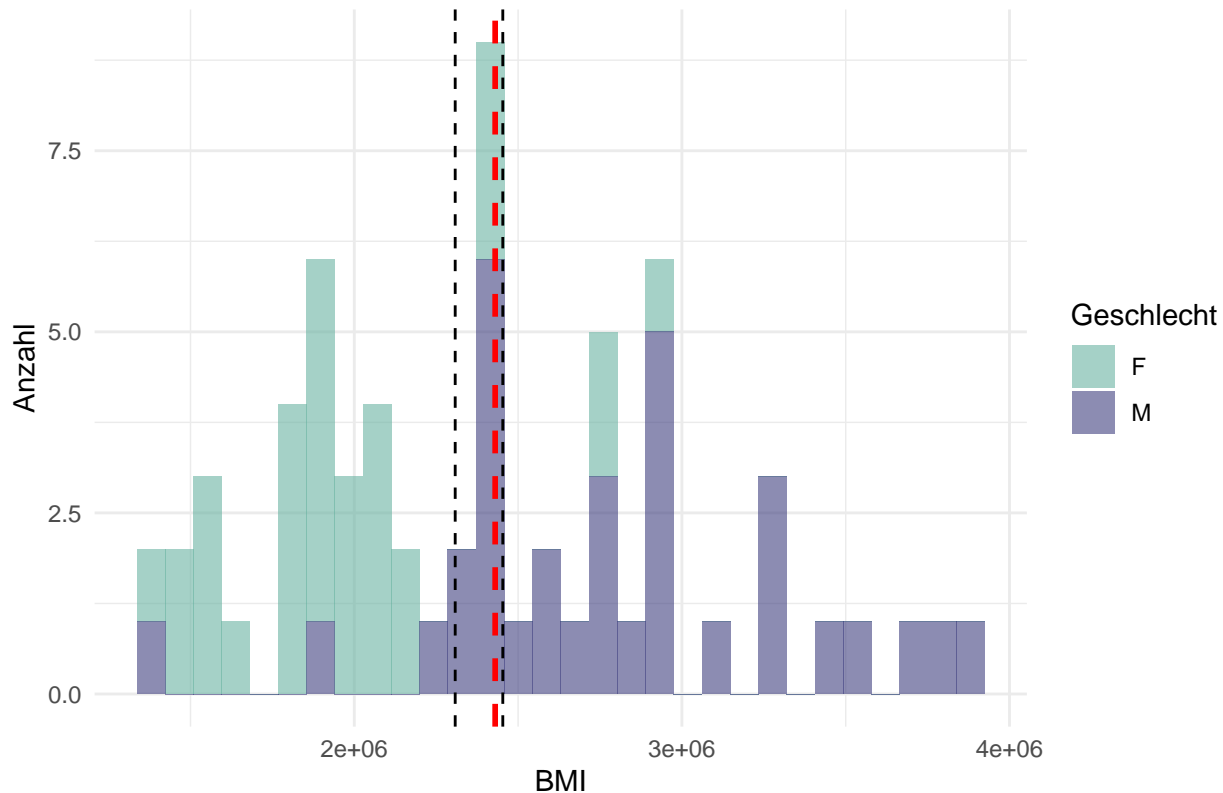
```



```
geom_vline(aes(xintercept = median(`D-weight` * (`D-height`)^2)),
  data = data_iga[`R-sex` == "M"], linetype = "dashed") +
ylab("Anzahl") +
xlab("BMI") + # Einheit???
ggtitle("Histogramm BMI Iga") +
labs(fill = "Geschlecht") +
default_theme
```

`stat_bin()` using `bins = 30`. Pick better value with `binwidth`.

Histogramm BMI Iga



```
# death donator
tbl_death_d <- cbind(c(nrow(data_iga[`Pat death (0=alive, 1= dead)` == 0]),
  nrow(data_iga[`Pat death (0=alive, 1= dead)` == 1])),
  c(nrow(data_iga_pos[`Pat death (0=alive, 1= dead)` == 0]),
  nrow(data_iga_pos[`Pat death (0=alive, 1= dead)` == 1])),
  c(nrow(data_iga_neg[`Pat death (0=alive, 1= dead)` == 0]),
  nrow(data_iga_neg[`Pat death (0=alive, 1= dead)` == 1])))
colnames(tbl_death_d) <- c("iga_all", "iga_pos", "iga_neg")
rownames(tbl_death_d) <- c("Lebenspende", "Todspende")

p1 <- data.frame(group = c("Lebenspende", "Todspende"),
  value = c(nrow(data_iga[`Pat death (0=alive, 1= dead)` == 0]),
  nrow(data_iga[`Pat death (0=alive, 1= dead)` == 1])) %>%
ggplot(aes(x = "", y = value, fill = group)) +
geom_bar(width = 1, stat = "identity") +
coord_polar("y", start = 0) +
guides(fill=guide_legend(title = "Spender")) +
```

```

theme_void() +
two_scale_fill +
ggtitle("iga all")

p2 <- data.frame(group = c("Lebensspende", "Todspende"),
  value = c(nrow(data_iga_pos[Pat death (0=alive, 1= dead) == 0]),
    nrow(data_iga_pos[Pat death (0=alive, 1= dead) == 1])) %>%
  ggplot(aes(x = "", y = value, fill = group)) +
  geom_bar(width = 1, stat = "identity") +
  coord_polar("y", start = 0) +
  guides(fill=guide_legend(title = "Spender")) +
  theme_void() +
  two_scale_fill +
  ggtitle("iga +")

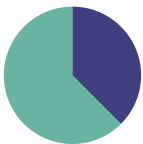
p3 <- data.frame(group = c("Lebensspende", "Todspende"),
  value = c(nrow(data_iga_neg[Pat death (0=alive, 1= dead) == 0]),
    nrow(data_iga_neg[Pat death (0=alive, 1= dead) == 1])) %>%
  ggplot(aes(x = "", y = value, fill = group)) +
  geom_bar(width = 1, stat = "identity") +
  coord_polar("y", start = 0) +
  guides(fill=guide_legend(title = "Spender")) +
  theme_void() +
  two_scale_fill +
  ggtitle("iga -")

patch <- p1 | p2 | p3
patch + plot_annotation(title = "Kreisdiagramme Spender")

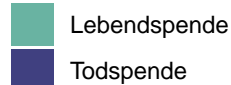
```

Kreisdiagramme Spender

iga all



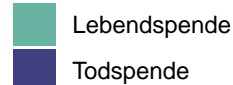
Spender



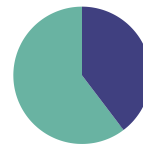
iga +



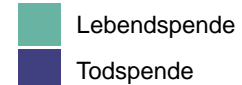
Spender



iga -



Spender



```

# ToDo: iga all + -
# deceased D.
# living D.
tbl_iga_1 <- summary(data_iga[graft loss (0=functional, 1=loss) == 1]$D-type)
tbl_iga_1 <- round(tbl_iga_1 / sum(tbl_iga_1), 3)
tbl_iga_2 <- summary(data_iga[graft loss (0=functional, 1=loss) == 0]$D-type)
tbl_iga_2 <- round(tbl_iga_2 / sum(tbl_iga_2), 3)
rbind("loss" = tbl_iga_1, "functional" = tbl_iga_2) %>%
  as.data.frame()

```

```

##           Cadaver Living
## loss      0.867  0.133
## functional 0.735  0.265

```

```

# dead/alive
tbl_iga_1 <- summary(data_iga[Pat death (0=alive, 1= dead) == 1]$D-type)

```

```
tbl_iga_1 <- round(tbl_iga_1 / sum(tbl_iga_1), 3)
tbl_iga_2<- summary(data_iga[Pat death (0=alive, 1= dead)` == 0]$D-type`)
tbl_iga_2 <- round(tbl_iga_2 / sum(tbl_iga_2),3) # in %
rbind("dead" = tbl_iga_1, "alive" = tbl_iga_2) %>%
  as.data.frame()
```

```
##          Cadaver Living
## dead      0.917  0.083
## alive     0.675  0.325
```

```
# BMI (mean.)
tbl_iga_bmi <- data.table(
  iga_all = mean(data_iga$D-weight` * (data_iga$D-height`)^2),
  iga_pos = mean(data_iga_pos$D-weight` * (data_iga_pos$D-height`)^2),
  iga_neg = mean(data_iga_neg$D-weight` * (data_iga_neg$D-height`)^2)
)
```

```
p1 <- data_iga %>%
  ggplot() +
  geom_boxplot(aes(x = "", y = mismatch_sum )) +
  default_theme +
  ggtitle("IGA all") +
  ylab("") +
  xlab("")
```

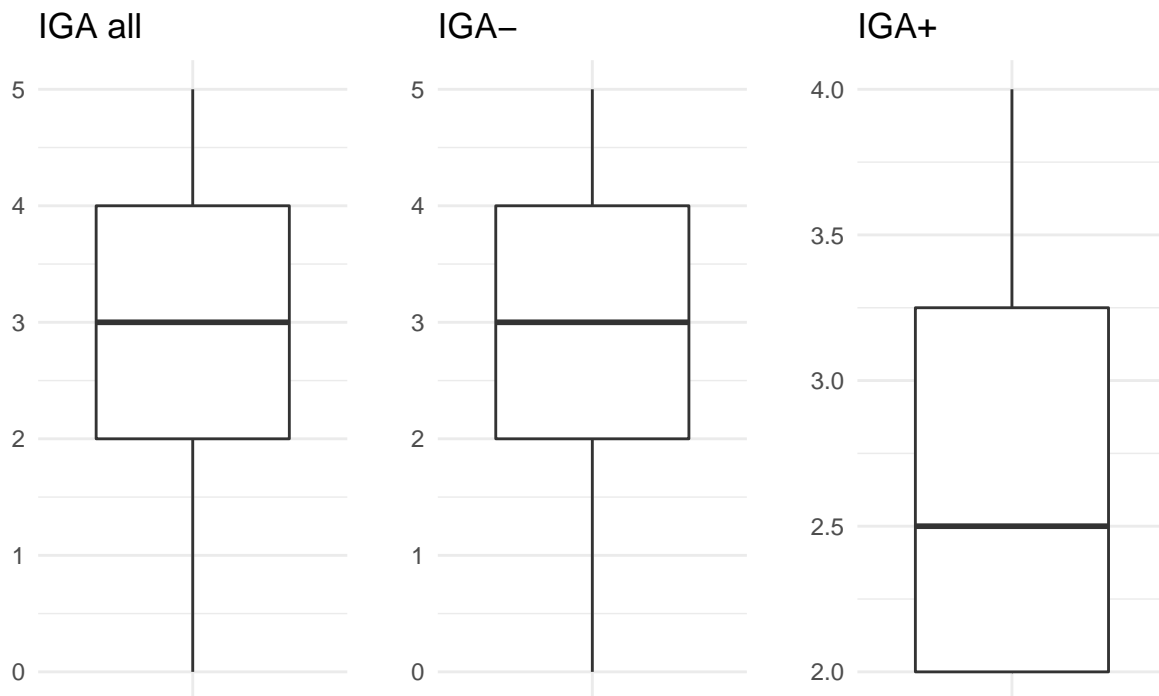
```
p2 <- data_iga_neg %>%
  ggplot() +
  geom_boxplot(aes(x = "", y = mismatch_sum)) +
  default_theme +
  ggtitle("IGA-") +
  ylab("") +
  xlab("")
```

```
p3 <- data_iga_pos %>%
  ggplot() +
  geom_boxplot(aes(x = "", y = mismatch_sum)) +
  default_theme +
  ggtitle("IGA+") +
  ylab("") +
  xlab("")
```

```
patch <- p1 | p2 | p3
patch + plot_annotation(title = "HLA mismatch summe")
```

```
## Warning: Removed 16 rows containing non-finite values (stat_boxplot).
## Warning: Removed 14 rows containing non-finite values (stat_boxplot).
## Warning: Removed 2 rows containing non-finite values (stat_boxplot).
```

HLA mismatch summe



```
# HLA-mm (0-6)
## data.table of mm-A, mm-B and mm-DR
tbl_iga_1 <- as.data.table(lapply(data_iga[, c("mm-A", "mm-B", "mm-DR")],
                                   as.numeric))

## calculate mean for each column
tbl_iga_2 <- apply(X = tbl_iga_1,
                  MARGIN = 2,
                  FUN = mean, na.rm = TRUE)
round(tbl_iga_2, 3)
```

```
## mm-A mm-B mm-DR
## 1.792 2.146 1.917
```

```
# mean of means
round(mean(tbl_iga_2),3)
```

```
## [1] 1.951
```

```
## sum of mm-A, mm-B and mm-DR
tbl_iga_3 <- apply(X = tbl_iga_1,
                  MARGIN = 1,
                  FUN = sum, na.rm = TRUE)
```

```
# mean value of col sum
mean(tbl_iga_3)
```

```
## [1] 4.390625
```

```
# median value of col sum
sd(tbl_iga_3)
```

```
## [1] 2.711628
```

```

# PRA current (mean)
tbl_iga_pra_curr_mean <- data.frame(
  iga_all = mean(data_iga$`Current PRA%`, na.rm = TRUE),
  iga_pos = mean(data_iga_pos$`Current PRA%`, na.rm = TRUE),
  iga_neg = mean(data_iga_neg$`Current PRA%`, na.rm = TRUE)
)

# PRA highest (mean)
tbl_iga_pra_high <- data.frame(
  iga_all = mean(data_iga$`Highest PRA%`, na.rm = TRUE),
  iga_pos = mean(data_iga_pos$`Highest PRA%`, na.rm = TRUE),
  iga_neg = mean(data_iga_neg$`Highest PRA%`, na.rm = TRUE)
)

# age donor (mean.)
tbl_iga_age_donor <- data.frame(
  iga_all = mean(data_iga$`D-age`),
  iga_pos = mean(data_iga_pos$`D-age`),
  iga_neg = mean(data_iga_neg$`D-age`)
)

# cold-ischemia time (hours)
# mean
tbl_iga_cis_mean <- data.frame(
  iga_all = mean(data_iga$`Cold ischaemic period hours`, na.rm = TRUE),
  iga_pos = mean(data_iga_pos$`Cold ischaemic period hours`, na.rm = TRUE),
  iga_neg = mean(data_iga_neg$`Cold ischaemic period hours`, na.rm = TRUE)
)

# median
tbl_iga_cis_median <- data.frame(
  iga_all = median(data_iga$`Cold ischaemic period hours`, na.rm = TRUE),
  iga_pos = median(data_iga_pos$`Cold ischaemic period hours`, na.rm = TRUE),
  iga_neg = median(data_iga_neg$`Cold ischaemic period hours`, na.rm = TRUE)
)

# standard error
tbl_iga_cis_sd <- data.frame(
  iga_all = sd(data_iga$`Cold ischaemic period hours`, na.rm = TRUE),
  iga_pos = sd(data_iga_pos$`Cold ischaemic period hours`, na.rm = TRUE),
  iga_neg = sd(data_iga_neg$`Cold ischaemic period hours`, na.rm = TRUE)
)

p1 <- data_iga %>%
  ggplot() +
  geom_boxplot(aes(x = `Cold ischaemic period hours`)) +
  coord_flip() +
  ggtitle("IGA (all)") +
  default_theme

p2 <- data_iga_pos %>%
  ggplot() +
  geom_boxplot(aes(x = `Cold ischaemic period hours`)) +
  coord_flip() +
  ggtitle("IGA +") +
  default_theme

```

```

p3 <- data_iga_neg %>%
  ggplot() +
  geom_boxplot(aes(x = `Cold ischaemic period hours`)) +
  coord_flip() +
  ggtitle("IGA -") +
  default_theme

p1$labels$x <- p2$labels$x <- p3$labels$x <- ""
p1$labels$y <- p2$labels$y <- p3$labels$y <- ""

patch <- p1 | p2 | p3
patch <- patch + plot_annotation(
  title = "Boxplot Cold Isch. Time h"
)
patch & xlim(0, 40) & theme(axis.ticks.x = element_blank(),
  axis.text.x = element_blank()) & xlab("Cold Isch Time")

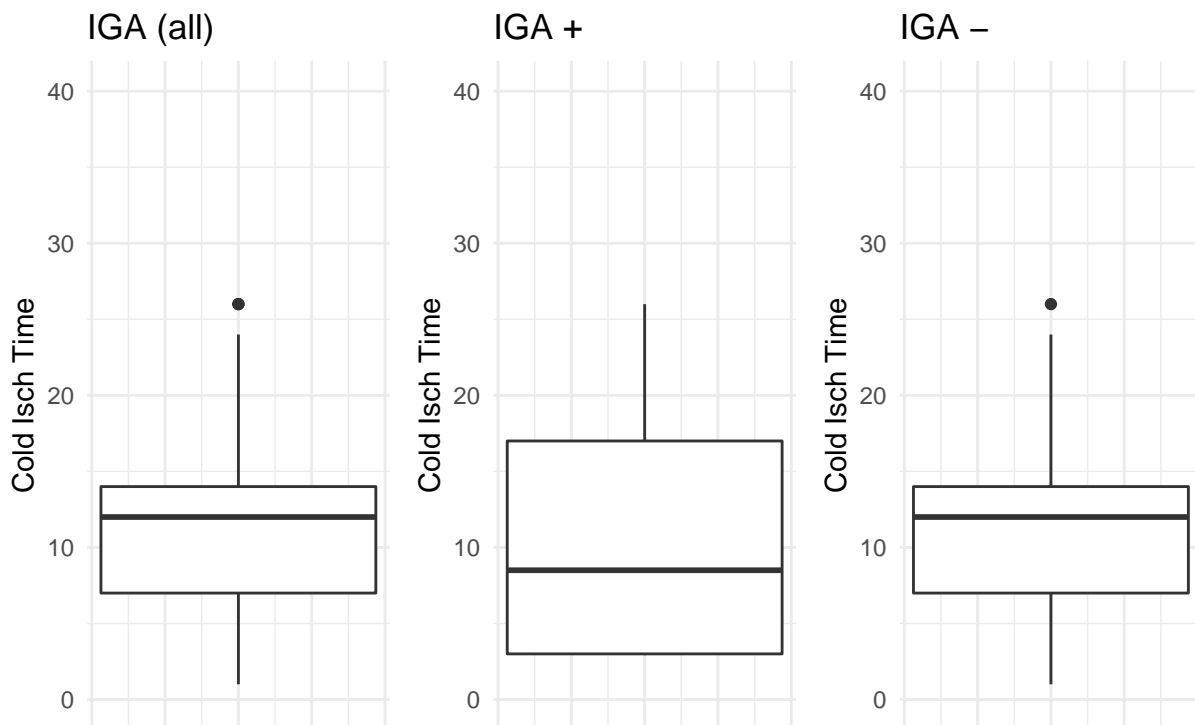
```

Warning: Removed 15 rows containing non-finite values (stat_boxplot).

Warning: Removed 2 rows containing non-finite values (stat_boxplot).

Warning: Removed 13 rows containing non-finite values (stat_boxplot).

Boxplot Cold Isch. Time h



```

# living vs dead donator
tbl_iga_don_living_abs <- data.frame(
  iga_all = nrow(data_iga[`D-type` == "Living"]),
  iga_pos = nrow(data_iga_pos[`D-type` == "Living"]),
  iga_neg = nrow(data_iga_neg[`D-type` == "Living"])
)

```

```

tbl_iga_don_dead_abs <- data.frame(
  iga_all = nrow(data_iga[`D-type` == "Cadaver"]),
  iga_pos = nrow(data_iga_pos[`D-type` == "Cadaver"]),
  iga_neg = nrow(data_iga_neg[`D-type` == "Cadaver"])
)

tbl_iga_don_living_rel <- data.frame(
  iga_all = nrow(data_iga[`D-type` == "Living"]) / nrow(data_iga),
  iga_pos = nrow(data_iga_pos[`D-type` == "Living"]) / nrow(data_iga_pos),
  iga_neg = nrow(data_iga_neg[`D-type` == "Living"]) / nrow(data_iga_neg)
)

tbl_iga_don_dead_rel <- data.frame(
  iga_all = nrow(data_iga[`D-type` == "Cadaver"]) / nrow(data_iga),
  iga_pos = nrow(data_iga_pos[`D-type` == "Cadaver"]) / nrow(data_iga_pos),
  iga_neg = nrow(data_iga_neg[`D-type` == "Cadaver"]) / nrow(data_iga_neg)
)

iga_table <- rbindlist(
  list(
    iga_dropout = data.frame(
      cbind(name = "dropout", tbl_iga_pat_drop)),
    iga_graftloss = data.frame(
      cbind(name = "graftloss", tbl_iga_graft_loss)),
    iga_graftloss_follow_up = data.frame(
      cbind(name = "graftloss_followup", tbl_iga_graft_loss_follow_up)),
    iga_pat_death = data.frame(
      cbind(name = "pat_death", tbl_iga_pat_death)),
    iga_don_dead_abs = data.frame(
      cbind(name = "don_dead_abs", tbl_iga_don_dead_abs)),
    iga_don_dead_rel = data.frame(
      cbind(name = "don_dead_rel", tbl_iga_don_dead_rel)),
    iga_don_living_abs = data.frame(
      cbind(name = "don_living_abs", tbl_iga_don_living_abs)),
    iga_don_living_rel = data.frame(
      cbind(name = "don_living_rel", tbl_iga_don_living_rel)),
    iga_cis_h_mean = data.frame(
      cbind(name = "cold_h_mean", tbl_iga_cis_mean)),
    iga_cis_h_sd = data.frame(
      cbind(name = "cold_h_sd", tbl_iga_cis_sd))
  ))

kable(iga_table, col.names = c("Merkmal", "IgA (all)", "IgA r+", "IgA r-")) %>%
  kable_styling(latex_options = "hold_position")

# TODO mittelwert wann recurrence?

```

§ 2.2 NTX

```

# follow_up mean
pmin(
  # follow up
  interval(data_ntx$Datum, (data_ntx$Datum + follow_up)) / years(1),

```

Merkmal	IgA (all)	IgA r+	IgA r-
dropout	27.000000	3.000000	24.000000
graftloss	15.000000	2.000000	13.000000
graftloss_followup	14.000000	2.000000	12.000000
pat_death	23.000000	1.000000	22.000000
don_dead_abs	49.000000	4.000000	45.000000
don_dead_rel	0.765625	0.666667	0.775862
don_living_abs	15.000000	2.000000	13.000000
don_living_rel	0.234375	0.333333	0.224137
cold_h_mean	11.612245	11.500000	11.622222
cold_h_sd	6.330535	10.969651	5.959289

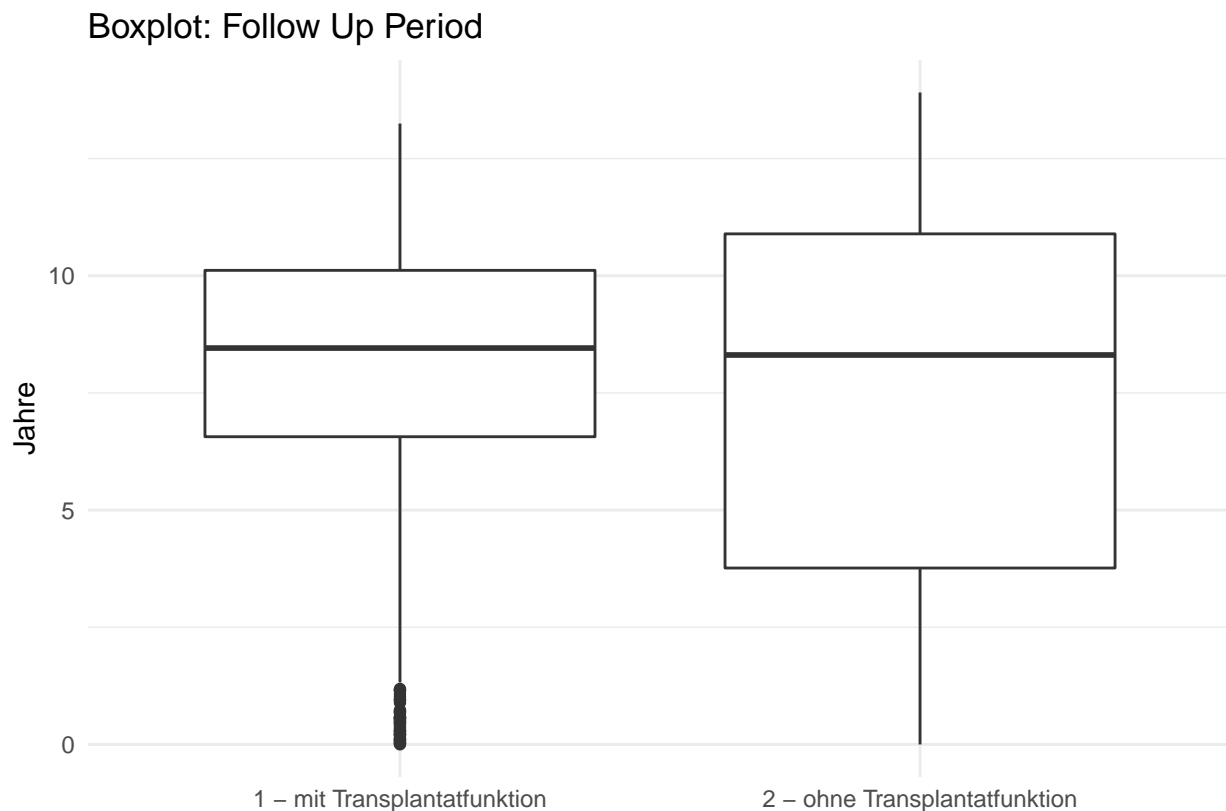
```

# last seen
interval(data_ntx$Datum, data_ntx$tdls) / years(1)
) %>%
  mean()

## [1] 7.41754

# ToDo; keine Stratifizierung
ggplot() +
  geom_boxplot(aes(x = data_ntx$`TX Status[NTX PatientenInformation]`,
                  y = interval(data_ntx$Datum, data_ntx$tdls) / years(1)), data = data_ntx) +
  default_theme +
  ggtitle("Boxplot: Follow Up Period") +
  ylab("Jahre") +
  xlab("")

```




```

# patient death within follow up
nrow(data_ntx[`Todesdatum[NTX PatientenInformation]` < (Datum + follow_up)])

## [1] 317

# patient drop out
nrow(data_ntx[`Date last seen[NTX PatientenInformation]` <= (Datum + follow_up) & `Patienten Status[NTX

## [1] 370

# patients with graft loss
nrow(data_ntx[data_ntx$`TX Status[NTX PatientenInformation]` == "2 - ohne Transplantatfunktion"])
```

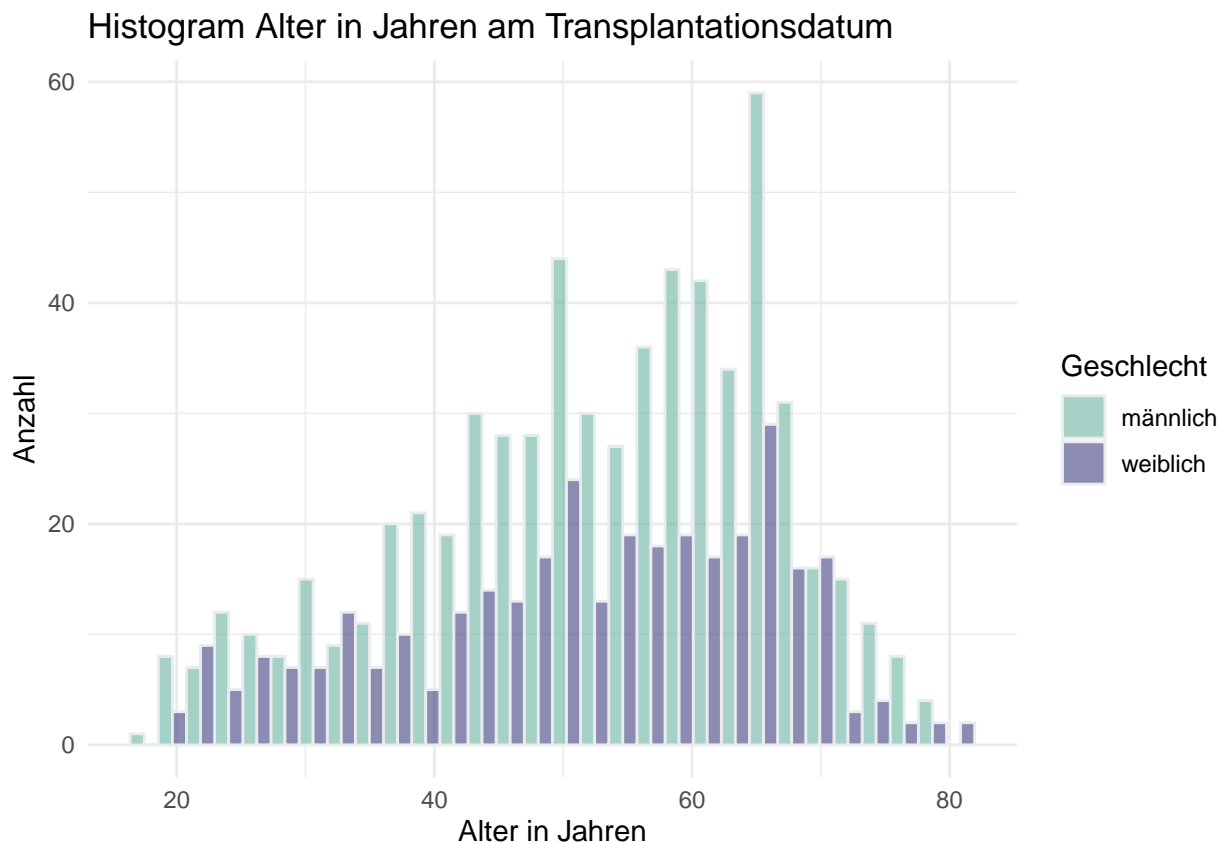
```

## [1] 181

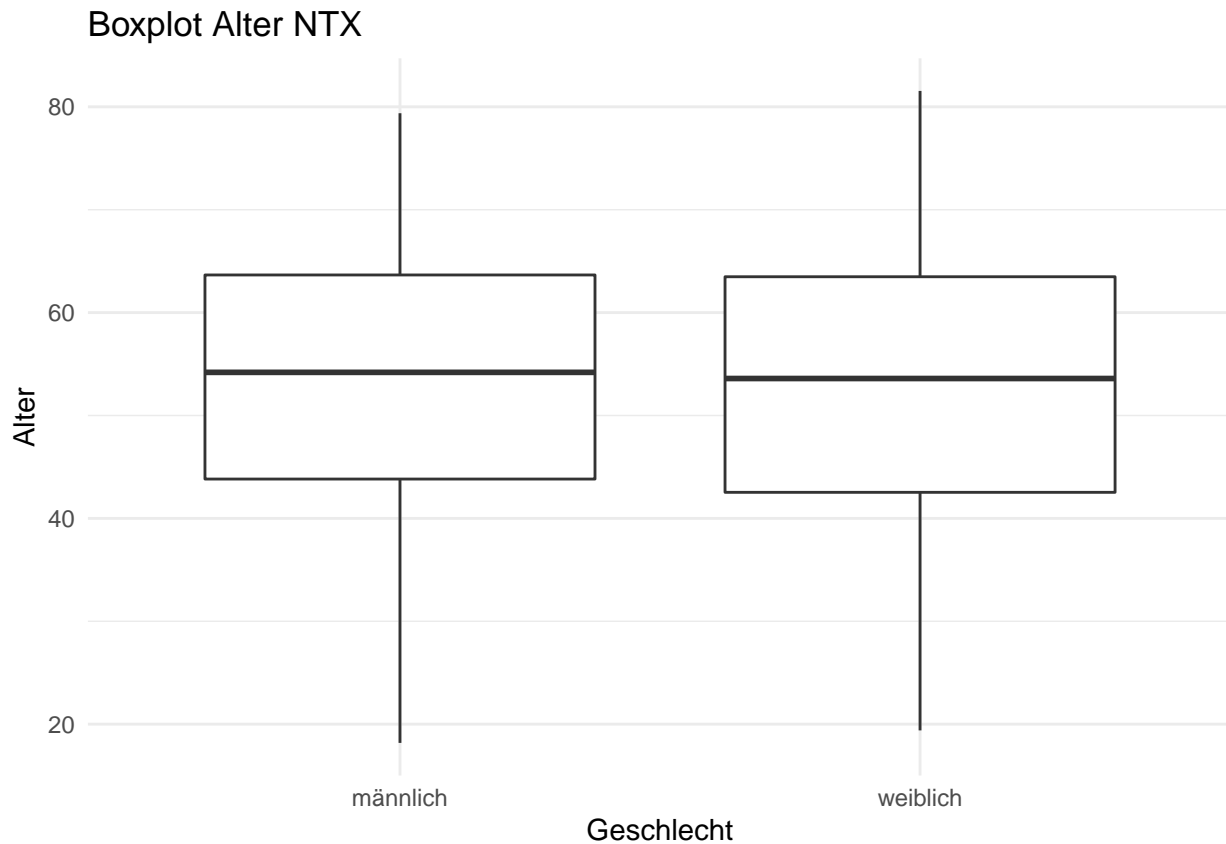
# age patients (yrs.)
ggplot(data = data_ntx) +
  geom_histogram(mapping = aes(x = R_age_Datum, fill = Geschlecht),
    color = "#e9ecef", alpha = 0.6, position = "dodge") +
  two_scale_fill +
  ylab("Anzahl") +
  xlab("Alter in Jahren") +
  ggtitle("Histogram Alter in Jahren am Transplantationsdatum") +
  labs(fill = "Geschlecht") +
  default_theme

## `stat_bin()` using `bins = 30`. Pick better value with `binwidth`.

```



```
ggplot(data = data_ntx) +
  geom_boxplot(aes(x = data_ntx$Geschlecht, y = data_ntx$R_age_Datum)) +
  default_theme +
  ggtitle("Boxplot Alter NTX") +
  xlab("Geschlecht") +
  ylab("Alter")
```

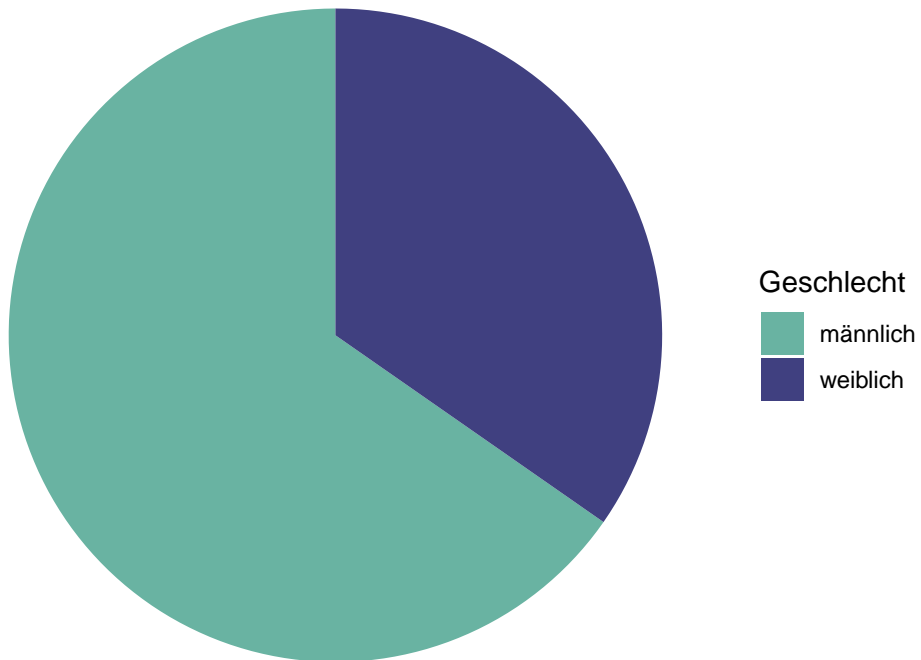


```
# male sex
summary(data_ntx$Geschlecht)
```

```
## männlich weiblich
##      627      333
```

```
data.frame(group = names(summary(data_ntx$Geschlecht)),
            value = c(summary(data_ntx$Geschlecht))) %>%
  ggplot(aes(x = "", y = value, fill = group)) +
  geom_bar(width = 1, stat = "identity") +
  coord_polar("y", start = 0) +
  guides(fill=guide_legend(title = "Geschlecht")) +
  theme_void() +
  two_scale_fill +
  ggtitle("NTX")
```

NTX



```
# BMI (mean.)  
## NO DATA AVAILABLE
```

```
# deceased D.  
# living D.  
## NO DATA AVAILABLE
```

```
# HLA-mm  
## NO DATA AVAILABLE
```

```
# PPR  
## NO DATA AVAILABLE
```

```
# living vs dead donator  
## NO DATA AVAILABLE
```

§ 3 Kaplan-Meier

§ 3.1 IGA

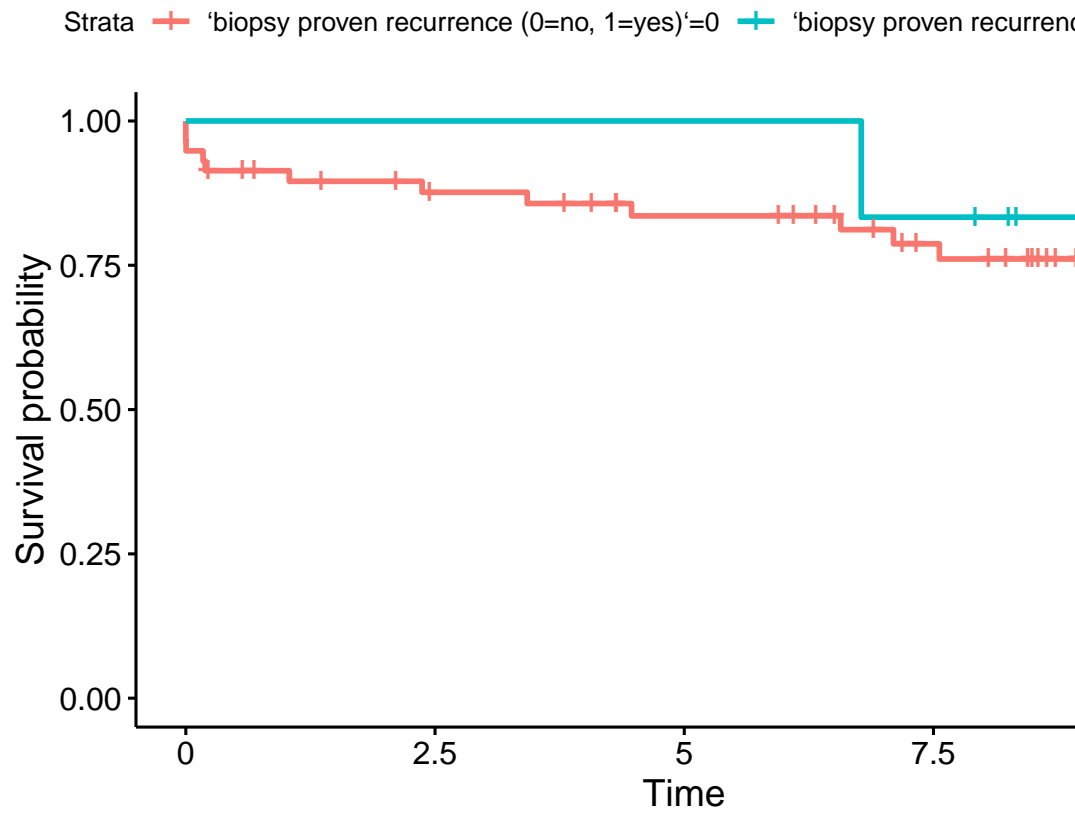
```
# functions in survival and survminer package need numeric-type input
data_iga[, time_date_biopsy := interval(`T-date`, `date of biopsy`) / years(1)]
data_iga[, time_t_dls := interval(`T-date`, `T-dls`) / years(1)]
data_iga[, time_date_birth := interval(`T-date`, `Date of birth`) / years(1)]
data_iga[, time_graft_loss := interval(`T-date`, `graft loss date`) / years(1)]
data_iga[, time_date_follow_up := interval(`T-date`, `T-date` + follow_up) / years(1)]
```

§ 3.1.1

- Event:
 - graft-loss within the follow up period.
- Censoring scheme:
 - if graft loss date after follow up period, censored by end of follow up.
 - if T-dls (date last seen) within follow up period, censored by T-dls.
- Time period:
 - 10 years after T-date (kidney transplantaion).

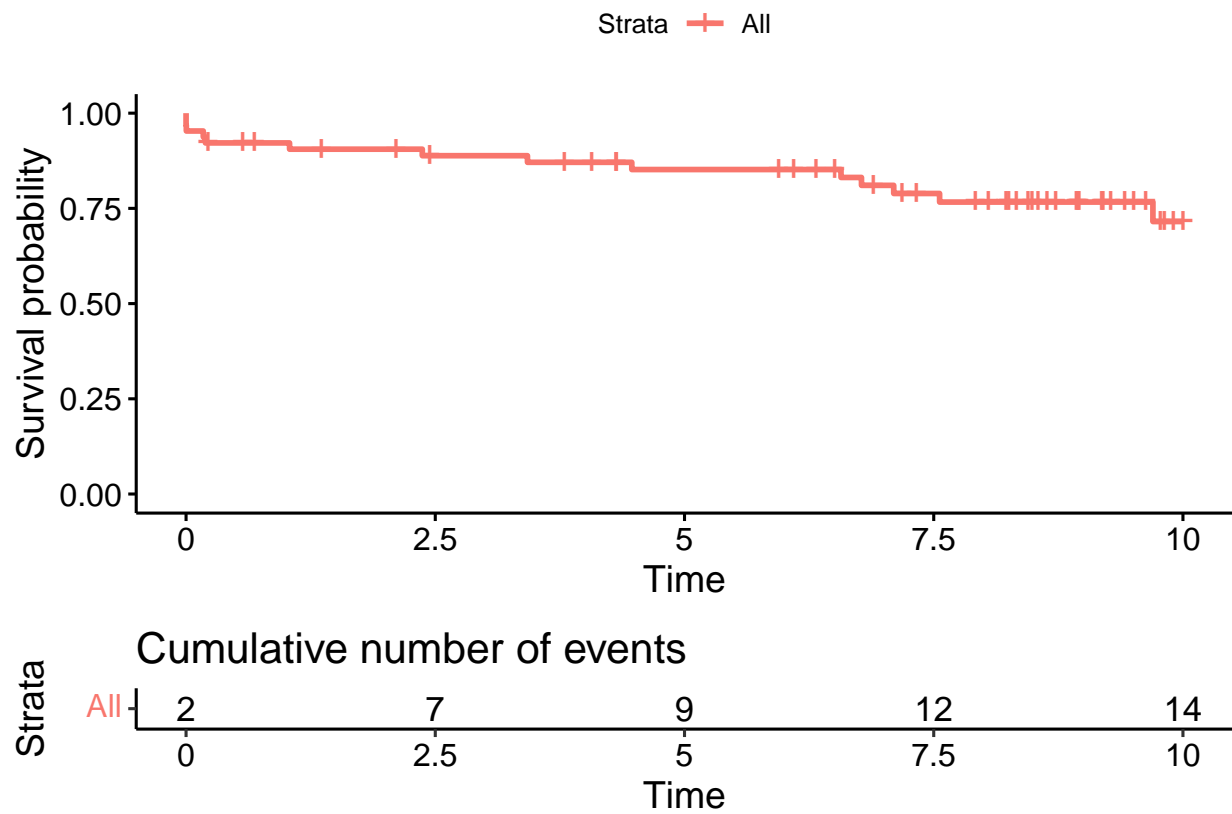
```
data_iga <- data_iga %>%
  mutate(status_date = case_when(
    ## graft-loss within follow up period
    !is.na(`graft loss date`) & `graft loss date` < `T-date` + follow_up ~ time_graft_loss,
    ## graft-loss after follow up period
    !is.na(`graft loss date`) & `graft loss date` > `T-date` + follow_up ~ time_date_follow_up,
    ## no graft-loss and last seen within follow up
    is.na(`graft loss date`) & !is.na(`T-dls`) & `T-dls` < `T-date` + follow_up ~ time_t_dls,
    ## no graft-loss and last seen after follow up
    is.na(`graft loss date`) & !is.na(`T-dls`) & `T-dls` > `T-date` + follow_up ~ time_date_follow_up,
    ## no graft loss and no last seen
    is.na(`graft loss date`) & is.na(`T-dls`) ~ time_date_follow_up
  )
)
data_iga <- data_iga %>%
  mutate(status = case_when(
    ## graft-loss within follow up period
    !is.na(`graft loss date`) & `graft loss date` < `T-date` + follow_up ~ 1,
    ## else censored
    TRUE ~ 0,
  )
)
data_iga_pos <- data_iga[`biopsy proven recurrence (0=no, 1=yes)` == 1]
data_iga_neg <- data_iga[`biopsy proven recurrence (0=no, 1=yes)` == 0]
```

```
model_iga_1 <- survfit(formula = Surv(time = status_date, event = status, type = "right") ~ data_iga$b
ggsurvplot(model_iga_1)
```

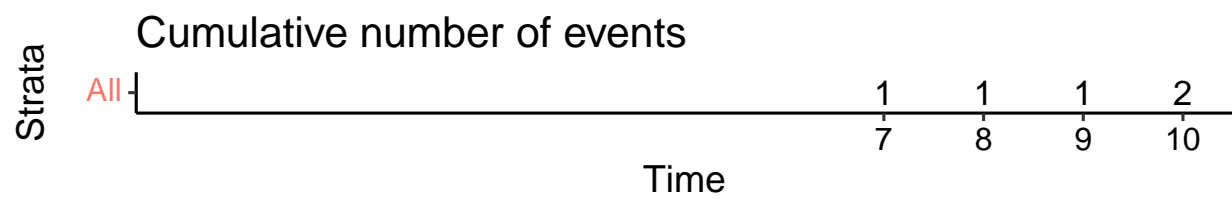
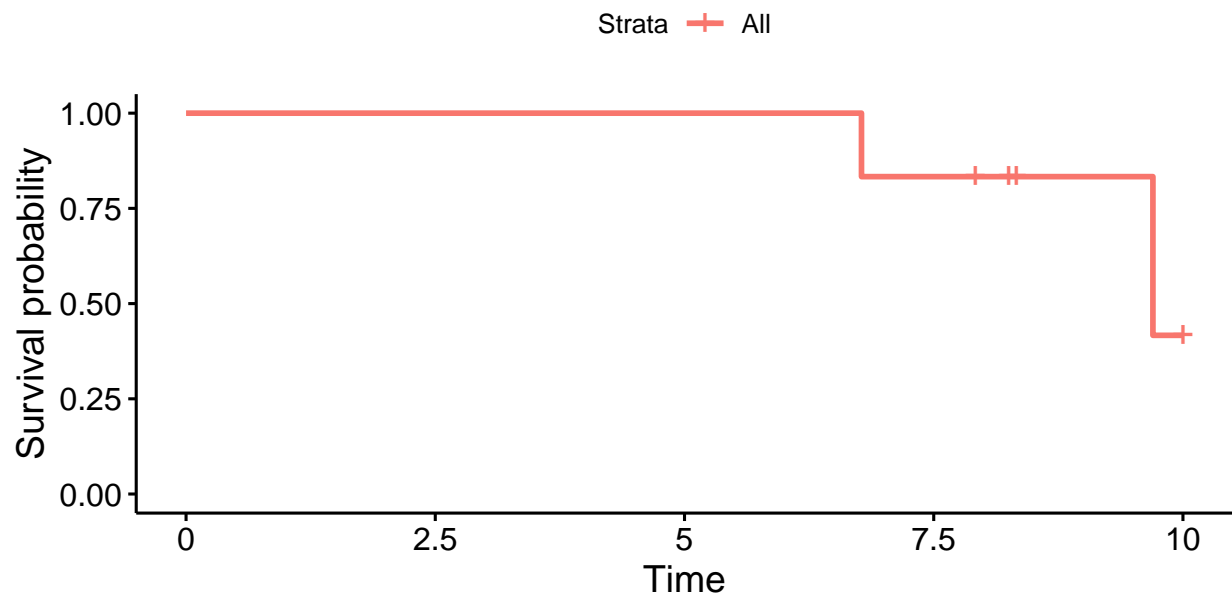


Overall kaplan-Meier curve

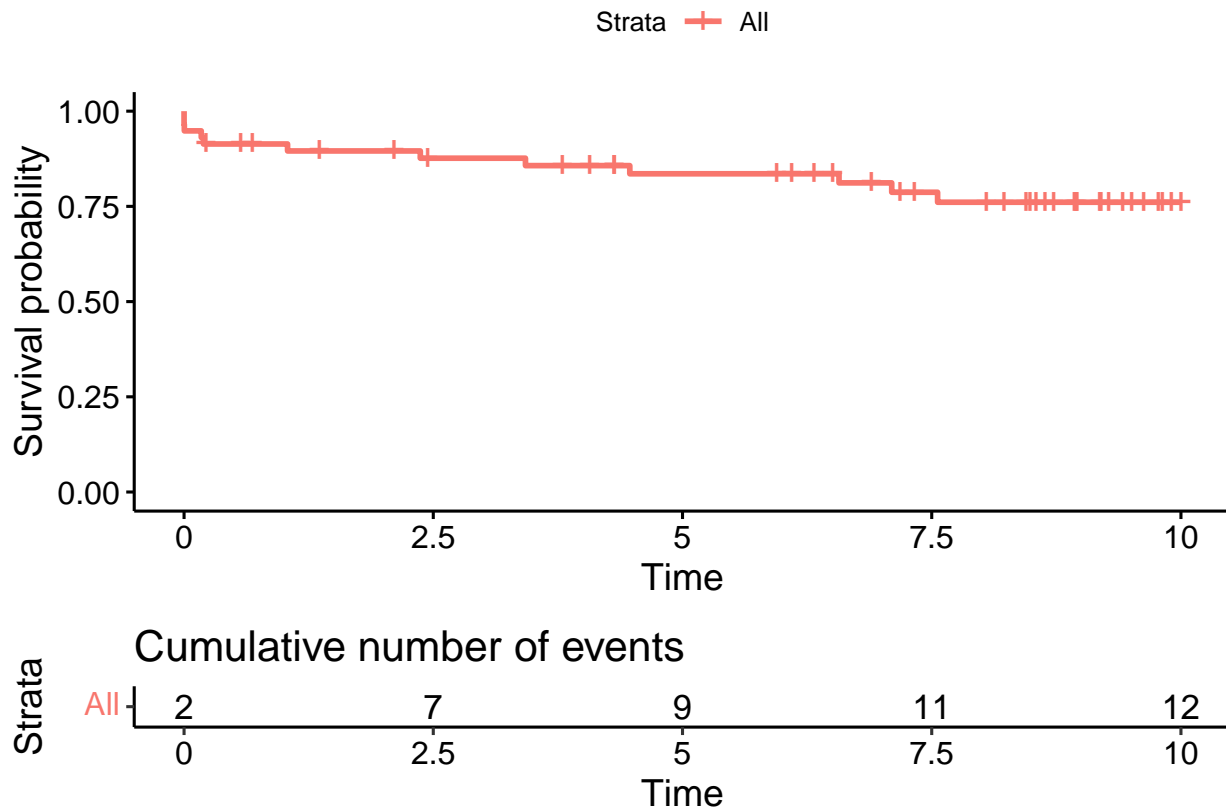
```
model_iga_1.1 <- survfit(formula = Surv(time = status_date,
                                     event = status, type = "right")~ 1,
                        data = data_iga)
ggsurvplot(model_iga_1.1,
            conf.int = FALSE,
            cumevents = TRUE)
```



```
## iga+
model_iga_1.1 <- survfit(formula = Surv(time = status_date,
                                         event = status, type = "right") ~ 1,
                         data = data_iga_pos)
ggsurvplot(model_iga_1.1,
            conf.int = FALSE,
            cumevents = TRUE)
```



```
## iga-
model_iga_1.2 <- survfit(formula = Surv(time = status_date,
                                     event = status, type = "right") ~ 1,
                        data = data_iga_neg)
ggsurvplot(model_iga_1.2,
            conf.int = FALSE,
            cumevents = TRUE)
```



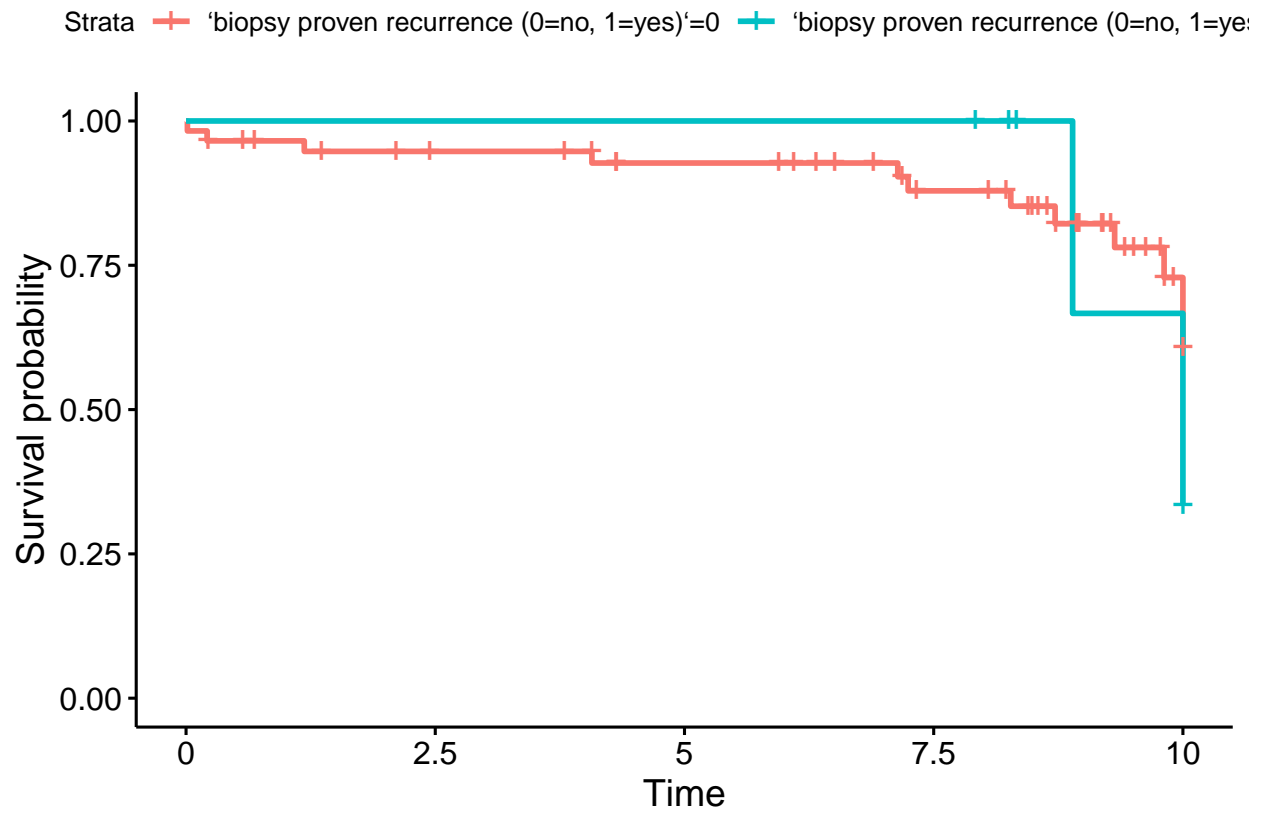
§ 3.1.2

- Event:
 - T-dls & 'Pat death (0=alive, 1= dead) (patient death).
- Censoring scheme:
 - T-dls (date last seen) within follow up period.
- Time period:
 - 10 years after T-date (kidney transplantation).

```
data_iga <- data_iga %>%
  mutate(status_date = case_when(
    ## patient death and death date within follow up
    (`Pat death (0=alive, 1= dead)` == 1) & `T-dls` < `T-date` + follow_up ~ time_t_dls,
    ## patient dead but after follow up
    (`Pat death (0=alive, 1= dead)` == 1) & `T-dls` > `T-date` + follow_up ~ time_date_follow_up,
    ## patient not death but dropped within follow up
    (`Pat death (0=alive, 1= dead)` == 0) & `T-dls` < `T-date` + follow_up ~ time_t_dls,
    ## patient not death but dropped after follow up
    (`Pat death (0=alive, 1= dead)` == 0) & `T-dls` > `T-date` + follow_up ~ time_date_follow_up,
    ## NOTE: T-dls never NA
  ))
)

model_iga_2 <- survfit(formula = Surv(time = status_date,
                                     event = status, type = "right") ~ data_iga$`biopsy proven recurrence`,
                      data = data_iga)

ggsurvplot(model_iga_2,
            conf.int = FALSE)
```

ASDFGHJKLKJHGFDSASDFGHJKLKJHGFDS

§ 3.1.3

- Event: graft loss and death
- Censoring scheme:
- Time period:

§ 3.2 IGA

§ 3.2.1

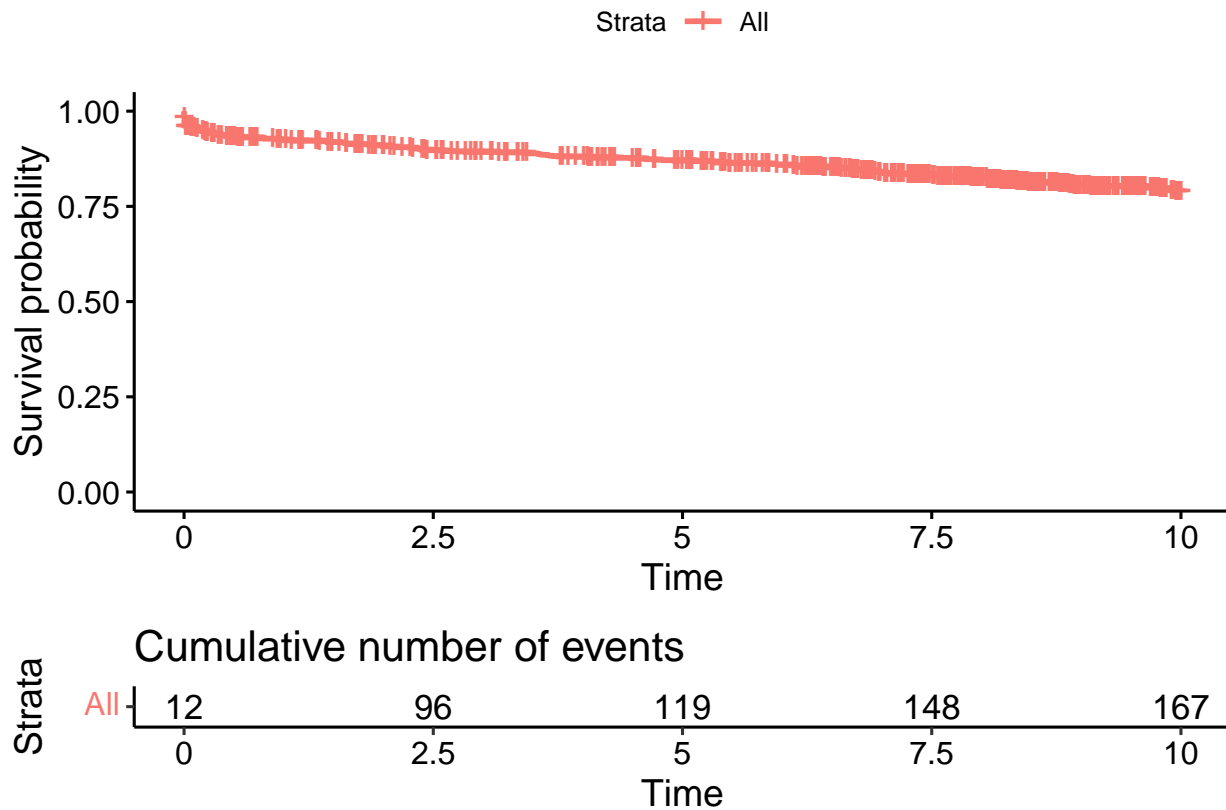
- Event: TX Status[NTX PatientenInformation]
- Censoring scheme: -if TX Status[NTX PatientenInformation] within follow up period then then event
 - if died within follow up and before no graft loss then censored
 - if last seen within follow up and before no graft loss then censored
 - if graft loss after follow up then censored
- Time period: 10 years after Datum

```
data_ntx <- data_ntx %>%
  mutate(status_date = case_when(
    ## patient experienced graft loss
    !is.na(Transplantatfunktionsende) & Transplantatfunktionsende <= (Datum + follow_up) ~ interval(Datum, Transplantatfunktionsende),
    ## patient died within follow up
    `Todesdatum[NTX PatientenInformation]` < (Datum + follow_up) ~ interval(Datum, `Todesdatum[NTX PatientenInformation]`),
    ## patient last seen within follow up
    `Date last seen[NTX PatientenInformation]` < (Datum + follow_up) ~ interval(Datum, `Date last seen[NTX PatientenInformation]`),
    ## else follow up
    TRUE ~ interval(Datum, (Datum + follow_up)) / years(1)
  ))

data_ntx <- data_ntx %>%
  mutate(status = case_when(
    ## patient experienced graft loss
    !is.na(Transplantatfunktionsende) & Transplantatfunktionsende <= (Datum + follow_up) ~ 1,
    ## patient died within follow up
    `Todesdatum[NTX PatientenInformation]` < (Datum + follow_up) ~ 0,
    ## patient last seen within follow up
    `Date last seen[NTX PatientenInformation]` < (Datum + follow_up) ~ 0,
    ## else follow up
    TRUE ~ 0
  ))

model_ntx_1 <- survfit(formula = Surv(time = status_date,
                                     event = status, type = "right") ~ 1,
                      data = data_ntx)

ggsurvplot(model_ntx_1,
            conf.int = FALSE,
            cumevents = TRUE)
```



§ 3.2.2

- Event: patient died within follow up period
- Censoring scheme:
 - patient died within follow up then event
 - patient dropped from study within follow up then censored

```
data_ntx <- data_ntx %>%
  mutate(status_date = case_when(
    ## patient died within follow up
    `Todesdatum[NTX PatientenInformation]` <= (Datum + follow_up) ~ interval(Datum, `Todesdatum[NTX Pat
    ## patient died after follow up
    `Todesdatum[NTX PatientenInformation]` > (Datum + follow_up) ~ interval(Datum, (Datum + follow_up))
    ## patient dropped within follow up
    `Date last seen[NTX PatientenInformation]` <= (Datum + follow_up) ~ interval(Datum, `Date last seen
    ## patient dropped after follow up
    `Date last seen[NTX PatientenInformation]` > (Datum + follow_up) ~ interval(Datum, (Datum + follow_u
  )
)

data_ntx <- data_ntx %>%
  mutate(status = case_when(
    ## patient died within follow up
    `Todesdatum[NTX PatientenInformation]` <= (Datum + follow_up) ~ 1,
    ## patient died after follow up
    `Todesdatum[NTX PatientenInformation]` > (Datum + follow_up) ~ 0,
    ## patient dropped within follow up
```

```

`Date last seen[NTX PatientenInformation]` <= (Datum + follow_up) ~ 0,
## patient dropped after follow up
`Date last seen[NTX PatientenInformation]` > (Datum + follow_up) ~ 0
)
)

```

```

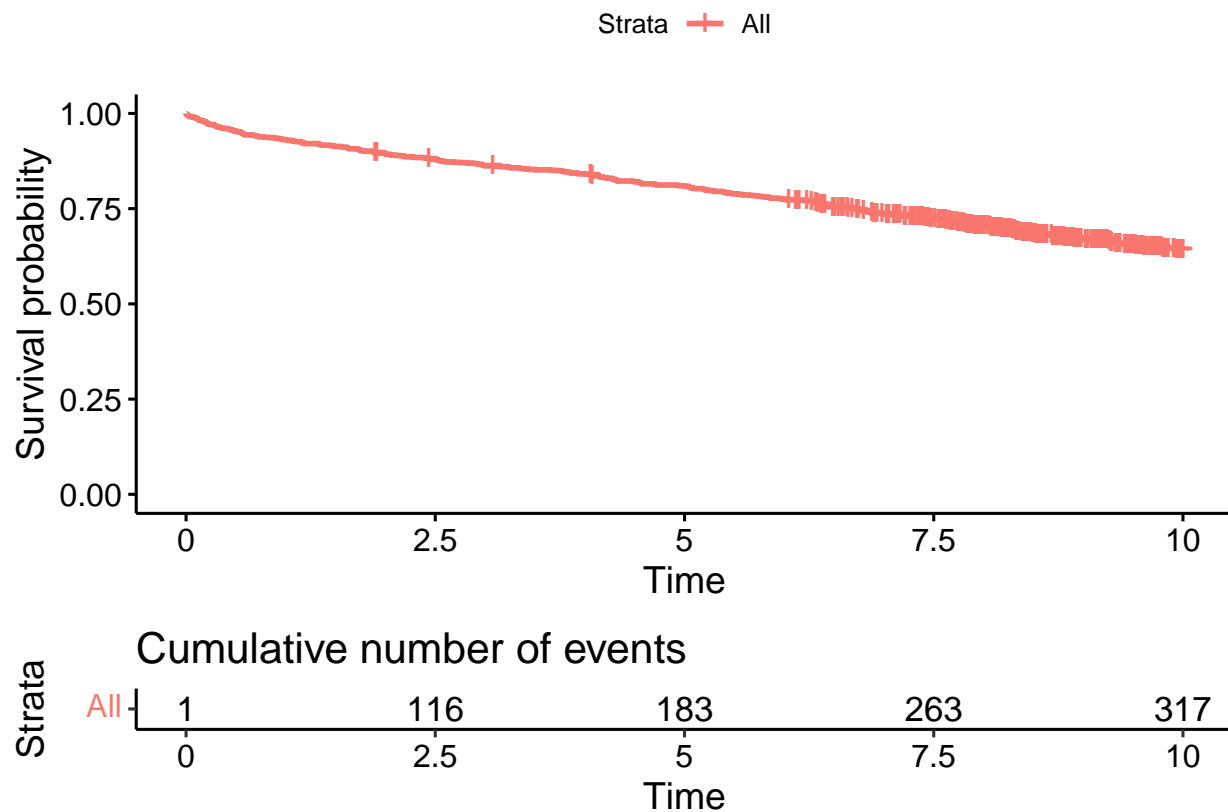
model_ntx_2 <- survfit(Surv(time = status_date, event = status,
                             type = "right") ~ 1, data = data_ntx)

```

```

ggsurvplot(model_ntx_2,
            conf.int = FALSE,
            cumevents = TRUE)

```



§ 4 Cox regression

§ 4.1 IGA

- Event:
 - graft loss within follow up
- Censoring scheme:
 - if patient dropped within follow up, then censored by time dropped
 - if patient patient experienced graft loss after follow up, then censored by follow up end
 - if patient experienced death within follow up, then censored by death date else by follow up end

```
data_iga <- data_iga %>%
  ## censor/event date
  mutate(status_date = case_when(
    ## patient dropped during follow up
    (`T-dls` <= `T-date` + follow_up) ~ `T-dls`,
    ## patient experienced graft loss but after follow up
    ((!is.na(`graft loss date`)) & (`graft loss date` > `T-date` + follow_up)) ~ `T-date` + follow_up,
    ## patient experienced graft loss within follow up
    ((!is.na(`graft loss date`)) & (`graft loss date` <= `T-date` + follow_up)) ~ `graft loss date`,
    ## patient experienced no graft loss within follow up, neither dropped
    (is.na(`graft loss date`) & (`T-dls` > `T-date` + follow_up)) ~ `T-date` + follow_up
  )) %>%
  ## status indicator
  mutate(status = case_when(
    ## patient dropped during follow up
    (`T-dls` <= `T-date` + follow_up) ~ 0,
    ## patient experienced graft loss but after follow up
    ((!is.na(`graft loss date`)) & (`graft loss date` > `T-date` + follow_up)) ~ 0,
    ## patient experienced graft loss within follow up
    ((!is.na(`graft loss date`)) & (`graft loss date` <= `T-date` + follow_up)) ~ 1,
    ## patient experienced no graft loss within follow up, neither dropped
    (is.na(`graft loss date`) & (`T-dls` > `T-date` + follow_up)) ~ 0
  ))
```

```
data_iga <- data_iga %>%
  ## censor/event date
  mutate(status_date = case_when(
    ## if graft loss within follow up
    `graft loss date` <= `T-date` + follow_up ~ `graft loss date`,
    ## else
    TRUE ~ `T-dls`
  )) %>%
  ## censor/event indicator
  mutate(status = case_when(
    ## if graft loss within follow up
    data_iga$`graft loss date` <= data_iga$`T-date` + follow_up ~ 1,
    ## else
    TRUE ~ 0
  ))
```

```
data_iga_pos <- data_iga[`biopsy proven recurrence (0=no, 1=yes)` == 1]
data_iga_neg <- data_iga[`biopsy proven recurrence (0=no, 1=yes)` == 0]
```

```
model_iga_cox <- coxph(data = data_iga, formula = Surv(time = as.numeric(status_date),
  event = status) ~ R_age_Tdate +
```

```

data_iga$`R-sex` +
data_iga$`D-type` +
## hla mismatch???
data_iga$`mm-A` +
data_iga$`Cold ischaemic period hours` +
data_iga$`D-age`)

```

```

## Warning in fitter(X, Y, istrat, offset, init, control, weights = weights, :
## Loglik converged before variable 5 ; coefficient may be infinite.

```

```
summary(model_iga_cox)
```

```

## Call:
## coxph(formula = Surv(time = as.numeric(status_date), event = status) ~
##       R_age_Tdate + data_iga$`R-sex` + data_iga$`D-type` + data_iga$`mm-A` +
##       data_iga$`Cold ischaemic period hours` + data_iga$`D-age`,
##       data = data_iga)
##
##      n= 38, number of events= 9
##      (26 observations deleted due to missingness)
##
##               coef exp(coef) se(coef)      z
## R_age_Tdate      -1.101e-02  9.891e-01  3.584e-02 -0.307
## data_iga$`R-sex`M      1.284e-01  1.137e+00  8.645e-01  0.149
## data_iga$`D-type`Living      NA      NA  0.000e+00      NA
## data_iga$`mm-A`1      -3.635e-02  9.643e-01  8.057e-01 -0.045
## data_iga$`mm-A`2      -1.833e+01  1.100e-08  1.197e+04 -0.002
## data_iga$`Cold ischaemic period hours` -5.066e-03  9.949e-01  6.792e-02 -0.075
## data_iga$`D-age`      4.052e-02  1.041e+00  2.901e-02  1.397
##
##               Pr(>|z|)
## R_age_Tdate      0.759
## data_iga$`R-sex`M      0.882
## data_iga$`D-type`Living      NA
## data_iga$`mm-A`1      0.964
## data_iga$`mm-A`2      0.999
## data_iga$`Cold ischaemic period hours`  0.941
## data_iga$`D-age`      0.162
##
##               exp(coef) exp(-coef) lower .95 upper .95
## R_age_Tdate      9.891e-01  1.011e+00  0.9220  1.061
## data_iga$`R-sex`M      1.137e+00  8.795e-01  0.2089  6.190
## data_iga$`D-type`Living      NA      NA      NA      NA
## data_iga$`mm-A`1      9.643e-01  1.037e+00  0.1988  4.677
## data_iga$`mm-A`2      1.100e-08  9.091e+07  0.0000  Inf
## data_iga$`Cold ischaemic period hours`  9.949e-01  1.005e+00  0.8709  1.137
## data_iga$`D-age`      1.041e+00  9.603e-01  0.9838  1.102
##
## Concordance= 0.701 (se = 0.095 )
## Likelihood ratio test= 3.73 on 6 df, p=0.7
## Wald test = 2.19 on 6 df, p=0.9
## Score (logrank) test = 3.02 on 6 df, p=0.8

```

```
cox.zph(model_iga_cox)
```

```
##               chisq df      p
```

```
## R_age_Tdate                2.2969  1 0.13
## data_iga$`R-sex`           2.5696  1 0.11
## data_iga$`mm-A`           0.0577  2 0.97
## data_iga$`Cold ischaemic period hours` 0.4896  1 0.48
## data_iga$`D-age`          0.5799  1 0.45
## GLOBAL                     3.6546  6 0.72
```

all p-values are relatively large, therefore the Null hypothesis of proportional hazards can not be .

\$ 4.2 NTX

```
data_ntx <- data_ntx %>%
  mutate(status_date = case_when(
    ## patient died within follow up
    `Todesdatum[NTX PatientenInformation]` <= (Datum + follow_up) ~ interval(Datum, `Todesdatum[NTX Pat
    ## patient died after follow up
    `Todesdatum[NTX PatientenInformation]` > (Datum + follow_up) ~ interval(Datum, Datum + follow_up) /
    ## patient dropped within follow up
    `Date last seen[NTX PatientenInformation]` <= (Datum + follow_up) ~ interval(Datum, `Date last seen
    ## patient dropped after follow up
    `Date last seen[NTX PatientenInformation]` > (Datum + follow_up) ~ interval(Datum, Datum + follow_up
  )
)
```

```
data_ntx <- data_ntx %>%
  mutate(status = case_when(
    ## patient died within follow up
    `Todesdatum[NTX PatientenInformation]` <= (Datum + follow_up) ~ 1,
    ## patient died after follow up
    `Todesdatum[NTX PatientenInformation]` > (Datum + follow_up) ~ 0,
    ## patient dropped within follow up
    `Date last seen[NTX PatientenInformation]` <= (Datum + follow_up) ~ 0,
    ## patient dropped after follow up
    `Date last seen[NTX PatientenInformation]` > (Datum + follow_up) ~ 0
  )
)
```

```
model_ntx_cox <- coxph(formula = Surv(time = status_date, event = status) ~ R_age_Datum +
  Geschlecht + `TX Status[NTX PatientenInformation]`,
  data = data_ntx)
```

```
summary(model_ntx_cox)
```

```
## Call:
## coxph(formula = Surv(time = status_date, event = status) ~ R_age_Datum +
##   Geschlecht + `TX Status[NTX PatientenInformation]`, data = data_ntx)
##
##   n= 960, number of events= 317
##
##                                     coef
## R_age_Datum                        0.066615
## Geschlechtweiblich                 -0.229876
## `TX Status[NTX PatientenInformation]`2 - ohne Transplantatfunktion 0.770638
```

```

##                                     exp(coef)
## R_age_Datum                        1.068884
## Geschlechtweiblich                0.794632
## `TX Status[NTX PatientenInformation]`2 - ohne Transplantatfunktion 2.161144
##                                     se(coef)
## R_age_Datum                        0.005387
## Geschlechtweiblich                0.122142
## `TX Status[NTX PatientenInformation]`2 - ohne Transplantatfunktion 0.130542
##                                     z
## R_age_Datum                        12.367
## Geschlechtweiblich                -1.882
## `TX Status[NTX PatientenInformation]`2 - ohne Transplantatfunktion 5.903
##                                     Pr(>|z|)
## R_age_Datum                        < 2e-16 ***
## Geschlechtweiblich                0.0598 .
## `TX Status[NTX PatientenInformation]`2 - ohne Transplantatfunktion 3.56e-09 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
##                                     exp(coef)
## R_age_Datum                        1.0689
## Geschlechtweiblich                0.7946
## `TX Status[NTX PatientenInformation]`2 - ohne Transplantatfunktion 2.1611
##                                     exp(-coef)
## R_age_Datum                        0.9356
## Geschlechtweiblich                1.2584
## `TX Status[NTX PatientenInformation]`2 - ohne Transplantatfunktion 0.4627
##                                     lower .95
## R_age_Datum                        1.0577
## Geschlechtweiblich                0.6255
## `TX Status[NTX PatientenInformation]`2 - ohne Transplantatfunktion 1.6733
##                                     upper .95
## R_age_Datum                        1.080
## Geschlechtweiblich                1.010
## `TX Status[NTX PatientenInformation]`2 - ohne Transplantatfunktion 2.791
##
## Concordance= 0.715 (se = 0.015 )
## Likelihood ratio test= 207.3 on 3 df, p=<2e-16
## Wald test              = 174.7 on 3 df, p=<2e-16
## Score (logrank) test = 186.7 on 3 df, p=<2e-16

```

```
cox.zph(model_ntx_cox)
```

```

##               chisq df      p
## R_age_Datum    3.473  1 0.062
## Geschlecht      0.302  1 0.582
## `TX Status[NTX PatientenInformation]` 0.759  1 0.384
## GLOBAL          4.233  3 0.237

```

all p-values are relatively large, therefore the Null hypothesis of proportional hazards can not be