#### 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"))</pre>
```

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)

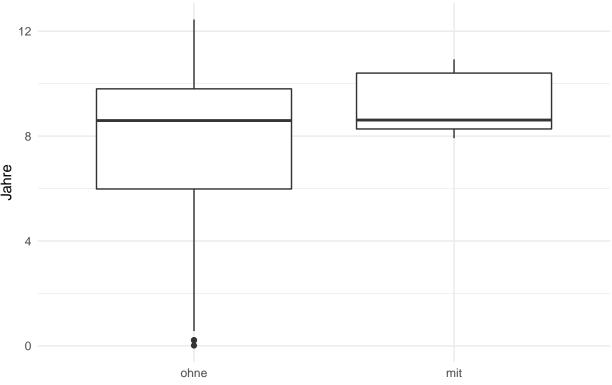
data_iga$mismatch_sum <- as.numeric(as.character(data_iga$`mm-A`)) + as.numeric(as.character(data_iga$`)

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]</pre>
```

#### § 2.1 IGA

```
# follow-up mean / median
tbl_iga_follow_up_mean <- data.frame(</pre>
  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(</pre>
  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_i
  default_theme +
  ggtitle("Boxplot: Follow Up Period") +
  ylab("Jahre") +
  xlab("") +
  scale_x_discrete(labels = c("ohne", "mit"))
```

#### Boxplot: Follow Up Period

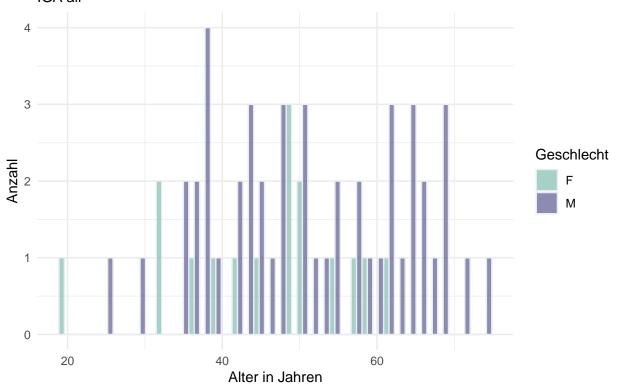


```
# follow_up mean/median (truncated)
tbl_iga_follow_up_mean_truncated <- data.frame(</pre>
 iga_all = mean((interval(data_iga$^T-date`, data_iga$follow_up_truncated) / years(1))),
 iga_pos = mean((interval(data_iga_pos$`T-date`, data_iga_pos$follow_up_truncated) / years(1))),
 iga_neg = mean((interval(data_iga_neg$`T-date`, data_iga_neg$follow_up_truncated) / years(1)))
tbl_iga_follow_up_median_truncated <- data.frame(</pre>
  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(</pre>
 iga all = nrow(data iga[(`T-dls` <= (`T-date` + follow up)) & `Pat death (0=alive, 1= dead)` == 1]),</pre>
  iga_pos = nrow(data_iga_pos[(`T-dls` <= (`T-date` + follow_up)) & `Pat death (0=alive, 1= dead)` == 1</pre>
  iga_neg = nrow(data_iga_neg[(`T-dls` <= (`T-date` + follow_up)) & `Pat death (0=alive, 1= dead)` == 1</pre>
# patient drop out within follow up
tbl_iga_pat_drop <- data.frame(</pre>
 iga_all = nrow(data_iga[(`T-dls` <= (`T-date` + follow_up)) & `Pat death (0=alive, 1= dead)` == 0]),</pre>
 iga_pos = nrow(data_iga_pos[(`T-dls` <= (`T-date` + follow_up)) & `Pat death (0=alive, 1= dead)` == 0</pre>
  iga_neg = nrow(data_iga_neg[('T-dls' <= ('T-date' + follow_up)) & 'Pat death (0=alive, 1= dead)' == 0</pre>
# patients with graft loss
tbl_iga_graft_loss <- data.frame(</pre>
```

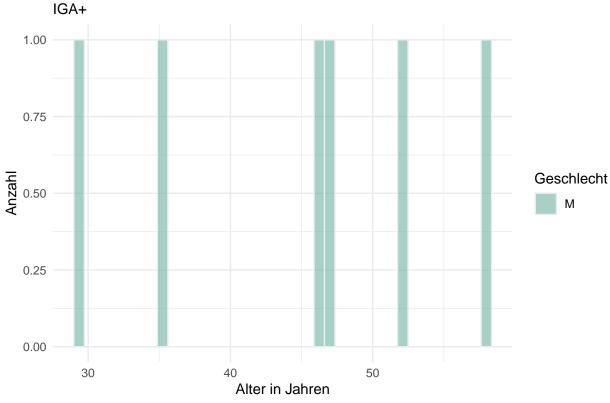
iga\_all = nrow(data\_iga[`graft loss (0=functial, 1=loss)` == 1]),

```
iga_pos = nrow(data_iga_pos[`graft loss (0=functial, 1=loss)` == 1]),
  iga_neg = nrow(data_iga_neg[`graft loss (0=functial, 1=loss)` == 1])
)
# patients with graft loss within follow up period
tbl_iga_graft_loss_follow_up <- data.frame(</pre>
  iga_all = nrow(data_iga[`graft loss date` < `T-date` + follow_up]),</pre>
  iga_pos = nrow(data_iga_pos[`graft loss date` < `T-date` + follow_up]),</pre>
  iga_neg = nrow(data_iga_neg[`graft loss date` < `T-date` + follow_up])</pre>
# 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
```

## Histogram Alter in Jahren am Transplantationsdatum IGA all

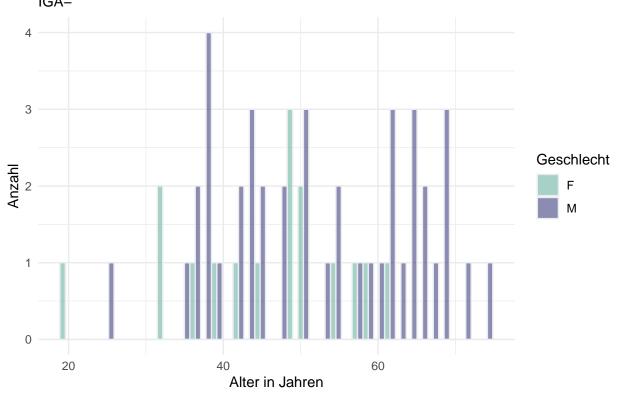


#### Histogram Alter in Jahren am Transplantationsdatum



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

## Histogram Alter in Jahren am Transplantationsdatum IGA-



```
## 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)
)</pre>
```

```
p1 <- ggplot(data = data_iga) +
  geom_boxplot(aes(x = R_age_Tdate)) +
  coord_flip() +</pre>
```

## boxpot

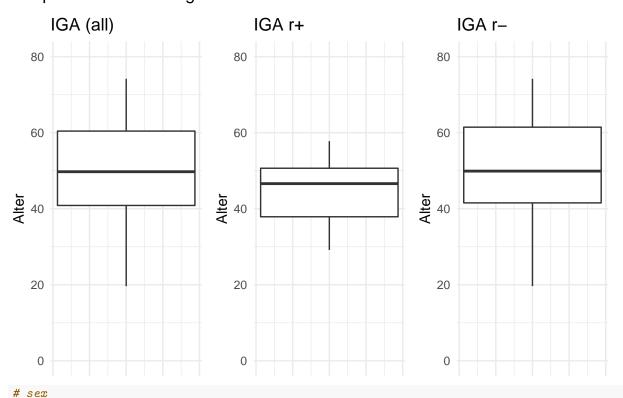
```
ggtitle("IGA (all)") +
default_theme

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()+</pre>
```

#### **Boxplot Altersverteilung**

ggtitle("iga all")



```
p2 <- data.frame(group = rownames(tbl_iga_sex_abs),</pre>
           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</pre>
patch + plot_annotation(title = "Kreisdiagramme Geschlecht (absolut)")
```

#### Kreisdiagramme Geschlecht (absolut)



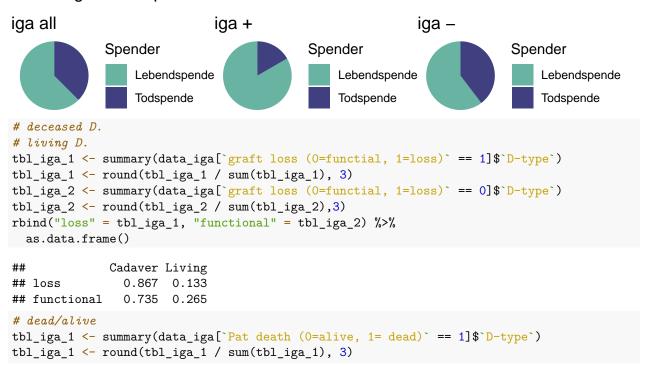
## `stat\_bin()` using `bins = 30`. Pick better value with `binwidth`.

# Histogramm BMI Iga 7.5 Geschlecht F M 2.5 0.0 2e+06 BMI

```
# 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")</pre>
rownames(tbl_death_d) <- c("Lebendspende", "Todspende")</pre>
p1 <- data.frame(group = c("Lebendspende", "Todspende"),</pre>
           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")) +
```

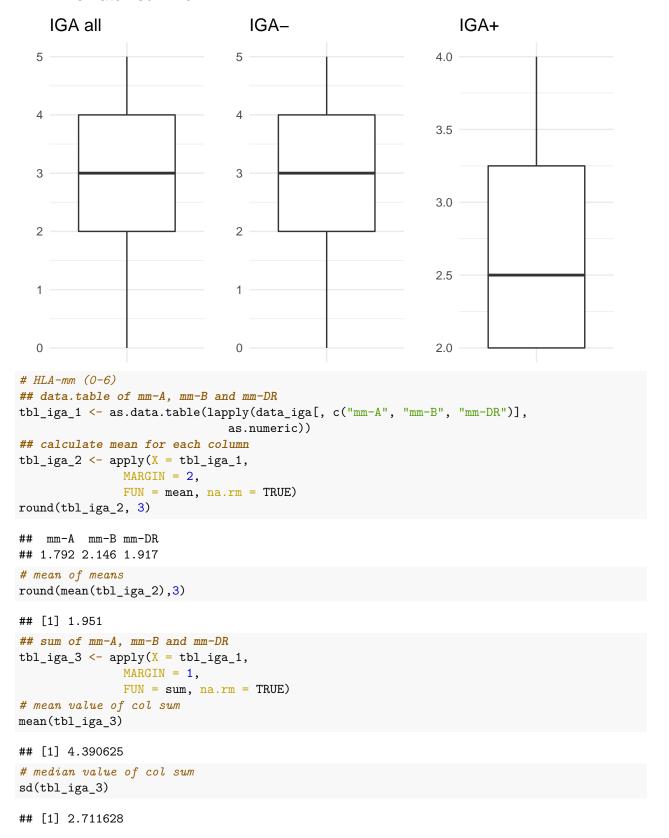
```
theme_void() +
  two scale fill +
  ggtitle("iga all")
p2 <- data.frame(group = c("Lebendspende", "Todspende"),</pre>
           value = c(nrow(data_iga_pos[`Pat death (0=alive, 1= dead)` == 0]),
                      nrow(data_iga_pos[`Pat death (0=alive, 1= dead)` == 1]))) %>%
  ggplot(aes(x = "", 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("Lebendspende", "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</pre>
patch + plot_annotation(title = "Kreisdiagramme Spender")
```

#### Kreisdiagramme Spender



```
tbl_iga_2<- summary(data_iga[`Pat death (0=alive, 1= dead)` == 0]$`D-type`)
tbl_iga_2 <- round(tbl_iga_2 / sum(tbl_iga_2),3) # in %
rbind("dead" = tbl_iga_1, "alive" = tbl_iga_2) %>%
  as.data.frame()
         Cadaver Living
## dead 0.917 0.083
## alive 0.675 0.325
# BMI (mean.)
tbl_iga_bmi <- data.table(</pre>
  iga_all = mean(data_iga$`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</pre>
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



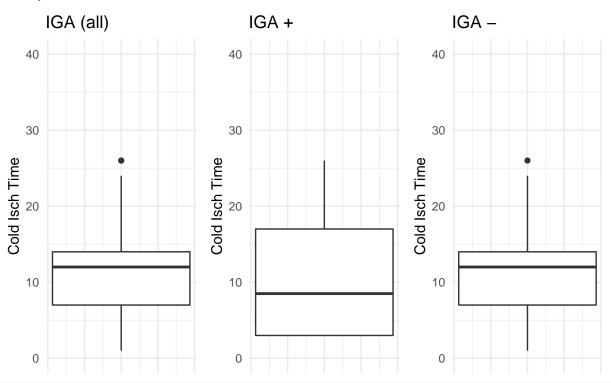
```
# PRA current (mean)
tbl_iga_pra_curr_mean <- data.frame(</pre>
  iga_all = mean(data_iga$`Current PRA%`, na.rm = TRUE),
  iga_pos = mean(data_iga_pos$`Current PRA%`, na.rm = TRUE),
  iga_neg = mean(data_iga_neg$`Current PRA%`, na.rm = TRUE)
# PRA highest (mean)
tbl_iga_pra_high <- data.frame(</pre>
  iga_all = mean(data_iga$`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(</pre>
  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(</pre>
 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$`Cold ischaemic period hours`, na.rm = TRUE)
# standard error
tbl_iga_cis_sd <- data.frame(</pre>
 iga_all = sd(data_iga$`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
```

## 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

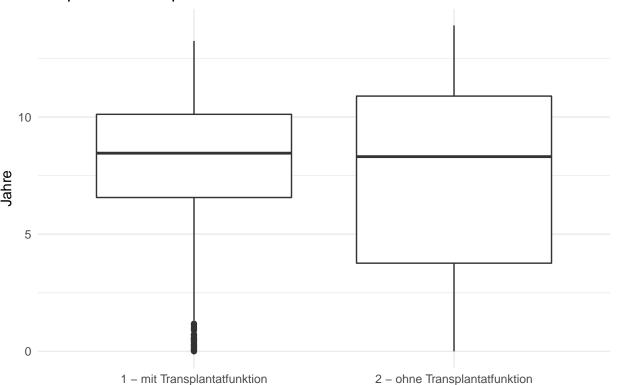
tbl_iga_don_living_abs <- data.frame(
    iga_all = nrow(data_iga[`D-type` == "Living"]),
    iga_pos = nrow(data_iga_pos[`D-type` == "Living"]),
    iga_neg = nrow(data_iga_neg[`D-type` == "Living"])
)</pre>
```

```
tbl_iga_don_dead_abs <- data.frame(</pre>
  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(</pre>
 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(</pre>
 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(</pre>
  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")
```

#### § 2.2 NTX

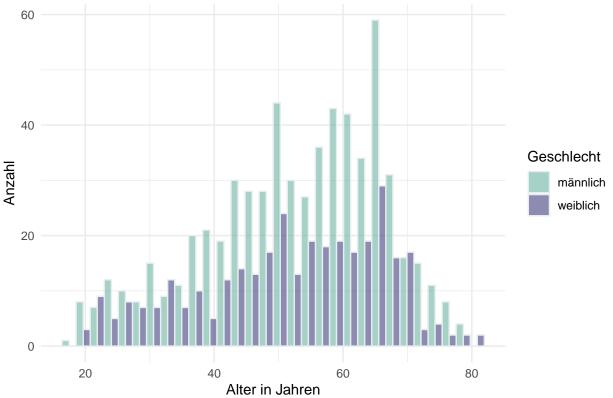
```
# follow_up mean
pmin(
    # follow up
interval(data_ntx$Datum, (data_ntx$Datum + follow_up)) / years(1),
# last seen
```

Merkmal	IgA (all)	IgA r+	IgA r-
dropout	27.000000	3.0000000	24.0000000
graftloss	15.000000	2.0000000	13.0000000
graftloss_followup	14.000000	2.0000000	12.0000000
pat_death	23.000000	1.0000000	22.0000000
$don\_dead\_abs$	49.000000	4.0000000	45.0000000
don_dead_rel	0.765625	0.6666667	0.7758621
don_living_abs	15.000000	2.0000000	13.0000000
don_living_rel	0.234375	0.3333333	0.2241379
cold_h_mean	11.612245	11.5000000	11.6222222
cold_h_sd	6.330535	10.9696551	5.9592895



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

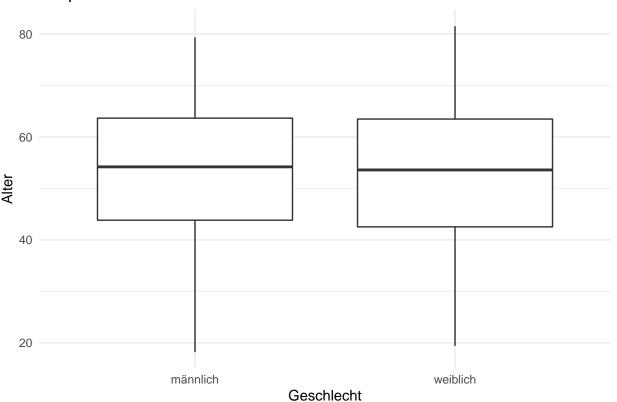
```
## [1] 317
# patient drop out
nrow(data_ntx[`Date last seen[NTX PatientenInformation]` <= (Datum + follow_up) & `Patienten Status[NTX
## [1] 370
# patients with graft loss
nrow(data_ntx$\text{TX Status[NTX PatientenInformation]}\text{ == "2 - ohne Transplantatfunktion"]})
## [1] 181
# age patients (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`.
     Histogram Alter in Jahren am Transplantationsdatum
  60
```



```
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")
```

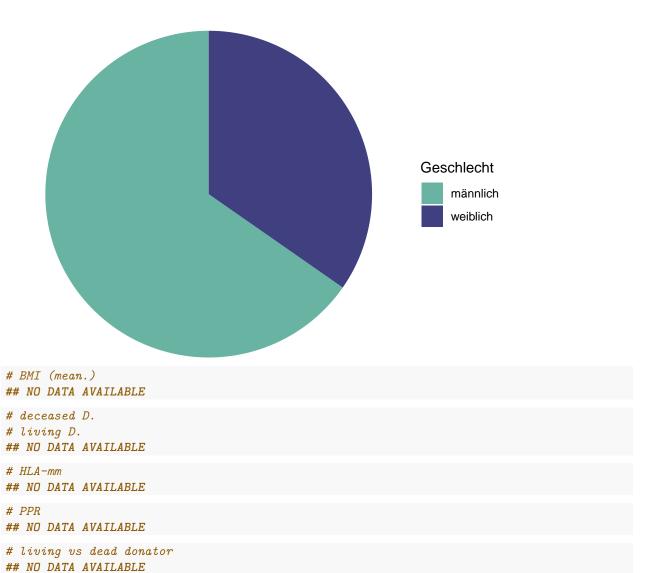
#### **Boxplot Alter NTX**



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

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

#### NTX



#### § 3 Kaplan-Meier

#### § 3.1 IGA

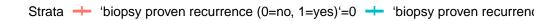
```
# functions in survival and survminer package need numeric-type input
data_iga[, time_date_biopsy := interval(`T-date`, `date of biopsy`) / years(1)]
data_iga[, time_t_dls := interval(`T-date`, `T-dls`) / years(1)]
data_iga[, time_date_birth := interval(`T-date`, `Date of birth`) / years(1)]
data_iga[,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

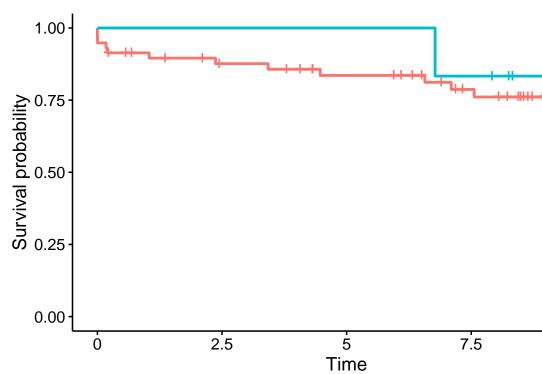
#### § 3.1.1

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

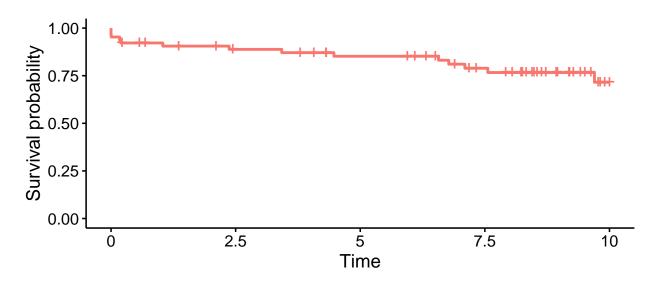
```
model_iga_1 <- survfit(formula = Surv(time = status_date, event = status, type = "right") ~ data_iga$`b
ggsurvplot(model_iga_1)</pre>
```



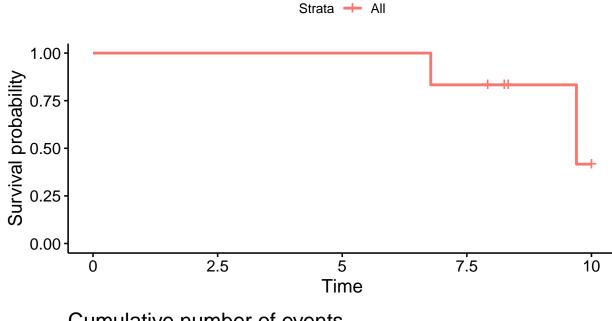


#### ${\bf Overall\ kaplan-Meier\ curve}$

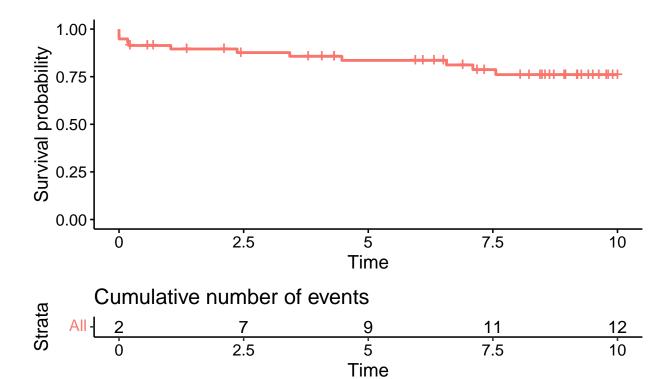




# Cumulative number of events All 2 7 9 12 14 0 2.5 5 7.5 10 Time



# Cumulative number of events All Time

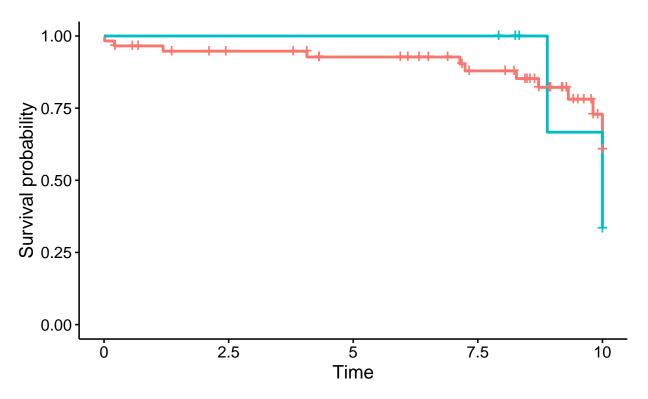


#### § 3.1.2

- 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,</pre>
                                   event = status, type = "right") ~ data_iga$`biopsy proven recurrence
                    data = data_iga)
ggsurvplot(model_iga_2,
           conf.int = FALSE)
```

Strata + 'biopsy proven recurrence (0=no, 1=yes)'=0 + 'biopsy proven recurrence (0=no, 1=yes)'=0



#### § 3.1.3

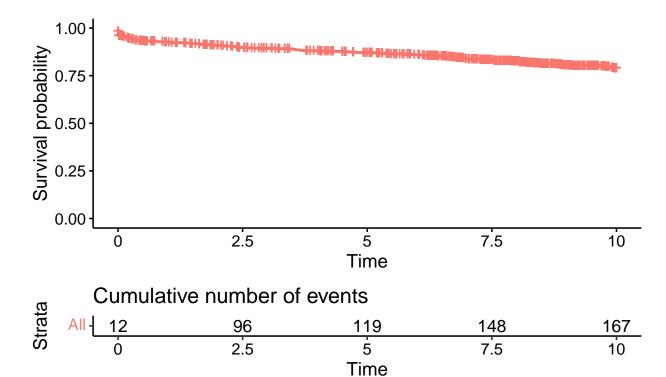
- Event: graft loss and death
- Censoring scheme:
- Time period:

#### § 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

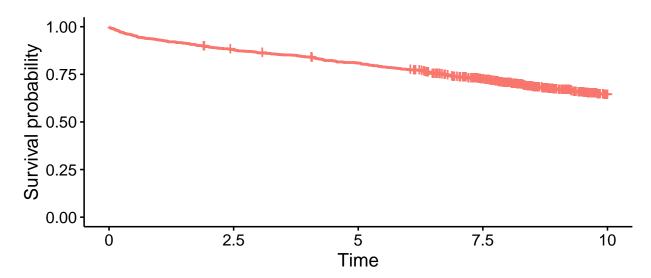
```
data_ntx <- data_ntx %>%
  mutate(status_date = case_when(
    ## patient experienced graft loss
    !is.na(Transplantatfunktionsende) & Transplantatfunktionsende <= (Datum + follow_up) ~ interval(Dat
    ## patient died within follow up
    `Todesdatum[NTX PatientenInformation]` < (Datum + follow_up) ~ interval(Datum, `Todesdatum[NTX Pati
    ## patiend last seen within follow up
    `Date last seen[NTX PatientenInformation]` < (Datum + follow_up) ~ interval(Datum, `Date last seen[
    ## 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,
   ## patiend 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,</pre>
                                       event = status, type = "right") ~ 1,
                       data = data_ntx)
ggsurvplot(model_ntx_1,
           conf.int = FALSE,
           cumevents = TRUE)
```

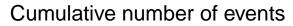


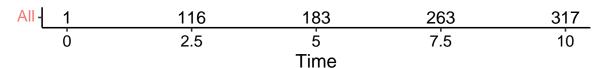
#### § 3.2.2

- 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_
  )
)
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
```







#### § 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,</pre>
    ## 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,</pre>
    ## else
   TRUE ~ 0
 ))
data_iga_pos <- data_iga[`biopsy proven recurrence (0=no, 1=yes)` == 1]</pre>
data_iga_neg <- data_iga[`biopsy proven recurrence (0=no, 1=yes)` == 0]</pre>
model_iga_cox <- coxph(data = data_iga, formula = Surv(time = as.numeric(status_date),</pre>
                                       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)
                                         -1.101e-02 9.891e-01 3.584e-02 -0.307
## R_age_Tdate
## data_iga$`R-sex`M
                                          1.284e-01 1.137e+00 8.645e-01 0.149
## data_iga$`D-type`Living
                                                            NA 0.000e+00
                                                 NA
                                                                              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|)
                                            0.759
## R_age_Tdate
## 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
                                         9.643e-01 1.037e+00
## data_iga$`mm-A`1
                                                                 0.1988
                                                                            4.677
## data_iga$`mm-A`2
                                          1.100e-08 9.091e+07
                                                               0.0000
                                                                              Inf
## data_iga$`Cold ischaemic period hours` 9.949e-01 1.005e+00
                                                                 0.8709
                                                                            1.137
                                          1.041e+00 9.603e-01 0.9838
                                                                            1.102
## data_iga$`D-age`
## Concordance= 0.701 (se = 0.095)
## Likelihood ratio test= 3.73 on 6 df.
                                          p = 0.7
                       = 2.19 on 6 df,
## Wald test
                                          p = 0.9
## Score (logrank) test = 3.02 on 6 df,
                                          p = 0.8
cox.zph(model_iga_cox)
```

 $\hbox{\tt \#\#}\qquad \qquad \hbox{\tt chisq df} \qquad p$ 

```
## 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_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_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
```

## `TX Status[NTX PatientenInformation]`2 - ohne Transplantatfunktion 0.770638

0.066615

-0.229876

## R\_age\_Datum

## Geschlechtweiblich

```
## R_age_Datum
                                                                       0.005387
## Geschlechtweiblich
                                                                       0.122142
## `TX Status[NTX PatientenInformation]`2 - ohne Transplantatfunktion 0.130542
##
## 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 '**' 0.05 '.' 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
## R_age_Datum
                                         3.473 1 0.062
                                         0.302 1 0.582
## Geschlecht
## `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
```

exp(coef)

1.068884

0.794632

2.161144 se(coef)

##

## R\_age\_Datum

## Geschlechtweiblich

## `TX Status[NTX PatientenInformation] `2 - ohne Transplantatfunktion