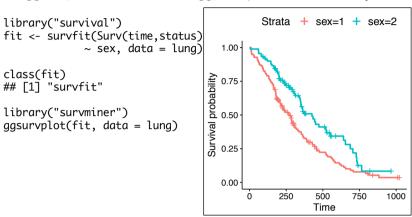
# **Creating Survival Plots**

## **Informative and Elegant** with survminer

#### **Survival Curves**

The ggsurvplot() function creates ggplot2 plots from survfit objects.



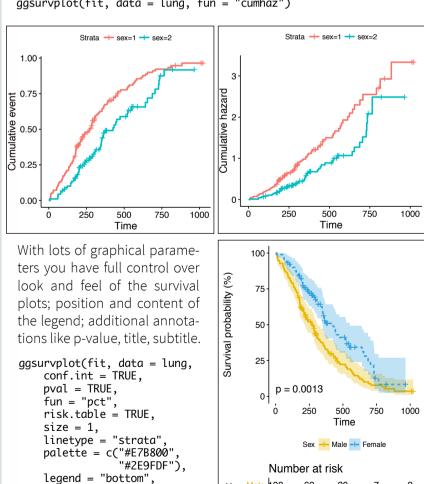
Use the **fun** argument to set the transformation of the survival curve. E.g. "event" for cumulative events, "cumhaz" for the cumulative hazard function or "pct" for survival probability in percentage.

```
ggsurvplot(fit, data = lung, fun = "event")
ggsurvplot(fit, data = lung, fun = "cumhaz")
```

legend.title = "Sex"

legend.labs = c("Male"

"Female"))



e 438

90

62

53

250

20

500

Time

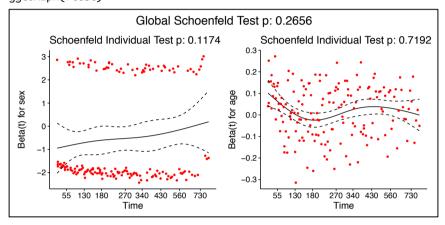
750

1000

## **Diagnostics of Cox Model**

The function cox.zph() from survival package may be used to test the proportional hazards assumption for a Cox regression model fit. The graphical verification of this assumption may be performed with the function **ggcoxzph()** from the **survminer** package. For each covariate it produces plots with scaled Schoenfeld residuals against the time.

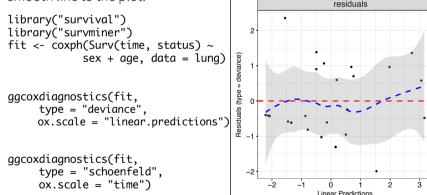
```
library("survival")
fit <- coxph(Surv(time, status) ~ sex + age, data = lung)
ftest <- cox.zph(fit)</pre>
ftest
              rho chisa
## sex
           0.1236 2.452 0.117
          -0.0275 0.129 0.719
## GLOBAL
               NA 2.651 0.266
library("survminer")
gqcoxzph(ftest)
```

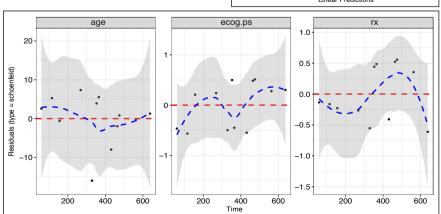


The function **ggcoxdiagnostics()** plots different types of residuals as a function of time, linear predictor or observation id. The type of residual is selected with type argument. Possible values are "martingale", "deviance", "score", "schoenfeld", "dfbeta", "dfbetas", and "scaledsch".

The **ox.scale** argument defines what shall be plotted on the OX axis. Possible values are "linear.predictions", "observation.id", "time".

Logical arguments hline and sline may be used to add horizontal line or smooth line to the plot.





### **Summary of Cox Model**

The function **ggforest()** from the **survminer** package creates a forest plot for a Cox regression model fit. Hazard ratio estimates along with confidence intervals and p-values are plotter for each variable.

```
library("survival")
library("survminer")
lung$age <- ifelse(lung$age > 70, ">70","<= 70")</pre>
fit <- coxph( Surv(time, status) ~ sex + ph.ecog + age, data = lung)</pre>
## coxph(formula = Surv(time, status) ~ sex+ph.ecog+age, data=lung)
##
##
              coef exp(coef) se(coef)
                                 0.168 -3.37 0.00075
## sex
            -0.567
                        0.567
## ph.ecog 0.470
                        1.600
                                 0.113 4.16 3.1e-05
## aae>70
            0.307
                       1.359
                                 0.187
                                         1.64 0.10175
                                               Forest plot for coxph model
## Likelihood ratio test=31.6 on
## n= 227, number of events= 164
                                                     0.57 (p.value 0.00075***)
ggforest(fit)
                                                              1.6 (p.value 0.000031**
                                         ph.ecoa
                                                                 1.4 (p.value 0.1)
                                         age>70
```

The function **ggcoxadjustedcurves()** from the **survminer** package plots Adjusted Survival Curves for Cox Proportional Hazards Model. Adjusted Survival Curves show how a selected factor influences survival estimated from a Cox model.

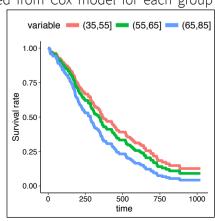
Note that these curves differ from Kaplan Meier estimates since they present expected ssurvival based on given Cox model.

```
library("survival")
                                                    variable Male Female
library("survminer")
lung$sex <- ifelse(lung$sex == 1,</pre>
                     "Male", "Female")
                                           0.75
fit <- coxph(Surv(time, status) ~</pre>
                sex + ph.ecog + age,
                                          <u>R</u> 0.50 ⋅
               data = luna)
ggcoxadjustedcurves(fit, data=lung,
                                           0.25
               variable=lung$sex)
```

Note that it is not necessary to include the grouping factor in the Cox model. Survival curves are estimated from Cox model for each group

defined by the factor independently. lung\$age3 <- cut(lung\$age,</pre> c(35,55,65,85)) ggcoxadjustedcurves(fit, data=lung,

variable=lung\$age3)



Hazard ratio