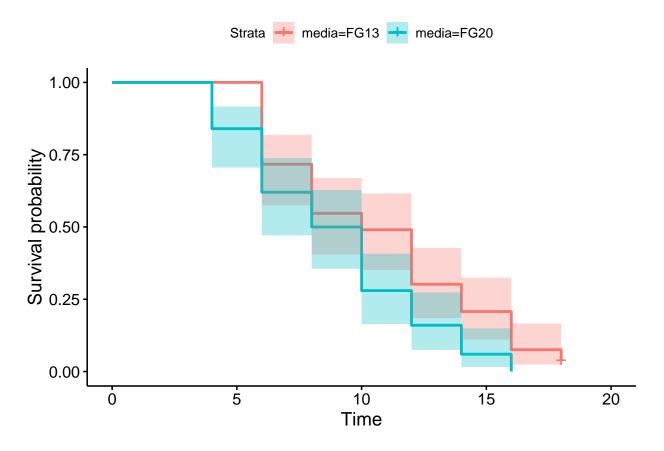
Metabolite

Jonas Gehrlein 29 nov 2018

```
data_met$Survobj <- with(data_met,Surv(data_met$Time, event = data_met$Status))
km <- survfit(Survobj ~ media, data = data_met,conf.type = "log-log", error = "greenwood")
s_km <- summary(km)
ggsurvplot(km, conf.int = TRUE)</pre>
```



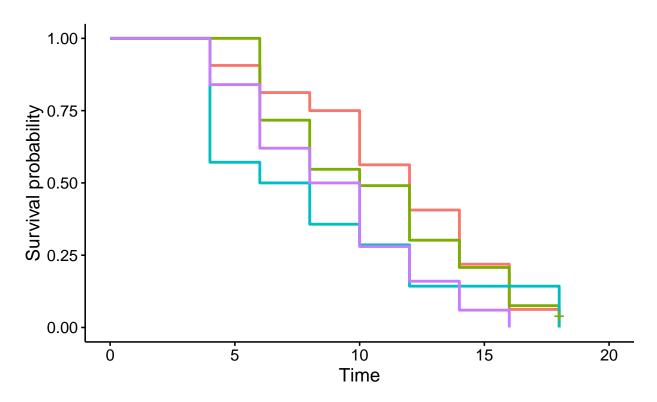
```
fit <- coxph(Survobj ~ media, data = data_met)
summary(fit)</pre>
```

```
## Call:
## coxph(formula = Survobj ~ media, data = data_met)
```

```
##
    n= 103, number of events= 101
##
##
              coef exp(coef) se(coef)
##
                                          z Pr(>|z|)
## mediaFG20 0.5178
                      1.6784
                             0.2061 2.513
                                               0.012 *
## mediaOP50
                NA
                          NA
                               0.0000
                                         NA
                                                  NA
## Signif. codes: 0 '***' 0.001 '**' 0.05 '.' 0.1 ' ' 1
##
##
            exp(coef) exp(-coef) lower .95 upper .95
## mediaFG20
                1.678
                          0.5958
                                     1.121
## mediaOP50
                                        NA
                                                  NA
                   NA
                              NA
## Concordance= 0.571 (se = 0.031)
                  (max possible= 0.999 )
## Rsquare= 0.059
## Likelihood ratio test= 6.28 on 1 df,
                                          p=0.01
## Wald test
                       = 6.31 on 1 df,
                                          p=0.01
## Score (logrank) test = 6.44 on 1 df,
                                          p=0.01
```

Alo lower survival for FG20 if only including survival and higher survival for OP50 which is higher than if we had compared with fly gut which isn't only metabolites. If we include metabolites in a cox model without OP50

```
data_2_fly <- subset(data_2, media != 'OP50' )
data_2_fly$Survobj <- with(data_2_fly,Surv(data_2_fly$Time, event = data_2_fly$Status))
km_fly <- survfit(Survobj ~ media + Metabolite, data = data_2_fly,conf.type = "log-log", error = "green ggsurvplot(km_fly)</pre>
```



No big difference on the plot between metabolite or without if we fit a model of both media and metabolite

```
fit_fly_meta <- coxph(Survobj ~ media + Metabolite , data = data_2_fly)
summary(fit_fly_meta)</pre>
```

```
## Call:
## coxph(formula = Survobj ~ media + Metabolite, data = data_2_fly)
##
    n= 149, number of events= 147
##
##
##
                 coef exp(coef) se(coef)
                                             z Pr(>|z|)
                                0.17022 2.915 0.00356 **
## mediaFG20 0.49618
                        1.64243
## mediaOP50
                   NA
                                0.00000
                                            NA
## Metabolite 0.08789
                        1.09186 0.18127 0.485 0.62778
## ---
## Signif. codes: 0 '***' 0.001 '**' 0.05 '.' 0.1 ' ' 1
##
##
              exp(coef) exp(-coef) lower .95 upper .95
## mediaFG20
                  1.642
                            0.6089
                                      1.1765
                                                 2.293
## mediaOP50
                     NA
                                NA
                                                    NA
                                          NA
## Metabolite
                  1.092
                            0.9159
                                      0.7654
                                                 1.558
##
## Concordance= 0.59 (se = 0.027)
## Rsquare= 0.059
                   (max possible= 1 )
## Likelihood ratio test= 9.07 on 2 df,
                                           p=0.01
## Wald test
                        = 9.31 on 2 df,
                                           p=0.01
```

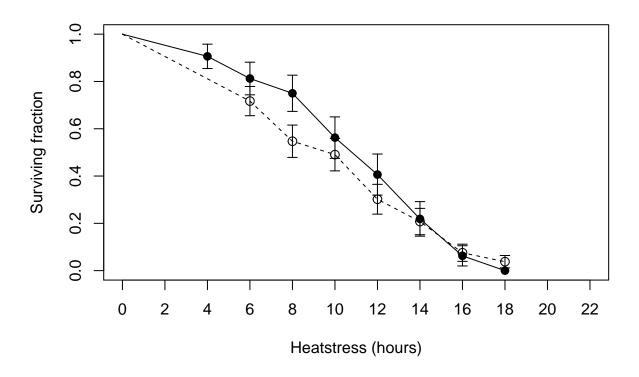
```
## Score (logrank) test = 9.5 on 2 df, p=0.009
AIC(fit_fly_meta)
## [1] 1193.565
fit_fly_med <- coxph(Survobj ~ media , data = data_2_fly)</pre>
summary(fit_fly_med)
## Call:
## coxph(formula = Survobj ~ media, data = data_2_fly)
##
##
    n= 149, number of events= 147
##
##
               coef exp(coef) se(coef)
                                            z Pr(>|z|)
## mediaFG20 0.5080
                       1.6619
                                0.1684 3.016 0.00256 **
## mediaOP50
                 NΑ
                           NA
                                0.0000
                                           NA
                                                    NΑ
## Signif. codes: 0 '***' 0.001 '**' 0.05 '.' 0.1 ' ' 1
##
             exp(coef) exp(-coef) lower .95 upper .95
                           0.6017
## mediaFG20
                 1.662
                                       1.195
## mediaOP50
                    NA
                                NA
                                          NA
                                                    NA
## Concordance= 0.586 (se = 0.025)
## Rsquare= 0.058
                    (max possible= 1 )
## Likelihood ratio test= 8.83 on 1 df,
                                            p=0.003
## Wald test
                        = 9.1 \text{ on } 1 \text{ df},
                                           p=0.003
## Score (logrank) test = 9.28 on 1 df,
                                           p=0.002
AIC(fit_fly_med)
```

[1] 1191.803

It certainly isn't a good predictor in the model. and is centered around 0. But very uncertain about the exact size on the size of the predictor. The only real difference is for FG13 where metabolite has worse survival

```
km_meta <- survfit(Survobj ~ media, data = data_meta,conf.type = "log-log", error = "greenwood")
s_km_meta <- summary(km_meta)</pre>
df_{13}_{meta} \leftarrow data.frame(c(0,s_km_meta$time[1:7]),c(1,s_km_meta$surv[1:7]),
                        c(0,s_km_meta$std.err[1:7]))
colnames(df_13_meta) <- c('Time', 'Surv', 'Std.error')</pre>
df_20_meta <- data.frame(c(0,s_km_meta$time[8:14]),c(1,s_km_meta$surv[8:14]),
                        c(0,s km meta\$std.err[8:14]))
colnames(df_20_meta) <- c('Time', 'Surv', 'Std.error')</pre>
plot(df_13_bac$Time[2:9],df_13_bac$Surv[2:9], pch = 16, cex = 1.2, xlab = 'Heatstress (hours)',
     ylab = 'Surviving fraction',
     main = expression('Comparision between fly gut 20 without and with living bacteria'),
     xlim = c(0,22), ylim = c(0,1), xaxp = c(0,22,11))
lines(df_13_bac$Time,df_13_bac$Surv)
arrows(df_13_bac$Time, df_13_bac$Surv-df_13_bac$Std.error, df_13_bac$Time,
       df_13_bac$Surv+df_13_bac$Std.error, length=0.05, angle=90, code=3, col = 'black')
## Warning in arrows(df_13_bac$Time, df_13_bac$Surv - df_13_bac$Std.error, :
## zero-length arrow is of indeterminate angle and so skipped
points(df_13_meta$Time[2:9],df_13_meta$Surv[2:9], pch = 1, cex = 1.2)
lines(df_13_meta$Time,df_13_meta$Surv, lty = 2)
arrows(df_13_meta$Time, df_13_meta$Surv-df_13_meta$Std.error, df_13_meta$Time, df_13_meta$Surv+df_13_me
## Warning in arrows(df_13_meta$Time, df_13_meta$Surv -
## df 13 meta$Std.error, : zero-length arrow is of indeterminate angle and so
## skipped
```

Comparision between fly gut 20 without and with living bacteria



df_20_meta\$Std.error, : zero-length arrow is of indeterminate angle and so

skipped

Comparision between fly gut 20 without and with living bacteria

