

FRET Classifier and Validation

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Load patient data

```
load("COIN_Final.Rdata")
```

Encode, Impute Covariates

Isolate those covariates known to be needed for Risk Signatures etc. Impute by replacing NA by the average of other patients.

```
encoded_data <- with(patient_data, data.frame(SUMLES, WBC, NEUT, SHB0, FRET, FRET.cohort))

encoded_data$RAS.Mutation <- ifelse(patient_data$RAS=="Mutation", 1, 0)
encoded_data$mlivonly <- ifelse(patient_data$mlivonly=="Yes", 1, 0)
encoded_data$PIK3CA.Mutation <- ifelse(patient_data$PIK3CA=="Mutation", 1, 0)

# Get means from the 398
df <- encoded_data[encoded_data$FRET.cohort==1,]
df <- subset(df, select = -c(FRET.cohort))
trained_means <- colMeans(df, na.rm = T)

# remove FRET.cohort col
encoded_data <- subset(encoded_data, select = -c(FRET.cohort))

# Apply to all data
#encoded_data[] <- lapply(encoded_data, NA2mean)
encoded_data[] <- replaceNA(encoded_data, trained_means)
```

Load risk signature classifier

```
Class_Predictor <-
  read.table("SP_Classifier_Score/COIN_withFRET_July16/RiskScore_formula.txt",
            sep = '*', col.names = c("Weight", "Covariate"),
            stringsAsFactors = F, skip = 2, fill = T)

kable(Class_Predictor)
```

Weight	Covariate
S=(0.006338408651)	SUMLES
+ (-0.142238099638)	WBC
+ (0.211925854316)	NEUT
+ (0.769846490411)	SHB0
+ (9.895611265279)	FRET
+ (-0.728162098653)	mlivonly
+ (-1.151128618935)	PIK3CA.Mutation
- (0.697856603639)	

```

# Load output from Saddle Point Signature and tidy up
Class_Predictor_Betas <-
  read.table("SP_Classifier_Score/COIN_withFRET_July16/SETCV_L2/betas_with_errors_optimised.txt",
    sep = ' ', col.names = c("Covariate", "beta", "pm", "error",
                             "rank", "z.score", "p.value"),
    stringsAsFactors = F)

Class_Predictor_Betas <- subset(Class_Predictor_Betas, select=-pm)
Class_Predictor_Betas$Covariate <- sub(":", "", Class_Predictor_Betas$Covariate)
Class_Predictor_Betas$beta <- as.numeric(
  sub("^beta\\[[\\d+]=", "", Class_Predictor_Betas$beta))
Class_Predictor_Betas$rank <- last_number(Class_Predictor_Betas$rank,
  decimals = F, negs = F)
Class_Predictor_Betas$z.score <- last_number(Class_Predictor_Betas$z.score,
  decimals = T, negs = F)
Class_Predictor_Betas$p.value <- last_number(Class_Predictor_Betas$p.value,
  decimals = T, negs = F)

# Create new cols as required
Class_Predictor_Betas$description <- Class_Predictor_Betas$Covariate
Class_Predictor_Betas$description <-
  sub("SUMLES", "Sum of longest diameter", Class_Predictor_Betas$description)
Class_Predictor_Betas$description <-
  sub("WBC", "White blood cell count", Class_Predictor_Betas$description)
Class_Predictor_Betas$description <-
  sub("NEUT", "Neutrophil count", Class_Predictor_Betas$description)
Class_Predictor_Betas$description <-
  sub("SHBO", "Haemaglobin (CTC grade)", Class_Predictor_Betas$description)
#Class_Predictor_Betas$description <-
# sub("FRET", "HER2-HER3 FRET efficiency", Class_Predictor_Betas$description)
Class_Predictor_Betas$description <-
  sub("mlivonly", "Liver-only metastases", Class_Predictor_Betas$description)
Class_Predictor_Betas$description <-
  sub("PIK3CA.Mutation", "PIK3CA mutation", Class_Predictor_Betas$description)

Class_Predictor_Betas$HR <- exp(Class_Predictor_Betas$beta)
Class_Predictor_Betas$HR.U95.CI <- exp(Class_Predictor_Betas$beta+Class_Predictor_Betas$error)
Class_Predictor_Betas$HR.L95.CI <- exp(Class_Predictor_Betas$beta-Class_Predictor_Betas$error)

# Add the constant as new row
Class_Predictor_Betas <- rbind(Class_Predictor_Betas,
  list("Constant", NA, NA, NA, NA, 0, "Constant", 1, 1, 1))

Class_Predictor_Betas$Weight <- Class_Predictor$Weight
Class_Predictor_Betas$Weight <- str_remove(Class_Predictor_Betas$Weight, "S=\\(")
Class_Predictor_Betas$Weight <- str_remove(Class_Predictor_Betas$Weight, "\\+ \\(")
Class_Predictor_Betas$Weight <- sub("\\- \\(", "-", Class_Predictor_Betas$Weight)
Class_Predictor_Betas$Weight <- str_remove(Class_Predictor_Betas$Weight, "\\)")
Class_Predictor_Betas$Weight <- as.numeric(Class_Predictor_Betas$Weight)

mydf <- data.frame(
  Covariate = Class_Predictor_Betas$description,
  #HazardRatio = signif(Class_Predictor_Betas$HR, digits = 3),

```

```

#HazardLower = signif(Class_Predictor_Betas$HR.L95.CI, digits = 3),
#HazardUpper = signif(Class_Predictor_Betas$HR.U95.CI, digits = 3),
# Pvalue = signif(Class_Predictor_Betas$p.value, digits = 2),
Pvalue = apply(Class_Predictor_Betas$p.value, JNCI_pvals),
Rank = Class_Predictor_Betas$rank,
Weight = round(Class_Predictor_Betas$Weight, digits = 5),
stringsAsFactors=FALSE
)

mydf <- mydf[order(mydf$Rank),]

#plotHRTTable(mydf, useClass = F, useWeight = T, useRank = T, xmin = -4, xmax = 5,
#             cex = 1.0, rowspacing = 0.5, col = 1,
#             hrTicks = c(0.5, 1, 2),
#             classpos = -0.8, hrpos = 1.6, pvaluepos = 3.4,
#             weightpos = 4.47, rankpos = -0.7)

# Put blank entries for the constant
mydf$Pvalue[8] = ""
mydf$Rank[8] = ""

# Reorder cols
mydf <- subset(mydf, select = c(Covariate, Rank, Pvalue, Weight))

kable(mydf, row.names=F, col.names = c("Covariate", "Rank", "p-value", "Weight"), booktabs = T) %>%
  kable_styling(latex_options = "striped")

```

Covariate	Rank	p-value	Weight
Sum of longest diameter	1	<0.001***	0.00634
Neutrophil count	2	0.01*	0.21193
White blood cell count	3	0.04*	-0.14224
Haemaglobin (CTC grade)	4	<0.001***	0.76985
PIK3CA mutation	5	<0.001***	-1.15113
Liver-only metastases	6	<0.001***	-0.72816
FRET	7	0.003**	9.89561
Constant			-0.69786

Calculate Risk Score

```
patient_data$Class_Pred_Score <- calculateRiskScore(Class_Predictor, encoded_data)
```

ROC

ROC curve for the 398 training data, these are the ones that we have ground truth FRET OS Class for.

```
library(pROC)
```

```
## Warning: package 'pROC' was built under R version 3.5.3
```

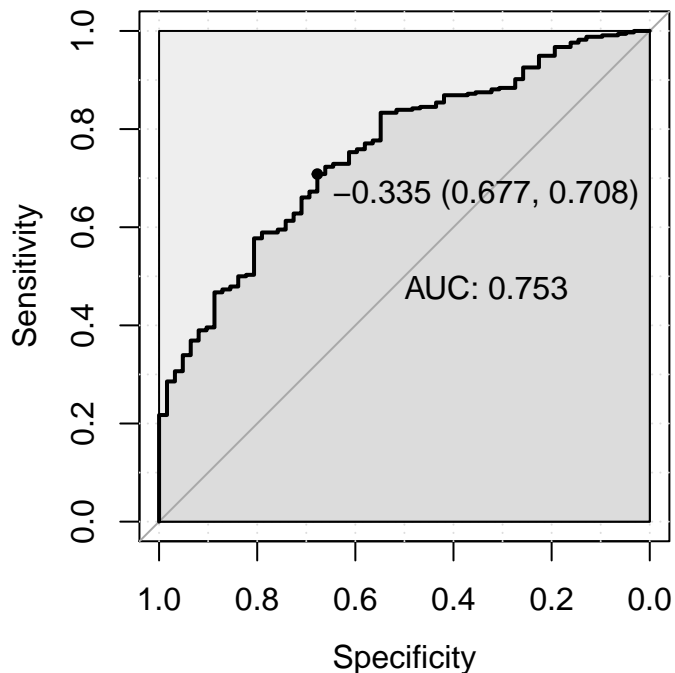
```
## Type 'citation("pROC")' for a citation.
```

```
##
## Attaching package: 'pROC'

## The following objects are masked from 'package:stats':
##
##      cov, smooth, var

pROC_obj <- roc(patient_data$Class.FRET.OS, patient_data$Class_Pred_Score,
  smoothed = TRUE,
  # arguments for ci
  ci=FALSE, ci.alpha=0.9, stratified=FALSE,
  # arguments for plot
  plot=TRUE, auc.polygon=TRUE, max.auc.polygon=TRUE, grid=TRUE,
  print.auc=TRUE, print.thres=TRUE)

## Setting levels: control = 1, case = 2
## Setting direction: controls < cases
```



Choose an optimal threshold

```
optimal <- coords(pROC_obj, x="best", input="threshold", best.method="youden")

## Warning in coords.roc(pROC_obj, x = "best", input = "threshold",
## best.method = "youden"): An upcoming version of pROC will set the
## 'transpose' argument to FALSE by default. Set transpose = TRUE explicitly
## to keep the current behavior, or transpose = FALSE to adopt the new one
## and silence this warning. Type help(coords_transpose) for additional
## information.
```

```
optimal
```

```
##   threshold specificity sensitivity  
## -0.3347168   0.6774194   0.7083333  
threshold = optimal[1]
```

Classify all patients

```
patient_data$Pred.Class <- ifelse(patient_data$Class_Pred_Score>threshold, 2, 1)
```

Confusion table for the 398

```
GT <- patient_data$Class.FRET.OS  
Pred <- patient_data$Pred.Class
```

```
table(GT, Pred)
```

```
##      Pred  
## GT      1      2  
##   1  42  20  
##   2  98 238
```

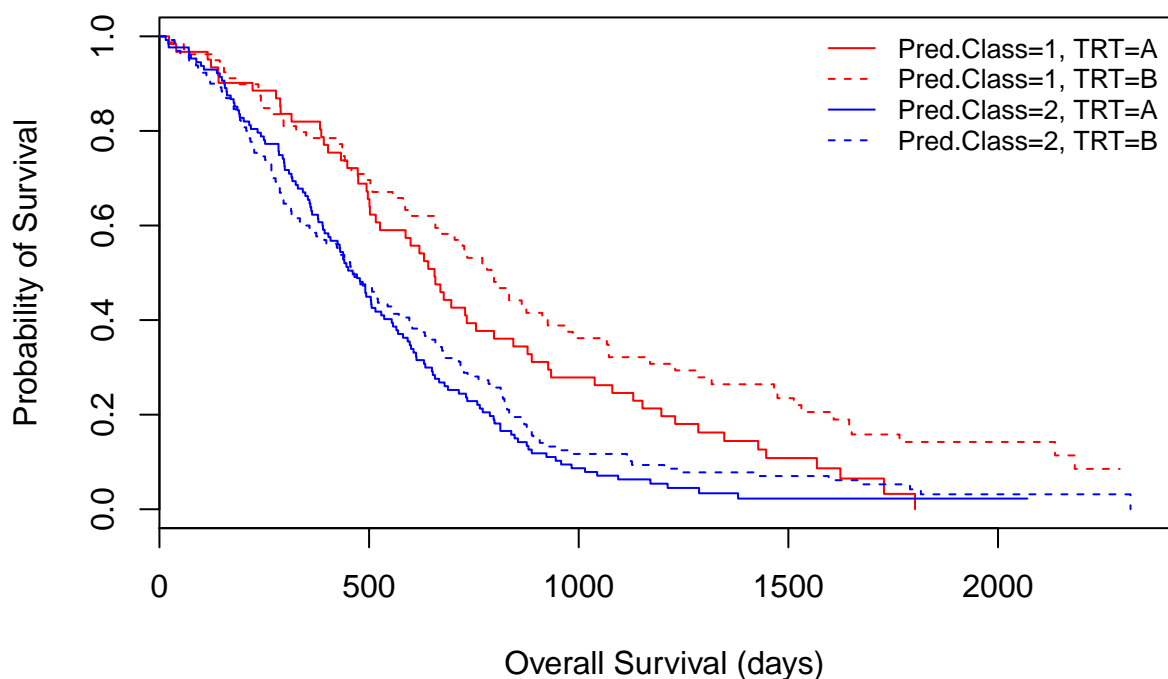
Analyse 398 training set

```
data <- patient_data[patient_data$FRET.cohort==1,]
```

```
library(survival)
```

```
data$SurvObj.pfs <- with(data, Surv(pfstime, pfsevent))  
data$SurvObj.os <- with(data, Surv(ostime, osevent))  
col=c("red", "red", "blue", "blue")  
lty=c(1,2,1,2)
```

```
km <- survfit(SurvObj.os ~ Pred.Class + TRT, data=data)  
plot(km, col=col, xlab="Overall Survival (days)", ylab="Probability of Survival", lty = lty)  
legend("topright", col=col, legend = names(km$strata), lty = lty, bty="n", cex=0.8)
```



```
print(km)
```

```
## Call: survfit(formula = SurvObj.os ~ Pred.Class + TRT, data = data)
```

```
##
```

	n	events	median	0.95LCL	0.95UCL
Pred.Class=1, TRT=A	61	58	656	516	844
Pred.Class=1, TRT=B	79	67	792	658	975
Pred.Class=2, TRT=A	128	123	461	403	536
Pred.Class=2, TRT=B	130	125	461	374	571

```
survdifff(SurvObj.os ~ Pred.Class, data=data)
```

```
## Call:
```

```
## survdifff(formula = SurvObj.os ~ Pred.Class, data = data)
```

```
##
```

	N	Observed	Expected	(O-E) ² /E	(O-E) ² /V
Pred.Class=1	140	125	175	14.5	28.6
Pred.Class=2	258	248	198	12.9	28.6

```
##
```

```
## Chisq= 28.6 on 1 degrees of freedom, p= 9e-08
```

```
survdifff(SurvObj.os[Pred.Class==1] ~ TRT[Pred.Class==1], data=data)
```

```
## Call:
```

```
## survdifff(formula = SurvObj.os[Pred.Class == 1] ~ TRT[Pred.Class ==  
## 1], data = data)
```

```
##
```

	N	Observed	Expected	(O-E) ² /E	(O-E) ² /V

```
## TRT[Pred.Class == 1]=A 61      58      47.2      2.46      4.07
## TRT[Pred.Class == 1]=B 79      67      77.8      1.49      4.07
##
## Chisq= 4.1 on 1 degrees of freedom, p= 0.04
```

```
survdifff(SurvObj.os[Pred.Class==2] ~ TRT[Pred.Class==2], data=data)
```

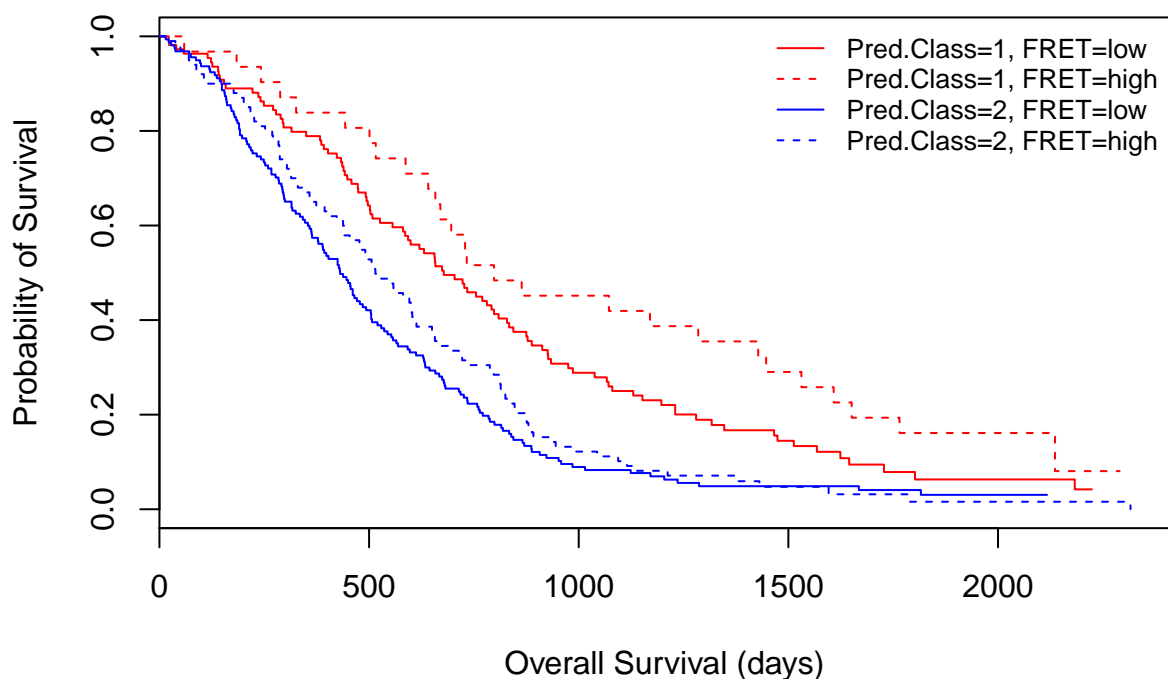
```
## Call:
## survdiff(formula = SurvObj.os[Pred.Class == 2] ~ TRT[Pred.Class ==
##      2], data = data)
##
##              N Observed Expected (O-E)^2/E (O-E)^2/V
## TRT[Pred.Class == 2]=A 128      123      116      0.439      0.842
## TRT[Pred.Class == 2]=B 130      125      132      0.385      0.842
##
## Chisq= 0.8 on 1 degrees of freedom, p= 0.4
```

```
ggrisktable(km, data=data) + theme_cleantable()
```

Number at risk					
Pred.Class=1, TRT=A	61	40	17	6	0
Pred.Class=1, TRT=B	79	55	27	16	8
Pred.Class=2, TRT=A	128	57	11	2	1
Pred.Class=2, TRT=B	130	61	15	9	2

```
#Split FRET by Tertiles
breaks <- quantile(data$FRET, probs = c(0.0, 0.67, 1.0), na.rm = T)
data$FRETeff <- cut(data$FRET, breaks = breaks, labels = c("low", "high"),
                    include.lowest = T)
data$FRET_raw <- data$FRET
data$FRET <- data$FRETeff

km <- survfit(SurvObj.os ~ Pred.Class + FRET, data=data)
plot(km, col=col, lty=lty, xlab="Overall Survival (days)", ylab="Probability of Survival")
legend("topright", col=col, legend = names(km$strata), lty=lty, bty="n", cex=0.8)
```



```
print(km)
```

```
## Call: survfit(formula = SurvObj.os ~ Pred.Class + FRET, data = data)
```

```
##
```

	n	events	median	0.95LCL	0.95UCL
## Pred.Class=1, FRET=low	109	98	678	586	829
## Pred.Class=1, FRET=high	31	27	798	658	1531
## Pred.Class=2, FRET=low	158	151	431	378	504
## Pred.Class=2, FRET=high	100	97	515	441	613

```
survdifff(SurvObj.os ~ Pred.Class, data=data)
```

```
## Call:
```

```
## survdifff(formula = SurvObj.os ~ Pred.Class, data = data)
```

```
##
```

	N	Observed	Expected	(O-E) ² /E	(O-E) ² /V
## Pred.Class=1	140	125	175	14.5	28.6
## Pred.Class=2	258	248	198	12.9	28.6

```
##
```

```
## Chisq= 28.6 on 1 degrees of freedom, p= 9e-08
```

```
survdifff(SurvObj.os[Pred.Class==1] ~ FRET[Pred.Class==1], data=data)
```

```
## Call:
```

```
## survdifff(formula = SurvObj.os[Pred.Class == 1] ~ FRET[Pred.Class ==  
## 1], data = data)
```

```
##
```

	N	Observed	Expected	(O-E) ² /E	(O-E) ² /V
##					


```
## FRET[Pred.Class == 1]=low 109      98      89.1      0.88      3.15
## FRET[Pred.Class == 1]=high 31       27      35.9      2.19      3.15
##
## Chisq= 3.2  on 1 degrees of freedom, p= 0.08
```

```
survdifff(SurvObj.os[Pred.Class==2] ~ FRET[Pred.Class==2], data=data)
```

```
## Call:
## survdiff(formula = SurvObj.os[Pred.Class == 2] ~ FRET[Pred.Class ==
##      2], data = data)
##
##              N Observed Expected (O-E)^2/E (O-E)^2/V
## FRET[Pred.Class == 2]=low 158      151      141      0.647      1.52
## FRET[Pred.Class == 2]=high 100       97      107      0.858      1.52
##
## Chisq= 1.5  on 1 degrees of freedom, p= 0.2
```

```
ggrisktable(km, data=data) + theme_cleantable()
```

	Number at risk				
Pred.Class=1, FRET=low	109	70	30	13	4
Pred.Class=1, FRET=high	31	25	14	9	4
Pred.Class=2, FRET=low	158	66	14	7	2
Pred.Class=2, FRET=high	100	52	12	4	1

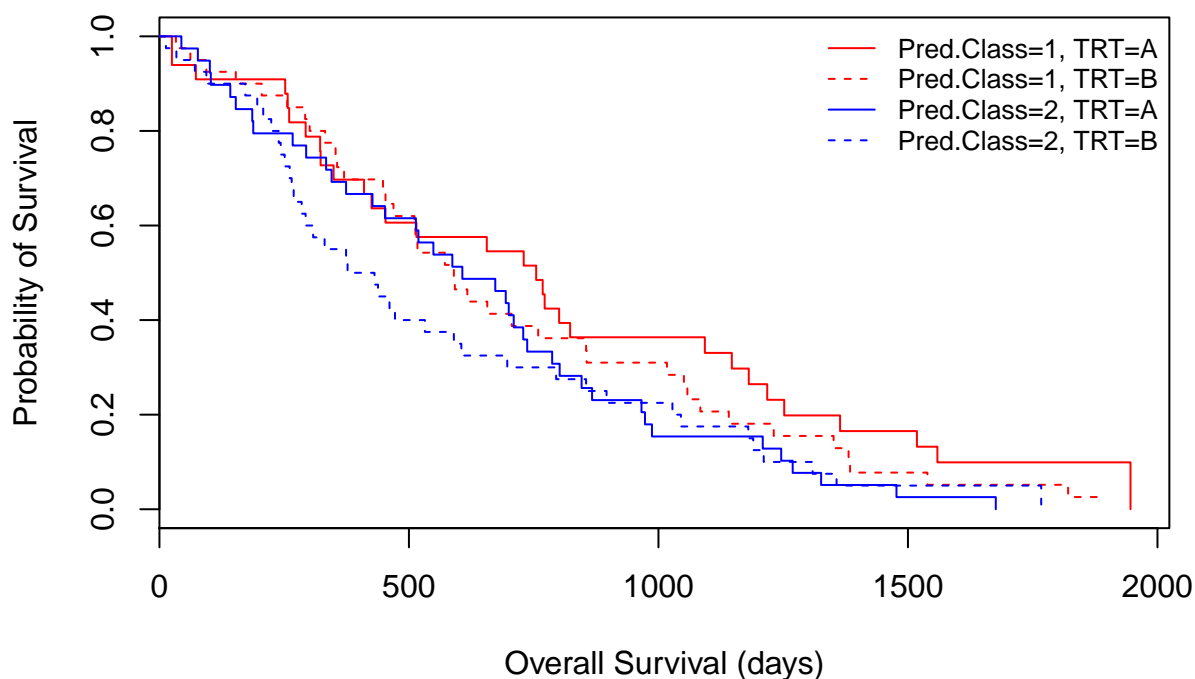
Analyse COIN III 152 validation set

```
data <- patient_data[patient_data$FRET.cohort==2,]
```

```
library(survival)
```

```
data$SurvObj.pfs <- with(data, Surv(pfstime, pfsevent))
data$SurvObj.os <- with(data, Surv(ostime, osevent))
col=c("red", "red", "blue", "blue")
lty=c(1,2,1,2)
```

```
km <- survfit(SurvObj.os ~ Pred.Class + TRT, data=data)
plot(km, col=col, xlab="Overall Survival (days)", ylab="Probability of Survival", lty = lty)
legend("topright", col=col, legend = names(km$strata), lty = lty, bty="n", cex=0.8)
```



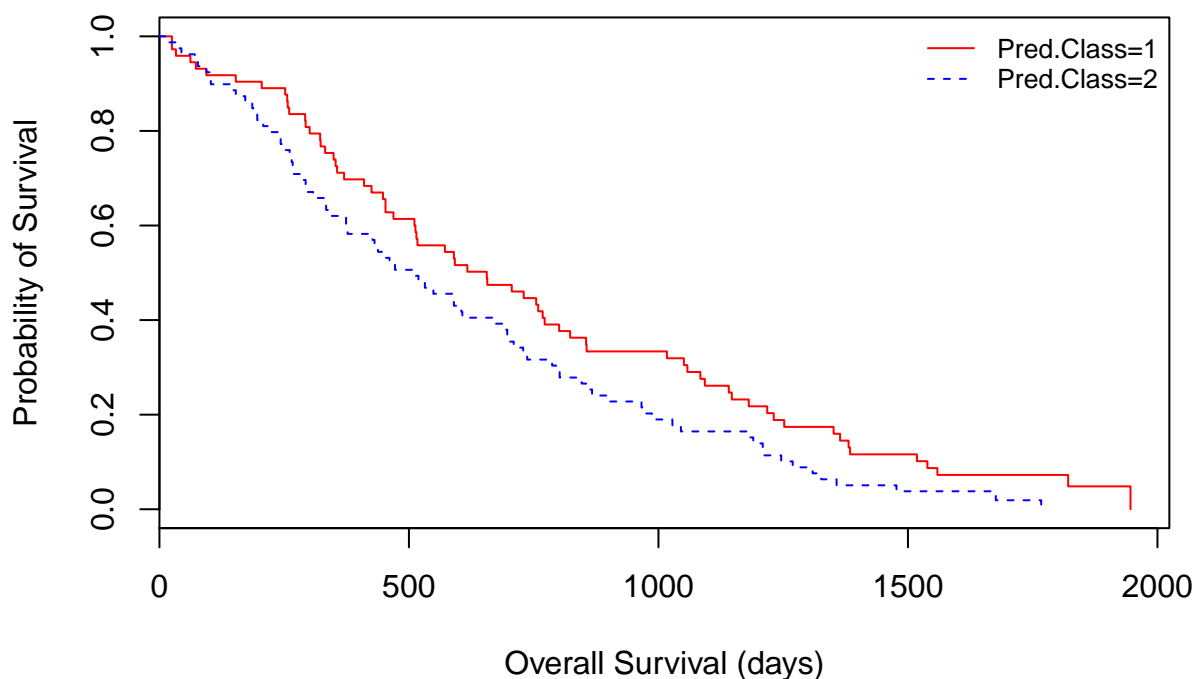
```
print(km)
```

```
## Call: survfit(formula = SurvObj.os ~ Pred.Class + TRT, data = data)
##
##              n events median 0.95LCL 0.95UCL
## Pred.Class=1, TRT=A 33      30    755    425    1181
## Pred.Class=1, TRT=B 40      38    590    469    856
## Pred.Class=2, TRT=A 39      39    607    452    787
## Pred.Class=2, TRT=B 40      39    404    285    697
```

```
ggrisktable(km, data=data) + theme_cleantable()
```

Number at risk					
Pred.Class=1, TRT=A	33	20	11	5	0
Pred.Class=1, TRT=B	40	24	12	3	0
Pred.Class=2, TRT=A	39	24	6	1	0
Pred.Class=2, TRT=B	40	16	9	2	0

```
col=c("red", "blue")
lty=c(1,2)
km <- survfit(SurvObj.os ~ Pred.Class, data=data)
plot(km, col=col, xlab="Overall Survival (days)", ylab="Probability of Survival", lty = lty)
legend("topright", col=col, legend = names(km$strata), lty = lty, bty="n", cex=0.8)
```



```
print(km)
```

```
## Call: survfit(formula = SurvObj.os ~ Pred.Class, data = data)
```

```
##
```

	n	events	median	0.95LCL	0.95UCL
Pred.Class=1	73	68	656	511	823
Pred.Class=2	79	78	514	374	697

```
survdif(SurvObj.os ~ Pred.Class, data=data)
```

```
## Call:
```

```
## survdif(formula = SurvObj.os ~ Pred.Class, data = data)
```

```
##
```

	N	Observed	Expected	(O-E)^2/E	(O-E)^2/V
Pred.Class=1	73	68	80.4	1.90	4.35
Pred.Class=2	79	78	65.6	2.33	4.35

```
##
```

```
## Chisq= 4.4 on 1 degrees of freedom, p= 0.04
```

```
survdif(SurvObj.os[Pred.Class==1] ~ TRT[Pred.Class==1], data=data)
```

```
## Call:
```

```
## survdif(formula = SurvObj.os[Pred.Class == 1] ~ TRT[Pred.Class ==  
## 1], data = data)
```

```
##
```

	N	Observed	Expected	(O-E)^2/E	(O-E)^2/V
TRT[Pred.Class == 1]=A	33	30	33.6	0.395	0.804
TRT[Pred.Class == 1]=B	40	38	34.4	0.387	0.804

```
##
## Chisq= 0.8 on 1 degrees of freedom, p= 0.4
survdifff(SurvObj.os[Pred.Class==2] ~ TRT[Pred.Class==2], data=data)

## Call:
## survdiff(formula = SurvObj.os[Pred.Class == 2] ~ TRT[Pred.Class ==
## 2], data = data)
##
##               N Observed Expected (O-E)^2/E (O-E)^2/V
## TRT[Pred.Class == 2]=A 39         39   40.2    0.0385    0.0814
## TRT[Pred.Class == 2]=B 40         39   37.8    0.0410    0.0814
##
## Chisq= 0.1 on 1 degrees of freedom, p= 0.8
ggrisktable(km, data=data) + theme_cleantable()
```

Number at risk					
Pred.Class=1	73	44	23	8	0
Pred.Class=2	79	40	15	3	0

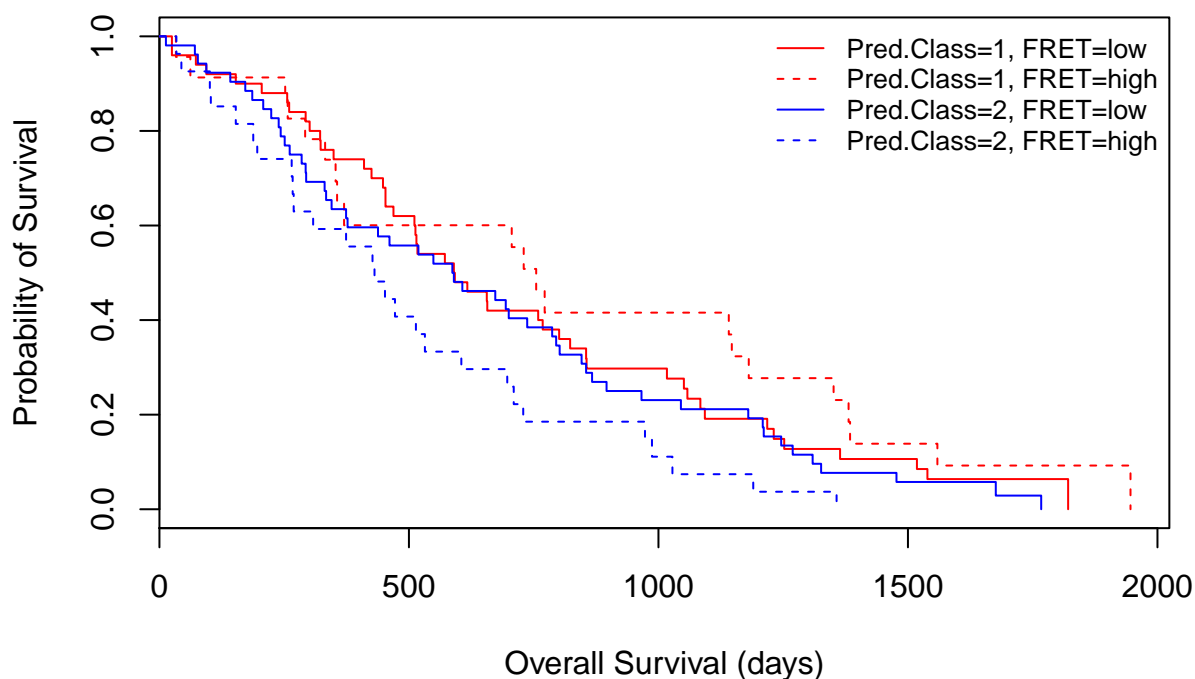
KM for raw FRET split by class

In Fig 4, FRET was split by tertiles of the original 398 patients. Use these same splits.

```
col=c("red", "red", "blue", "blue")
lty=c(1,2,1,2)

#Split FRET by Tertiles, or comment ot use same breaks as the 398
breaks <- quantile(data$FRET, probs = c(0.0, 0.67, 1.0), na.rm = T)
data$FRETeff <- cut(data$FRET, breaks = breaks, labels = c("low", "high"),
                    include.lowest = T)
data$FRET_raw <- data$FRET
data$FRET <- data$FRETeff

km <- survfit(SurvObj.os ~ Pred.Class + FRET, data=data)
plot(km, col=col, lty=lty, xlab="Overall Survival (days)", ylab="Probability of Survival")
legend("topright", col=col, legend = names(km$strata), lty=lty, bty="n", cex=0.8)
```



```
print(km)
```

```
## Call: survfit(formula = SurvObj.os ~ Pred.Class + FRET, data = data)
```

```
##
```

	n	events	median	0.95LCL	0.95UCL
## Pred.Class=1, FRET=low	50	47	591	469	855
## Pred.Class=1, FRET=high	23	21	755	356	1351
## Pred.Class=2, FRET=low	52	51	588	374	802
## Pred.Class=2, FRET=high	27	27	431	269	697

```
survdifff(SurvObj.os ~ Pred.Class, data=data)
```

```
## Call:
```

```
## survdifff(formula = SurvObj.os ~ Pred.Class, data = data)
```

```
##
```

	N	Observed	Expected	(O-E) ² /E	(O-E) ² /V
## Pred.Class=1	73	68	80.4	1.90	4.35
## Pred.Class=2	79	78	65.6	2.33	4.35

```
##
```

```
## Chisq= 4.4 on 1 degrees of freedom, p= 0.04
```

```
survdifff(SurvObj.os[Pred.Class==1] ~ FRET[Pred.Class==1], data=data)
```

```
## Call:
```

```
## survdifff(formula = SurvObj.os[Pred.Class == 1] ~ FRET[Pred.Class ==  
## 1], data = data)
```

```
##
```

	N	Observed	Expected	(O-E) ² /E	(O-E) ² /V
##					

```
## FRET[Pred.Class == 1]=low 50      47      43.1      0.362      1.04
## FRET[Pred.Class == 1]=high 23      21      24.9      0.624      1.04
##
##  Chisq= 1  on 1 degrees of freedom, p= 0.3
```

```
survdifff(SurvObj.os[Pred.Class==2] ~ FRET[Pred.Class==2], data=data)
```

```
## Call:
## survdiff(formula = SurvObj.os[Pred.Class == 2] ~ FRET[Pred.Class ==
##      2], data = data)
##
##              N Observed Expected (O-E)^2/E (O-E)^2/V
## FRET[Pred.Class == 2]=low 52      51      57.7      0.782      3.12
## FRET[Pred.Class == 2]=high 27      27      20.3      2.226      3.12
##
##  Chisq= 3.1  on 1 degrees of freedom, p= 0.08
```

```
ggrisktable(km, data=data) + theme_cleanable()
```

Number at risk					
Pred.Class=1, FRET=low	50	31	14	5	0
Pred.Class=1, FRET=high	23	13	9	3	0
Pred.Class=2, FRET=low	52	29	12	3	0
Pred.Class=2, FRET=high	27	11	3	0	0

Session Information

```
sessionInfo()
```

```
## R version 3.5.1 (2018-07-02)
## Platform: x86_64-w64-mingw32/x64 (64-bit)
## Running under: Windows 10 x64 (build 16299)
##
## Matrix products: default
##
## locale:
## [1] LC_COLLATE=English_United Kingdom.1252
## [2] LC_CTYPE=English_United Kingdom.1252
## [3] LC_MONETARY=English_United Kingdom.1252
## [4] LC_NUMERIC=C
## [5] LC_TIME=English_United Kingdom.1252
##
## attached base packages:
## [1] stats      graphics  grDevices  utils      datasets  methods    base
##
## other attached packages:
## [1] survival_2.42-3  pROC_1.15.3      kableExtra_1.1.0
## [4] knitr_1.20       survminer_0.4.6  ggpubr_0.2
## [7] magrittr_1.5     ggplot2_3.1.1    filesstrings_3.0.0
## [10] stringr_1.3.1
##
## loaded via a namespace (and not attached):
## [1] zoo_1.8-4        tidyselect_0.2.5  purrr_0.2.5
```

## [4] splines_3.5.1	lattice_0.20-35	colorspace_1.3-2
## [7] generics_0.0.2	vctrs_0.2.0	viridisLite_0.3.0
## [10] htmltools_0.3.6	yaml_2.2.0	survMisc_0.5.5
## [13] rlang_0.4.0	pillar_1.4.1	glue_1.3.0
## [16] withr_2.1.2	matrixStats_0.54.0	lifecycle_0.1.0
## [19] plyr_1.8.4	munsell_0.5.0	gtable_0.2.0
## [22] rvest_0.3.4	evaluate_0.12	labeling_0.3
## [25] broom_0.5.2	Rcpp_1.0.1	xtable_1.8-4
## [28] readr_1.3.1	checkmate_1.9.3	scales_1.0.0
## [31] backports_1.1.2	webshot_0.5.1	km.ci_0.5-2
## [34] gridExtra_2.3	hms_0.4.2	digest_0.6.18
## [37] stringi_1.2.4	dplyr_0.8.3	KMsurv_0.1-5
## [40] grid_3.5.1	rprojroot_1.3-2	tools_3.5.1
## [43] lazyeval_0.2.1	tibble_2.1.1	crayon_1.3.4
## [46] tidyr_1.0.0	pkgconfig_2.0.2	zeallot_0.1.0
## [49] Matrix_1.2-14	xml2_1.2.2	data.table_1.12.2
## [52] strex_1.0.1	httr_1.4.1	assertthat_0.2.0
## [55] rmarkdown_1.10	rstudioapi_0.10	R6_2.3.0
## [58] nlme_3.1-137	compiler_3.5.1	