QuantProject

```
## Loading required package: ggpubr

## Loading required package: magrittr

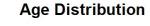
## Loading required package: lattice

## ## Attaching package: 'caret'

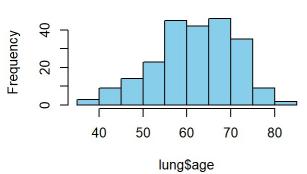
## The following object is masked from 'package:survival':
## ## cluster

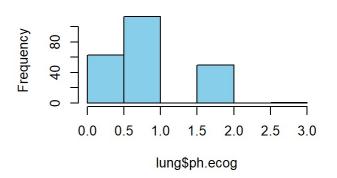
## [1] 2 2 1 2 2 1

par(mfrow = c(2,2))
hist(lung$age, main = "Age Distribution", col = "skyblue")
hist(lung$ph.ecog, main = "Distributio of ph.ecog Scores", col = "skyblue")
hist(lung$wt.loss, main = "Distributio of Weight Loss", col = "skyblue")
hist(lung$ph.karno, main = "Distributio of ph.karno Scores", col = "skyblue")
```



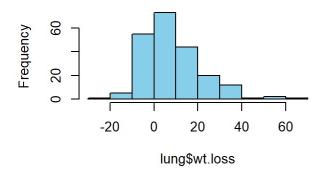
Distributio of ph.ecog Scores

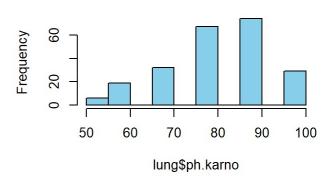




Distributio of Weight Loss

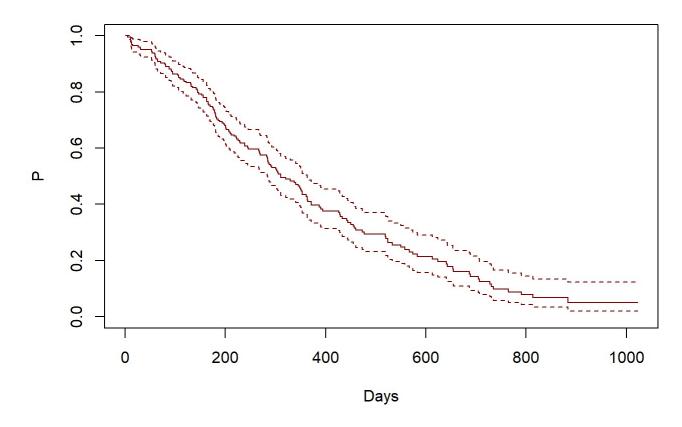
Distributio of ph.karno Scores





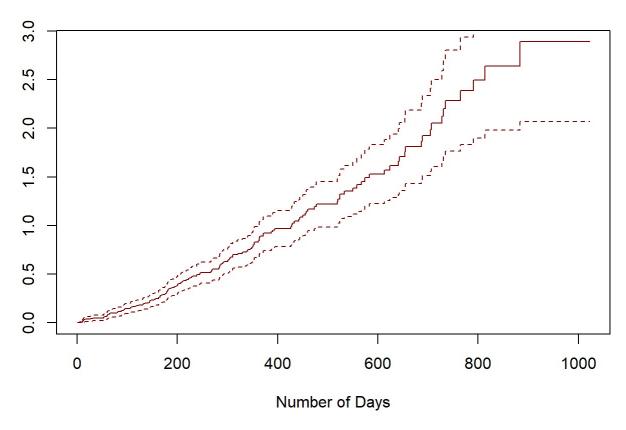
```
surv <- Surv(time = lung$time, event = lung$status)
all <- survfit(surv ~ 1, data = lung)
plot(all, main = "KM Survival Estimate", xlab = "Days", ylab = "P", col = "darkred")</pre>
```

KM Survival Estimate

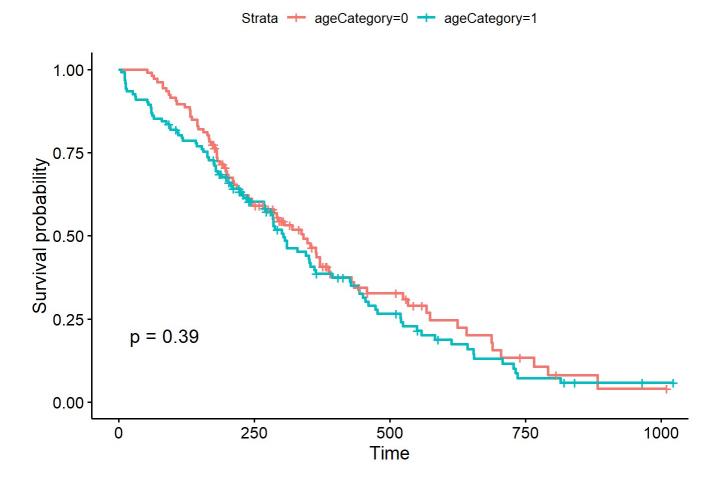


```
#cumulative hazard function
hazard <- plot(all, fun="cumhaz", col = "darkred", main = "Cumulative Hazard Function
", xlab = "Number of Days")</pre>
```

Cumulative Hazard Function

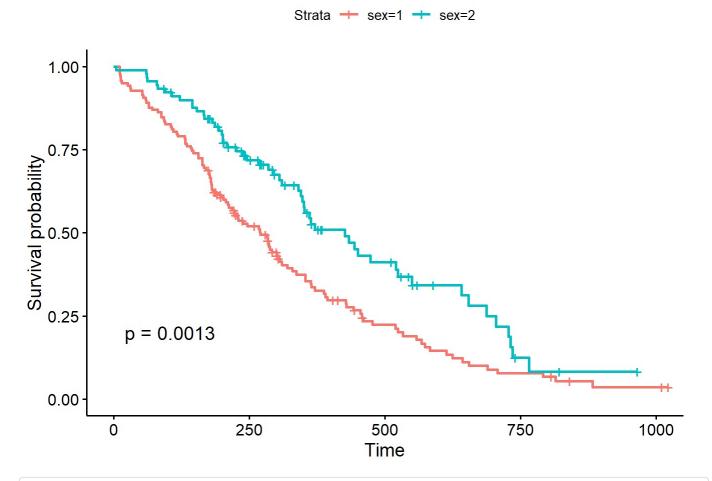


```
#surival function of age
#divided into two groups those over and under the mean age
ageCategory <- as.numeric(lung$age>mean(lung$age))
age <- survfit(surv ~ ageCategory, data = lung)
ageplot <-ggsurvplot(age,data = lung, pval = TRUE,linetype = "solid")
ageplot</pre>
```

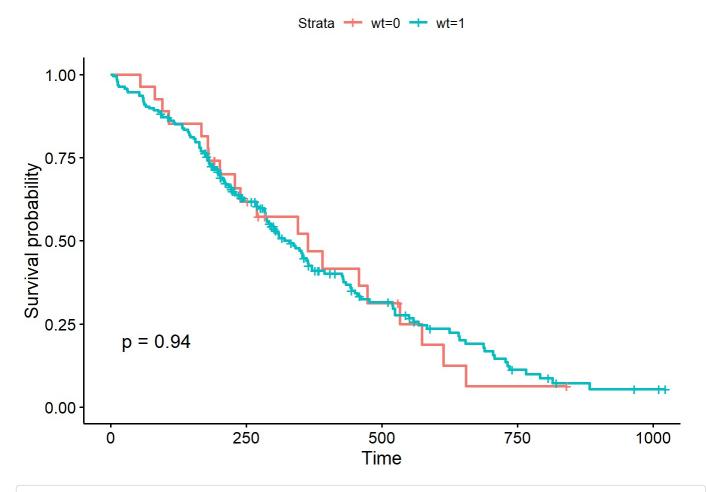


```
#survival plot for sex

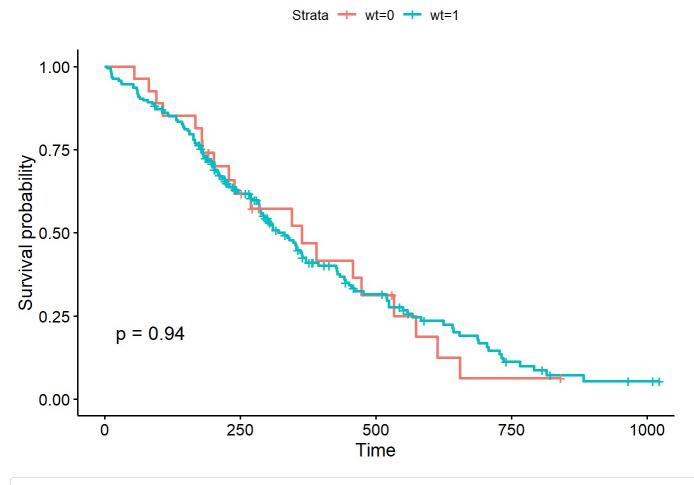
sexFit <- survfit(surv ~ sex, data = lung)
sexplot <- ggsurvplot(sexFit,data = lung, pval = TRUE,linetype = "solid")
sexplot</pre>
```



```
#weight loss, 1 = gained or no loss, 0 = lost weight
wt <- as.numeric(lung$wt.loss >= 0)
wtFit <- survfit(surv ~ wt, data = lung)
wtplot <- ggsurvplot(wtFit,data = lung, pval = TRUE,linetype = "solid")
wtplot</pre>
```



```
#weight loss, 1 = gained or no loss, 0 = lost weight
wt <- as.numeric(lung$wt.loss >= 0)
wtFit <- survfit(surv ~ wt, data = lung)
wtplot <- ggsurvplot(wtFit,data = lung, pval = TRUE,linetype = "solid")
wtplot</pre>
```



```
library(flexsurv)
exp <- flexsurvreg(surv~1, dist="exp")
gamma <- flexsurvreg(surv~1, dist="gamma")
genGamma <- flexsurvreg(surv~1, dist="gengamma")
log <- flexsurvreg(surv~1, dist="lognormal")

AICs <- c(exp$AIC, gamma$AIC, genGamma$AIC, log$AIC)
logs <- c(exp$loglik, gamma$loglik, genGamma$loglik, log$loglik)
dists <- c("Exponential", "Gamma", "Generalized Gammal", "log")
table <- rbind(dists, AICs, logs)

par(mfrow = c(2,2))
plot(exp,col = 5,main = "Exponential Fitted Curve")

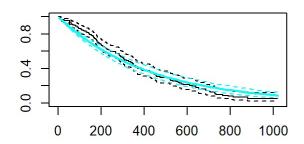
plot(gamma, col = 2, main = "Gamma Fitted Curve")

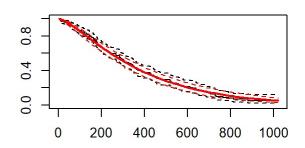
plot(genGamma, col = 3, main = "Generalized Gamma Fitted Curve")

plot(log, col = 4, main = "Log Fitted Curve")</pre>
```



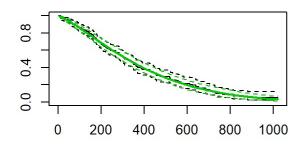
Gamma Fitted Curve

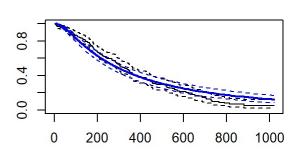




Generalized Gamma Fitted Curve

Log Fitted Curve





table

Survival proportions can be accurately modeled using the gamma distrabution

```
#creating cox survival object
lung <-na.omit(lung)
survCox <- Surv(time = lung$time, event = lung$status)</pre>
```

```
#cox univariant model with sex only
coxSex <- coxph(survCox ~ lung$sex, data = lung)
summary(coxSex)</pre>
```

```
## Call:
## coxph(formula = survCox ~ lung$sex, data = lung)
##
##
  n= 167, number of events= 120
##
##
            coef exp(coef) se(coef) z Pr(>|z|)
## Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
        exp(coef) exp(-coef) lower .95 upper .95
## lung$sex 0.6193 1.615 0.4212 0.9104
##
## Concordance= 0.567 (se = 0.025)
## Likelihood ratio test= 6.25 on 1 df, p=0.01
## Wald test = 5.94 on 1 df, p=0.01
## Score (logrank) test = 6.05 on 1 df, p=0.01
```

```
#cox univariant model with age only
coxAge <-coxph(survCox ~ lung$age, data = lung)
summary(coxAge)</pre>
```

```
## Call:
## coxph(formula = survCox ~ lung$age, data = lung)
##
  n= 167, number of events= 120
##
##
             coef exp(coef) se(coef) z Pr(>|z|)
## lung$age 0.01989 1.02009 0.01075 1.851 0.0642 .
## ---
## Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
         exp(coef) exp(-coef) lower .95 upper .95
##
## lung$age 1.02 0.9803 0.9988 1.042
##
## Concordance= 0.559 (se = 0.03)
## Likelihood ratio test= 3.52 on 1 df, p=0.06
## Wald test
                     = 3.43 on 1 df, p=0.06
## Score (logrank) test = 3.44 on 1 df, p=0.06
```

```
#cox univariant model with age only
coxecog <-coxph(survCox ~ lung$ph.ecog, data = lung)
summary(coxecog)</pre>
```

```
## Call:
## coxph(formula = survCox ~ lung$ph.ecog, data = lung)
##
##
  n= 167, number of events= 120
##
##
              coef exp(coef) se(coef) z Pr(>|z|)
## Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
            exp(coef) exp(-coef) lower .95 upper .95
## lung$ph.ecog 1.599 0.6255 1.232 2.075
##
## Concordance= 0.615 (se = 0.028)
## Likelihood ratio test= 12.41 on 1 df, p=4e-04
## Wald test = 12.44 on 1 df, p=4e-04
## Score (logrank) test = 12.61 on 1 df, p=4e-04
```

```
#multivariate cox model with stepwise selection of variables
Train <- createDataPartition(lung$status, p=0.75, list=FALSE)
training <- lung[ Train, ]
testing <- lung[ -Train, ]

res.cox <- coxph(survCox ~ lung$age + lung$sex + lung$ph.ecog +lung$ph.karno + lung$p
at.karno + lung$meal.cal + lung$wt.loss, data = training)
sp <- step(res.cox, data = training)</pre>
```

```
## Start: AIC=1002.07
## survCox ~ lung$age + lung$sex + lung$ph.ecog + lung$ph.karno +
      lung$pat.karno + lung$meal.cal + lung$wt.loss
##
##
                   Df
                         AIC
##
## - lung$meal.cal 1 1000.1
## - lung$age 1 1001.0
## <none>
                     1002.1
## - lung$pat.karno 1 1002.3
## - lung$wt.loss 1 1003.6
## - lung$ph.karno 1 1004.3
## - lung$sex 1 1008.0
## - lung$ph.ecog 1 1011.1
##
## Step: AIC=1000.08
## survCox ~ lung$age + lung$sex + lung$ph.ecog + lung$ph.karno +
##
      lung$pat.karno + lung$wt.loss
##
                   Df
                          AIC
## - lung$age
                   1 998.95
## <none>
                     1000.08
## - lung$pat.karno 1 1000.29
## - lung$wt.loss 1 1001.60
## - lung$ph.karno 1 1002.28
## - lung$sex 1 1006.29
## - lung$ph.ecog 1 1009.09
##
## Step: AIC=998.95
## survCox ~ lung$sex + lung$ph.ecog + lung$ph.karno + lung$pat.karno +
      lung$wt.loss
##
##
##
                   Df
                         AIC
## <none>
                       998.95
## - lung$pat.karno 1 999.34
## - lung$ph.karno 1 1000.53
## - lung$wt.loss 1 1000.74
## - lung$sex 1 1005.25
## - lung$ph.ecog
                   1 1007.83
```

```
#updated model based on results of the step selection
update.cox <- coxph(survCox ~ lung$sex + lung$ph.ecog + lung$ph.karno +
    lung$pat.karno + lung$wt.loss, data = lung)
summary(update.cox)</pre>
```

```
## Call:
## coxph(formula = survCox ~ lung$sex + lung$ph.ecog + lung$ph.karno +
##
      lung$pat.karno + lung$wt.loss, data = lung)
##
##
   n= 167, number of events= 120
##
                    coef exp(coef) se(coef) z Pr(>|z|)
##
           -0.558190 0.572244 0.199202 -2.802 0.00508 **
## lung$sex
## lung$ph.ecog 0.742983 2.102197 0.227604 3.264 0.00110 **
## lung$ph.karno 0.020366 1.020575 0.011080 1.838 0.06604 .
## lung$pat.karno -0.012401 0.987675 0.007978 -1.554 0.12008
## lung$wt.loss -0.014494 0.985611 0.007693 -1.884 0.05957 .
## ---
## Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
##
               exp(coef) exp(-coef) lower .95 upper .95
                 0.5722
## lung$sex
                           1.7475 0.3873
                                             0.8456
## lung$ph.ecog
                 2.1022
                           0.4757 1.3457
                                              3.2841
## lung$ph.karno 1.0206 0.9798 0.9987 1.0430
## lung$pat.karno 0.9877
                           1.0125 0.9724
                                             1.0032
## lung$wt.loss 0.9856 1.0146 0.9709 1.0006
##
## Concordance= 0.658 (se = 0.029)
## Likelihood ratio test= 27.28 on 5 df, p=5e-05
## Wald test = 26.89 on 5 df, p=6e-05
## Score (logrank) test = 27.64 on 5 df, p=4e-05
```

```
k<-1000
acc <- NULL

for (i in 1:k) {
    Train <- createDataPartition(lung$status, p=0.75, list=FALSE)
    training <- lung[ Train, ]
    testing <- lung[ -Train, ]

testSurv <- Surv(time = training$time, event = training$status)

update.coxT <- coxph(testSurv ~ sex + ph.ecog + ph.karno +
    pat.karno + wt.loss, data = training)

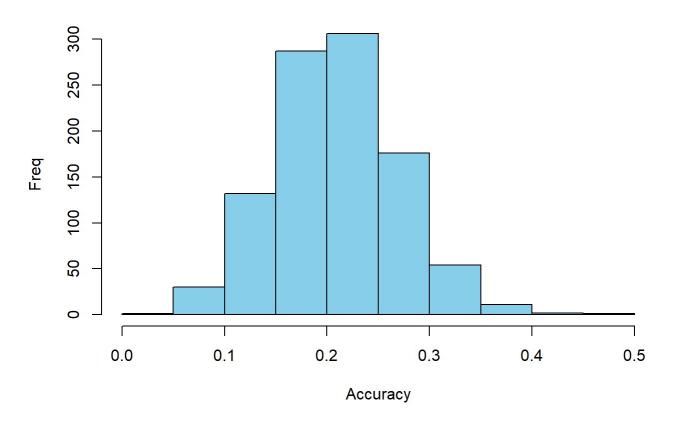
pred<-predict(update.coxT, newdata=testing,type = "risk")

results <- ifelse(pred > 0.5,1,0)
answers <- testing$status
misClasificError <- mean(answers != results)
acc[i]=1-misClasificError
}
mean(acc)</pre>
```

```
## [1] 0.2162927
```

```
hist(acc,xlab='Accuracy',ylab='Freq',
col='skyblue', main = "Accuracy of Cox Model 100 Iterations")
```

Accuracy of Cox Model 100 Iterations

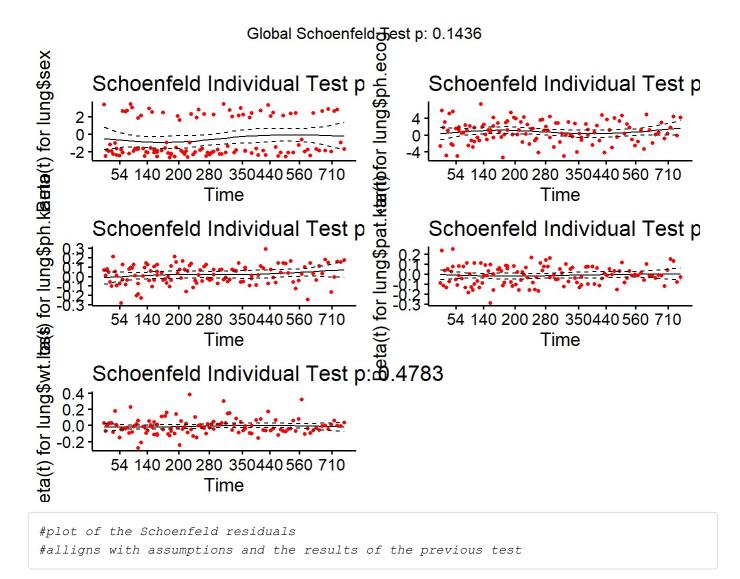


#testing cox model assumptions

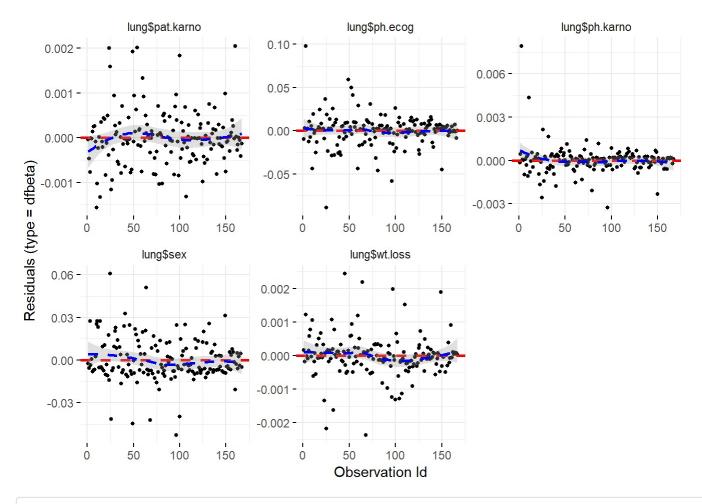
assump <- cox.zph(update.cox)
print("From these tests we find that we can consider all of the covarientes significa
nt ")</pre>

[1] "From these tests we find that we can consider all of the covarientes signific ant "

ggcoxzph(assump)



ggcoxdiagnostics(update.cox, type = "dfbeta", linear.predictions = FALSE, ggtheme = the
me minimal())



#testing for outliers in each of the covarients

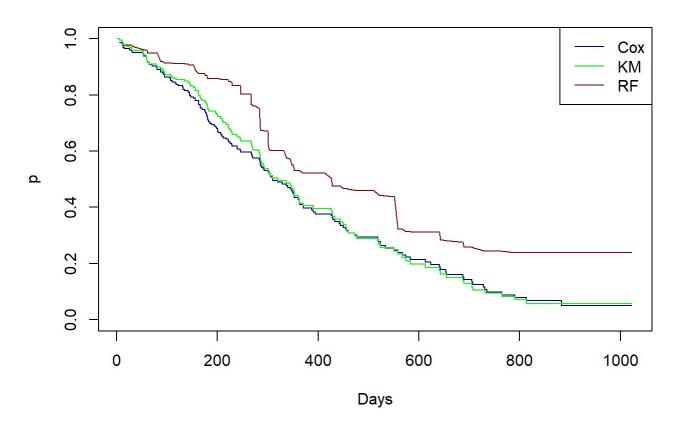
While age was significant when considered on its own after the stepwise process it is exclused from the covarients

```
#install.packages("ranger")
library(ranger)

#Fitting the random forest
ranger <- ranger(Surv(lung$time,lung$status==2) ~.,data=lung,num.trees = 500, importa
nce = "permutation",seed = 1)</pre>
```

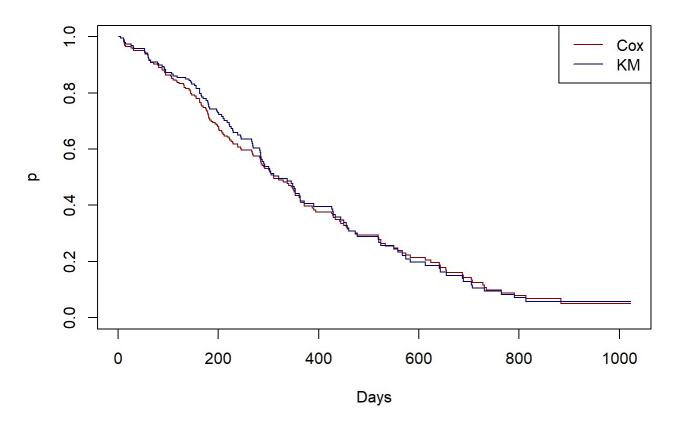
```
#ploting all models
coxfit <- survfit(update.cox)
rangePlot <- plot(ranger$unique.death.times,ranger$survival[1,], type = "l", ylim = c
(0,1),col = "darkred", main = "RM vs. Cox vs. KM", xlab = "Days", ylab = "p")
lines(all,conf.int = F, col = "navy")
lines(coxfit, col = "green", conf.int = F)
legend("topright",legend = c("Cox", "KM", "RF"), col = c("navy", "green", "darkred"),l
ty = 1)</pre>
```

RM vs. Cox vs. KM



```
plot(all, main = "Cox PHM vs. Kaplan-Meier", xlab = "Days", ylab = "p", col = "darkre
d", conf.int = F)
lines(survfit(update.cox),conf.int = F, col = "navy")
legend("topright",legend = c("Cox", "KM"), col = c("darkred", "navy"),lty = 1)
```

Cox PHM vs. Kaplan-Meier



```
data.frame(sort(ranger$variable.importance,decreasing = TRUE))
```

```
##
             sort.ranger.variable.importance..decreasing...TRUE.
## ph.ecog
                                                      0.0213567476
## pat.karno
                                                      0.0184760718
                                                      0.0057961043
## sex
## wt.loss
                                                      0.0026184009
## ph.karno
                                                      0.0022799733
## age
                                                      0.0001386971
## inst
                                                     -0.0006621027
## meal.cal
                                                     -0.0045072053
```

#shows the importance of each variable as determined by the random forest model