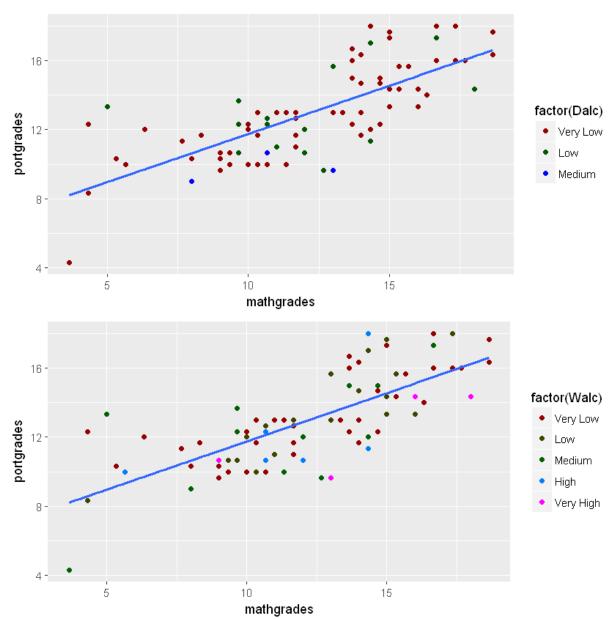
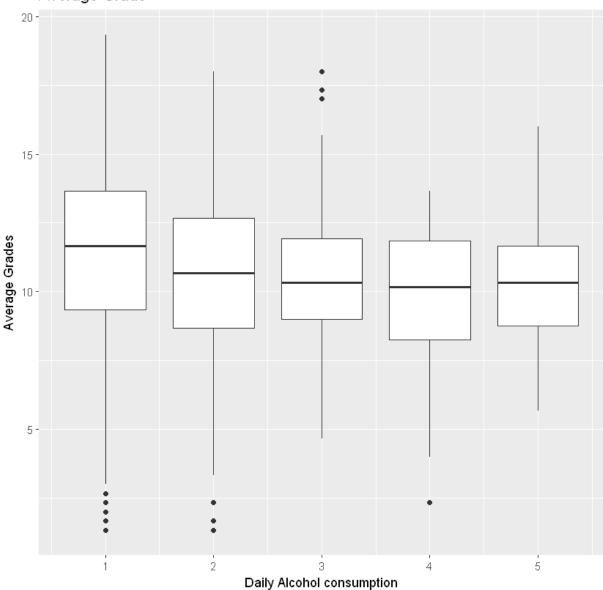
```
In [2]: library(ggplot2)
        library(plyr)
        library(dplyr)
        library(gridExtra)
        library(alluvial)
        library(extrafont)
        d1=read.table("C:/Program Files (x86)/R/R-2.2.0/student-mat.csv",sep=",",header=T
        d2=read.table("C:/Program Files (x86)/R/R-2.2.0/student-por.csv",sep=",",header=T
        #Following the suggestion of Carlo Ventrella, one of the attributes, "paid," is c
        #rather than student specific, so I am eliminating it from the list of attributes
        # are matched matched
        data.source=merge(d1,d2,by=c("school","sex","age","address","famsize","Pstatus",
                                      "Medu", "Fedu", "Mjob", "Fjob", "reason", "nursery", "inter
                                      "guardian", "guardian", "traveltime", "studytime", "failu
                                      "schoolsup", "famsup", "activities", "higher", "romantic"
                                     "famrel", "freetime", "goout", "Dalc", "Walc", "health", "a
        print(nrow(data.source))
```

In [3]: data.source\$mathgrades=rowMeans(cbind(data.source\$G1.x,data.source\$G2.x,data.sour data.source\$portgrades=rowMeans(cbind(data.source\$G1.y,data.source\$G2.y,data.sour data.source\$Dalc <- as.factor(data.source\$Dalc)</pre> data.source\$Dalc <- mapvalues(data.source\$Dalc,</pre> from = 1:5, to = c("Very Low", "Low", "Medium", "High", "Very H str1=ggplot(data.source, aes(x=mathgrades, y=portgrades)) + geom_point(aes(colour=factor(Dalc)))+ scale_colour_hue(l=25,c=150)+ geom_smooth(method = "lm", se = FALSE) data.source\$Walc <- as.factor(data.source\$Walc)</pre> data.source\$Walc <- mapvalues(data.source\$Walc,</pre> from = 1:5, to = c("Very Low", "Low", "Medium", "High", "Very H str2=ggplot(data.source, aes(x=mathgrades, y=portgrades))+ geom_point(aes(colour=factor(Walc)))+ scale_colour_hue(l=25,c=150)+ geom_smooth(method = "lm", se = FALSE) grid.arrange(str1,str2,nrow=2)

The following `from` values were not present in `x`: 4, 5



Average Grade



In [8]: failureind<-which(names(d3norepeats)=="failures")
 d3norepeats<-d3norepeats[,-failureind]</pre>

```
In [10]:
         # 1) multiple regression
         lm2<-lm(avggrades~., data=d3norepeats[,1:30])</pre>
         summary(1m2)
         Call:
         lm(formula = avggrades ~ ., data = d3norepeats[, 1:30])
         Residuals:
                              Median
              Min
                         10
                                           3Q
                                                    Max
         -10.8048
                    -1.5890
                              0.1455
                                       1.8676
                                                 8.4868
         Coefficients:
                             Estimate Std. Error t value Pr(>|t|)
         (Intercept)
                           10.6017499
                                       1.8139110
                                                    5.845 7.04e-09 ***
         schoolMS
                                                   -1.446
                           -0.3723012
                                       0.2574810
                                                           0.14854
         sexM
                           -0.1215773
                                       0.2255528
                                                   -0.539
                                                           0.59000
         age
                           -0.1198027
                                       0.0906465
                                                   -1.322
                                                           0.18661
         addressU
                            0.2930648
                                       0.2427090
                                                    1.207
                                                           0.22756
                                                    2.242
         famsizeLE3
                            0.4975437
                                       0.2219491
                                                           0.02522 *
         PstatusT
                            0.0640413
                                       0.3168516
                                                    0.202
                                                           0.83987
         Medu
                            0.1848194
                                       0.1396447
                                                    1.323
                                                           0.18600
         Fedu
                            0.1642152
                                       0.1234646
                                                    1.330
                                                           0.18383
         Mjobhealth
                            0.7195927
                                       0.4873500
                                                    1.477
                                                           0.14014
         Mjobother
                            0.0878501
                                       0.2874717
                                                    0.306
                                                           0.75998
         Mjobservices
                                                    1.060
                            0.3595006
                                       0.3392675
                                                           0.28959
                                                    0.239
         Mjobteacher
                            0.1089511
                                       0.4567083
                                                           0.81150
         Fjobhealth
                           -0.2298997
                                       0.6841685
                                                   -0.336
                                                           0.73693
         Fjobother
                                                    0.001
                            0.0004672
                                       0.4281211
                                                           0.99913
         Fjobservices
                           -0.2993809
                                       0.4493361
                                                   -0.666
                                                           0.50540
         Fjobteacher
                                                    1.925
                            1.1650673
                                       0.6051943
                                                           0.05452 .
         reasonhome
                                                    0.946
                            0.2418873
                                       0.2555691
                                                           0.34416
         reasonother
                            0.3335054
                                       0.3412013
                                                    0.977
                                                           0.32861
         reasonreputation 0.4954622
                                       0.2654763
                                                    1.866 0.06232 .
         guardianmother
                           -0.1771667
                                       0.2439794
                                                   -0.726
                                                           0.46793
         guardianother
                           -0.2497580
                                       0.4596099
                                                   -0.543
                                                           0.58698
         traveltime
                                                   -0.759
                           -0.1100901
                                       0.1449817
                                                           0.44785
         studytime
                                                    3.993 7.06e-05 ***
                            0.5080568
                                       0.1272501
         schoolsupyes
                           -1.5979812
                                       0.3190187
                                                   -5.009 6.56e-07 ***
         famsupyes
                           -0.3691163
                                       0.2091140
                                                   -1.765
                                                           0.07787 .
                           -0.6938560
                                       0.2413583
                                                   -2.875
                                                           0.00414 **
         paidyes
         activitiesyes
                            0.0980887
                                       0.2022346
                                                    0.485
                                                           0.62777
                                                    0.125
         nurseryyes
                            0.0309093
                                       0.2473158
                                                           0.90057
                                                    5.316 1.33e-07 ***
         higheryes
                            1.9236766
                                       0.3618564
         internetyes
                            0.4269133
                                       0.2569866
                                                    1.661
                                                           0.09701 .
         romanticyes
                           -0.4996783
                                       0.2097075
                                                   -2.383
                                                           0.01739 *
         famrel
                                                    1.486
                            0.1569738
                                       0.1056496
                                                           0.13768
         freetime
                           -0.0399315
                                       0.1029791
                                                   -0.388
                                                           0.69828
         goout
                           -0.2127131
                                       0.0995379
                                                   -2.137
                                                           0.03286 *
         Dalc
                                                   -0.938
                                                           0.34857
                           -0.1303141
                                       0.1389503
         Walc
                           -0.0210543
                                                   -0.193
                                                           0.84718
                                       0.1092166
         health
                           -0.1562929
                                       0.0709717
                                                   -2.202
                                                           0.02790 *
         absences
                           -0.0181346
                                       0.0162657
                                                   -1.115
                                                           0.26519
         Signif. codes:
                          0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
```

Residual standard error: 2.943 on 920 degrees of freedom

> Multiple R-squared: 0.2015, Adjusted R-squared: 0.1685 F-statistic: 6.11 on 38 and 920 DF, p-value: < 2.2e-16

In [12]: #2) Regression tree:

library(rpart)

library(DMwR)# I will be relying heavily on the DMwR library that comes with Torg rt2<-rpart(avggrades~., data=d3norepeats[,1:30])</pre>

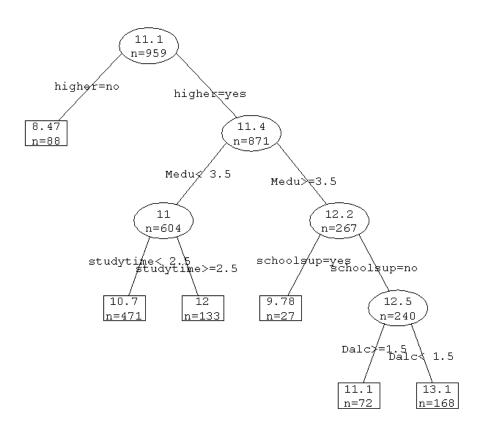
prettyTree(rt2)

Loading required package: lattice Loading required package: grid

Attaching package: 'DMwR'

The following object is masked from 'package:plyr':

join

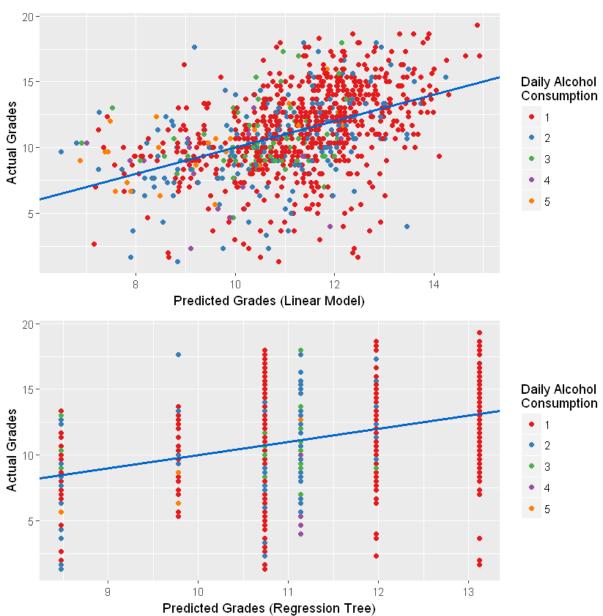


```
In [13]: #predictions lm.prediction
```

lm.predictions<-predict(lm2,d3norepeats)
rt.predictions<-predict(rt2,d3norepeats)
nmse.lm<-mean((lm.predictions-d3norepeats[,"avggrades"])^2)/mean((mean(d3norepeat
nmse.rt<-mean((rt.predictions-d3norepeats[,"avggrades"])^2)/mean((mean(d3norepeat
print(nmse.lm) #0.79
print(nmse.rt) #0.85</pre>

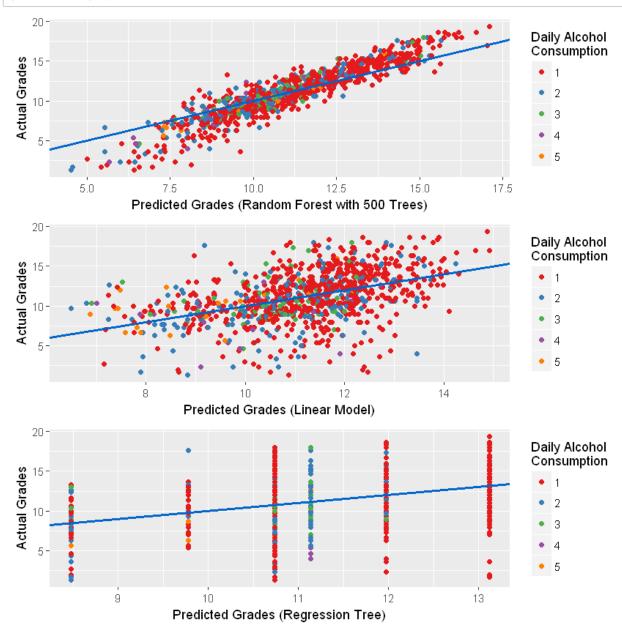
- [1] 0.7984877
- [1] 0.849412

In [14]: lmpltdata1=data.frame(cbind(lm.predictions,d3norepeats[,"avggrades"])) colnames(lmpltdata1)<-c("lm.predictions","avggrades")</pre> rtpltdata1=data.frame(cbind(rt.predictions,d3norepeats[,"avggrades"])) colnames(rtpltdata1)<-c("rt.predictions","avggrades")</pre> d3norepeats\$Dalc<-as.factor(d3norepeats\$Dalc)</pre> errplt.lt1=ggplot(lmpltdata1,aes(lm.predictions,avggrades))+ geom_point(aes(color=d3norepeats[,"Dalc"]))+ xlab("Predicted Grades (Linear Model)")+ ylab("Actual Grades")+ geom_abline(intercept=0,slope=1,color="#0066CC",size=1)+ #geom_smooth(method = "Lm", se = FALSE)+ scale_colour_brewer(palette = "Set1",name = "Daily Alcohol \nCo errplt.rt1=ggplot(rtpltdata1,aes(rt.predictions,avggrades))+ geom_point(aes(color=d3norepeats[,"Dalc"]))+ xlab("Predicted Grades (Regression Tree)")+ ylab("Actual Grades")+ geom abline(intercept=0,slope=1,color="#0066CC",size=1)+ #geom_smooth(method = "Lm", se = FALSE)+ scale_colour_brewer(palette = "Set1", name = "Daily Alcohol \nConsumption") grid.arrange(errplt.lt1,errplt.rt1,nrow=2)



In [16]: library(randomForest)
 set.seed(4543)
 rf2<-randomForest(avggrades~., data=d3norepeats[,1:30], ntree=500, importance=T)
 rf.predictions<-predict(rf2,d3norepeats)
 nmse.rf<-mean((rf.predictions-d3norepeats[,"avggrades"])^2)/mean((mean(d3norepeat print(nmse.rf)))</pre>

[1] 0.2038965



In []: