

Examining the Student Alcohol Consumption

Mine Your Business 

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| Data Source & Questions



UCI Machine Learning on Kaggle.com



How are students' personal background (family size, age, gender, etc.), if any, correlated to their alcohol consumption?



| Exploring the Data

➔ `dim(math)` 395 Records, 33 Variables

➔ Used structure function `str(math)`

➔ 16 int variables

16/33



Exploring the Data – Summary Statistics



Five Number Summary of Continuous Variables

| | age | Medu | Fedu | failures | absences |
|--------|------|------|------|----------|----------|
| Min | 15 | 0 | 0 | 0 | 0 |
| 1Q | 16 | 2 | 2 | 0 | 0 |
| Median | 17 | 3 | 2 | 0 | 4 |
| Mean | 16.7 | 2.75 | 2.52 | 0.33 | 5.71 |
| 3Q | 18 | 4 | 3 | 0 | 8 |
| Max | 22 | 4 | 4 | 3 | 75 |



Exploring the Data – Binary Responses



Counts of Binary Responses

| school | |
|--------|-----|
| GP | 349 |
| MS | 46 |

| sex | |
|-----|-----|
| F | 208 |
| M | 187 |

| address | |
|---------|-----|
| R | 88 |
| U | 307 |

| famsize | |
|---------|-----|
| GT3 | 281 |
| LE3 | 114 |

| Pstatus | |
|---------|-----|
| A | 41 |
| T | 354 |

| schoolsup | |
|-----------|-----|
| NO | 344 |
| YES | 242 |

| famsup | |
|--------|-----|
| NO | 153 |
| YES | 242 |

| paid | |
|------|-----|
| NO | 214 |
| YES | 181 |



Exploring the Data – Binary Responses 2



Counts of Binary Responses Continued

| activities | |
|------------|-----|
| NO | 194 |
| YES | 201 |

| nursery | |
|---------|-----|
| NO | 81 |
| YES | 314 |

| higher | |
|--------|-----|
| NO | 20 |
| YES | 375 |

| internet | |
|----------|-----|
| NO | 66 |
| YES | 329 |

| romantic | |
|----------|-----|
| NO | 263 |
| YES | 132 |

| Dalc | |
|------|-----|
| NO | 351 |
| YES | 44 |

| Walc | |
|------|-----|
| NO | 236 |
| YES | 159 |

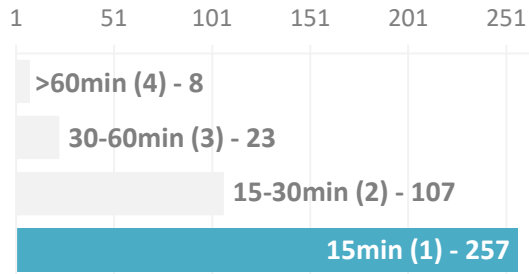


Exploring the Data – Likert-type Scale

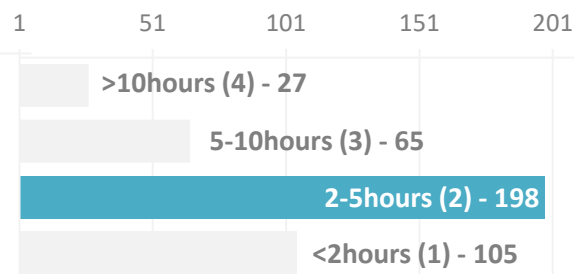


Responses in a Rating Scale

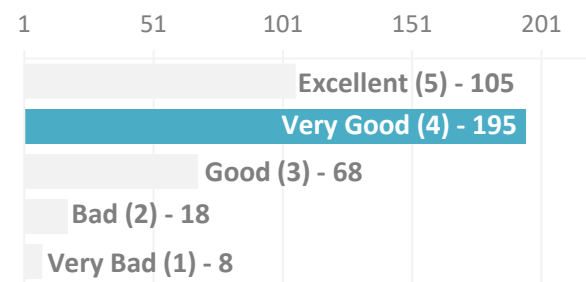
Travel Time



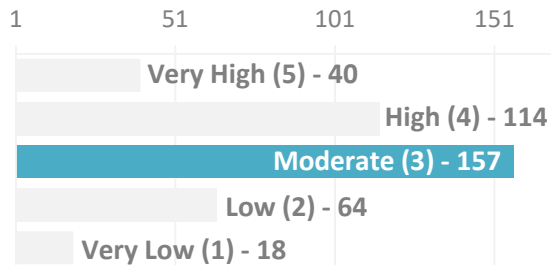
Study Time



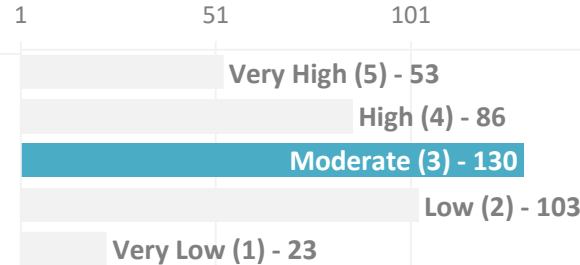
Quality of Family Relationships



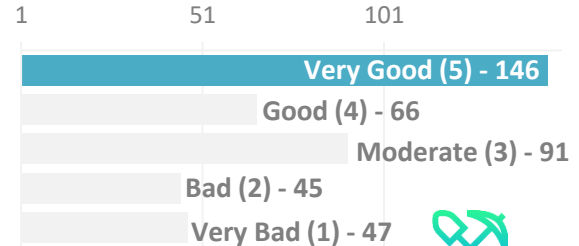
Free Time



Going out with Friends



Current Health Status

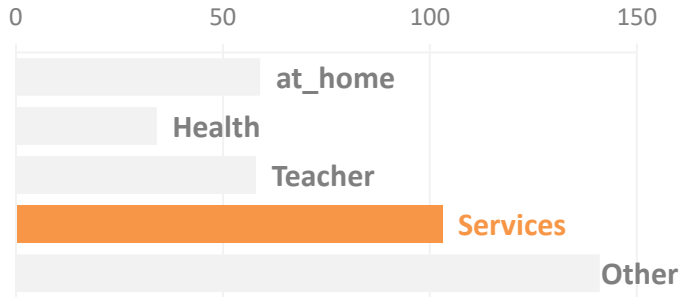


Exploring the Data – Nominal Response

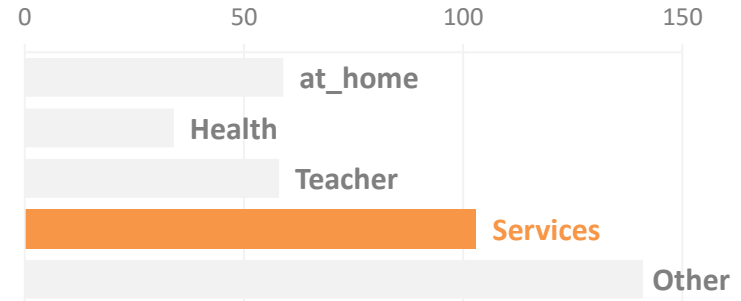


Nominal Response Variables

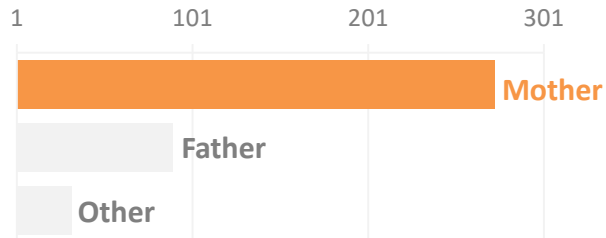
Father's Job



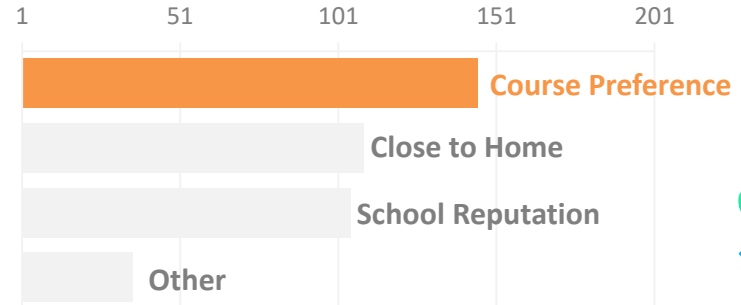
Mother's Job



Student's Guardian



Reason to Choose this School



| Converting Data – Categorical Variable



famrel, health, Mjob, Fjob, traveltime, studytime, freetime, goout, G1, G2, G3

```
math$famrel=as.factor(math$famrel)
math$health=as.factor(math$health)
math$Mjob=as.factor(math$Mjob)
math$Fjob=as.factor(math$Fjob)
math$traveltime=as.factor(math$traveltime)
math$studytime=as.factor(math$studytime)
math$freetime=as.factor(math$freetime)
math$goout=as.factor(math$goout)
math$G1=as.factor(math$G1)
math$G2=as.factor(math$G2)
math$G3=as.factor(math$G3)
```



| Converting Data – Binomial Variable

➔ Convert integer response to binomial “YES” and “NO”

```
math$Dalc[math$Dalc>2]="Yes"  
math$Dalc[math$Dalc<=2]="No"  
math$Walc[math$Walc>2]="Yes"  
math$Walc[math$Walc<=2]="No"  
math$Dalc=as.factor(math$Dalc)  
math$Walc=as.factor(math$Walc)
```



| Converting Data – Variables Contrasts

Two Variable Contrasts

```
contrasts(math$Dalc)
```

```
contrasts(math$Walc)
```

| Dalc | YES |
|------|-----|
| NO | 0 |
| YES | 1 |

| Walc | YES |
|------|-----|
| NO | 0 |
| YES | 1 |



| Choosing Predictors – Assumption

- ➔ sex: the student's gender
- ➔ Pstatus: the parent's cohabitation status
- ➔ romantic: the student's relationship status
- ➔ absences: the number of school absences
- ➔ failures: the number of past class failures
- ➔ famrel: the quality of a family relationship



Choosing Predictors – Stepwise Selection



Stepwise Regression

```
null = glm(math$Dalc ~ 1, family="binomial", data = math)
```

```
full = glm(math$Dalc ~  
math$school+math$sex+math$age+math$address+math$famsize+math$P  
status+math$Medu+math$Fedu+math$Mjob+math$Fjob+math$reason+m  
ath$guardian+math$traveltime+math$studytime+math$failures+math$sch  
oolsup+math$famsup+math$paid+math$activities+math$nursery+math$hi  
gher+math$internet+math$romantic+math$famrel+math$freetime+math$  
goout+math$health+math$absences+math$G1+math$G2+math$G3,  
family="binomial", data = math)
```

```
step.reg = step(null, scope=list(lower=null, upper=full), direction = 'both')
```

```
summary(step.reg)
```



| Choosing Predictors – Result

➔ Dalc AIC: 278.04 > **226.76**

➔ sex, goout (going out with friends), school, absences, traveltime, activities, higher (wants to take higher education), reason (reason to choose school), famsize, nuersery

➔ Walc AIC: 534.48 > **441.27**

➔ goout, Fjob (father's job), sex, absences, famrel, nursery, paid (extra paid classes within the course subject), traveltime, address (urban/rural), activities



| Choosing Predictors – Result Cont.

50% of our assumption

→ **sex, absences, famrel,** ~~Pstatus, romantic, failure~~

→ The Largest Odds Ratio – Male Students



| Model Selection – Bootstrapping



Limited data size



Sampling with replacement



4 resampled datasets from bootstrap

```
set.seed(14568)
```

```
train.dalc1=sample(nrow(math), 395 , replace=TRUE)
```

```
set.seed(23258)
```

```
train.dalc2=sample(nrow(math), 395 , replace=TRUE)
```

```
set.seed(36585)
```

```
train.dalc3=sample(nrow(math), 395 , replace=TRUE)
```

```
set.seed(45823)
```

```
train.dalc4=sample(nrow(math), 395 , replace=TRUE)
```



Mean accuracy



| Logistic Regression

➔ Using 4 resampled datasets from bootstrap

➔ Apply logistic model on each dataset



| Logistic Regression – Implementation



The R Code

```
glm.fit=glm(math$Dalc ~ math$sex + math$goout + math$school + math$absences +  
            math$travelttime + math$activities + math$higher + math$reason +  
            math$famsize + math$nursery,  
            data = math, subset=train.dalc1, family = "binomial")
```



The Prediction

```
glm.probs1=predict(glm.fit, math, type="response")  
glm.pred1=rep("No",395)  
glm.pred1[glm.probs1>.5]="Yes"
```



| Logistic Regression – Results

➔ `table(glm.pred1, test.truevalue)`

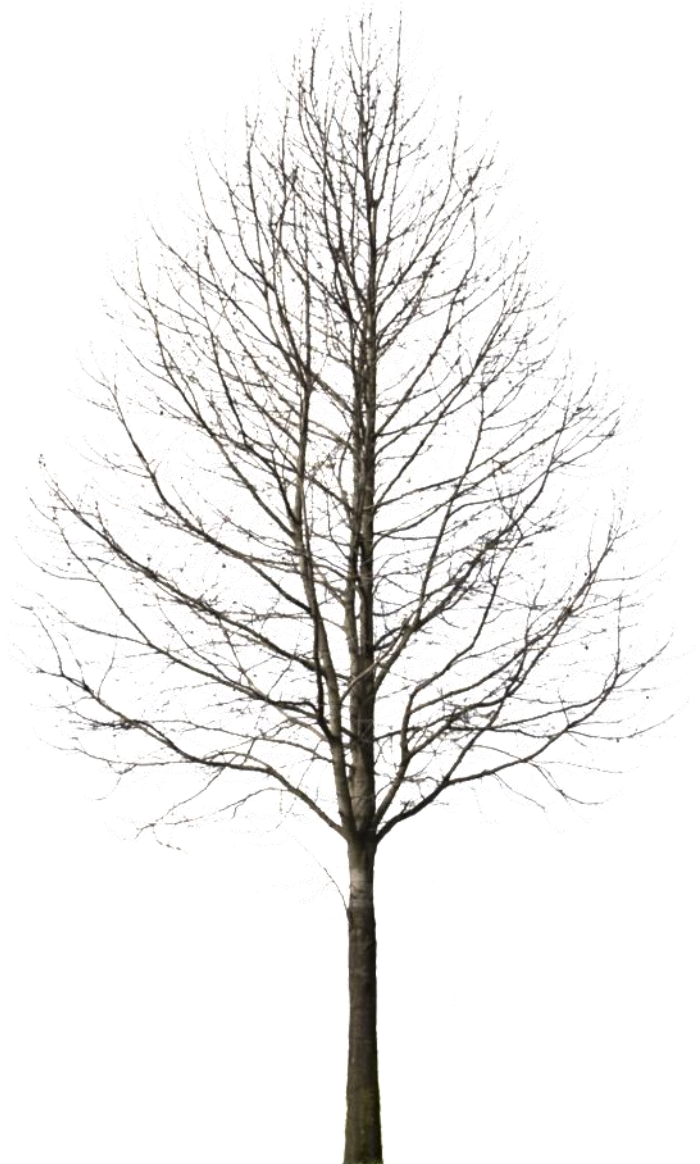
| | test.truevalue | |
|-----------|----------------|-----|
| | NO | YES |
| glm.pred1 | NO | 24 |
| | YES | 20 |



Classification Tree

➔ Definition

➔ Purpose



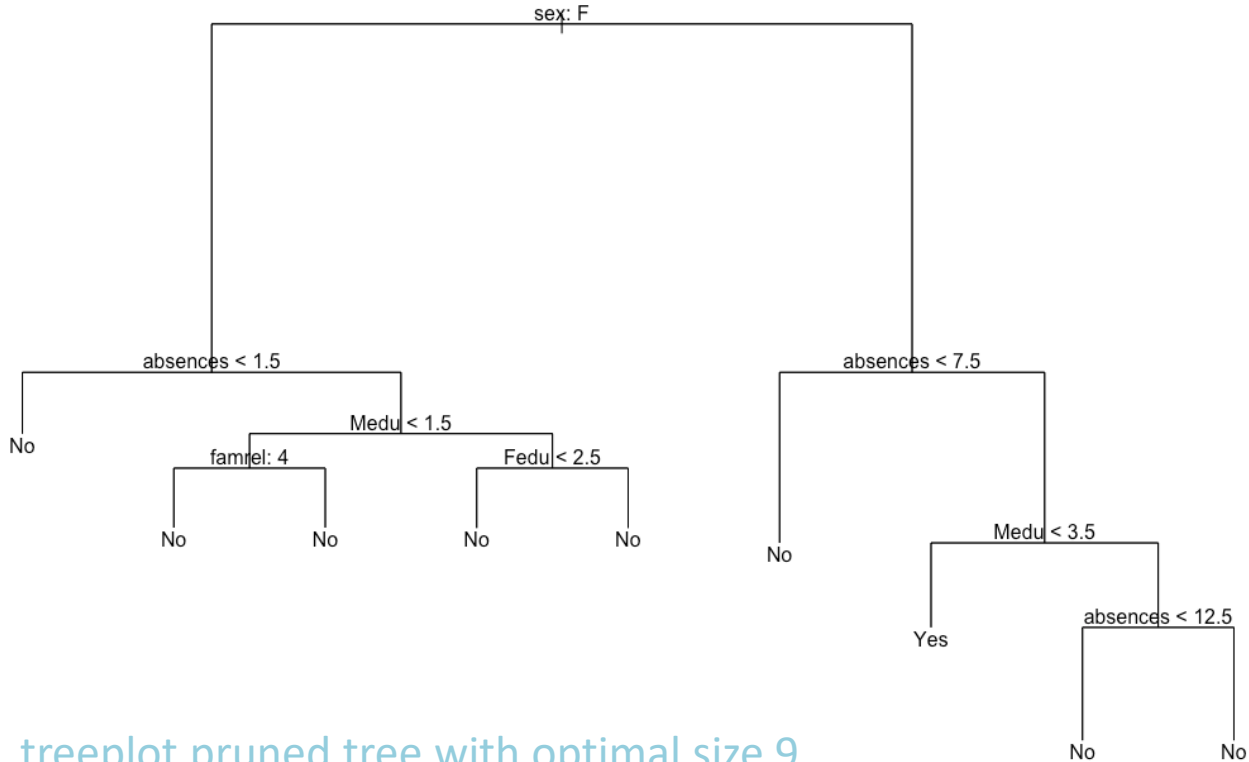
Classification Tree – Predictors

- famsize** LE3: ≤ 3 ; GT3: > 3
- nursery** YES; NO
- Pstatus** T: living together; A: living apart
- famrel** Numeric:
from 1 (Very Bad) – 5 (Excellent)
- Medu** 0: None; 1: Primary Edu. (4th grade)
2: 5th-9th grade; 3: Secondary Edu.
4: Higher Education
- Fedu** Same as Medu
- famsup** YES; NO
- absences** Count



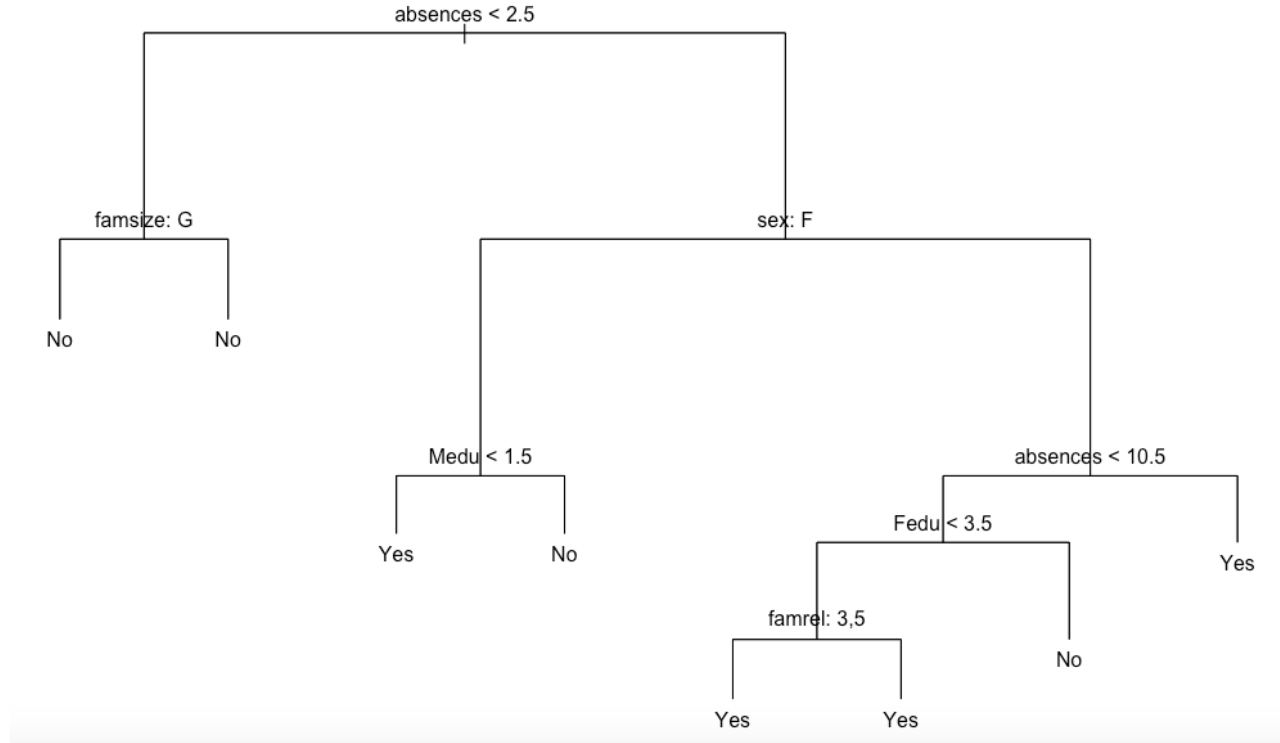
Classification Tree – Result (Weekday)

➔ Tree Diagram for the Weekday Alcohol Consumption



Classification Tree – Result (Weekend)

➔ Tree Diagram for the Weekend Alcohol Consumption



treeplot.pruned tree with optimal size 8



Classification Tree – Confusion Matrix



Confusion Matrix

```
mean(prunetree.pred==Dalc.test)
```

```
[1] 0.8810127
```

| | Dalc.test | |
|----------------|-----------|-----|
| | NO | YES |
| Prunetree.pred | NO | 392 |
| | YES | 29 |

```
mean(prunetree.pred==Walc.test)
```

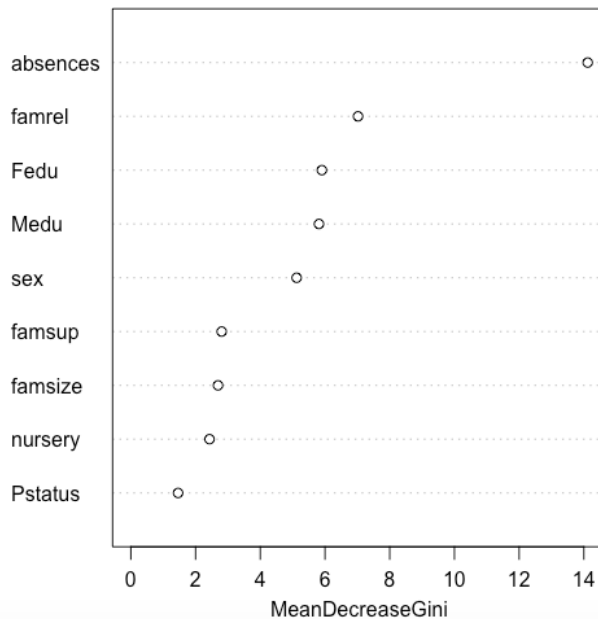
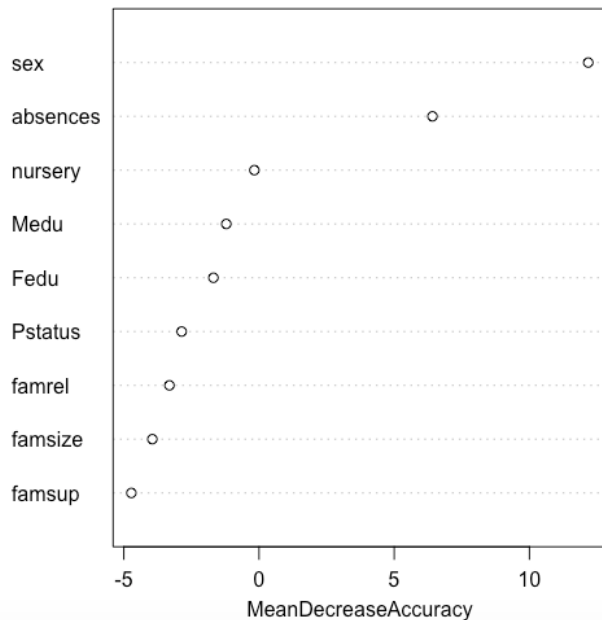
```
[1] 0.685544
```

| | Dalc.test | |
|----------------|-----------|-----|
| | NO | YES |
| Prunetree.pred | NO | 182 |
| | YES | 67 |



Classification Tree – Interpretation (Weekday)

➔ Interpretation (Weekday)



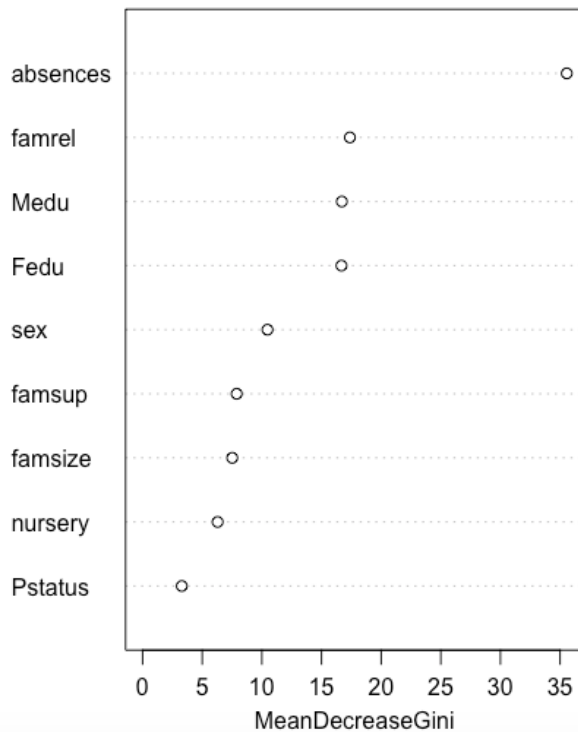
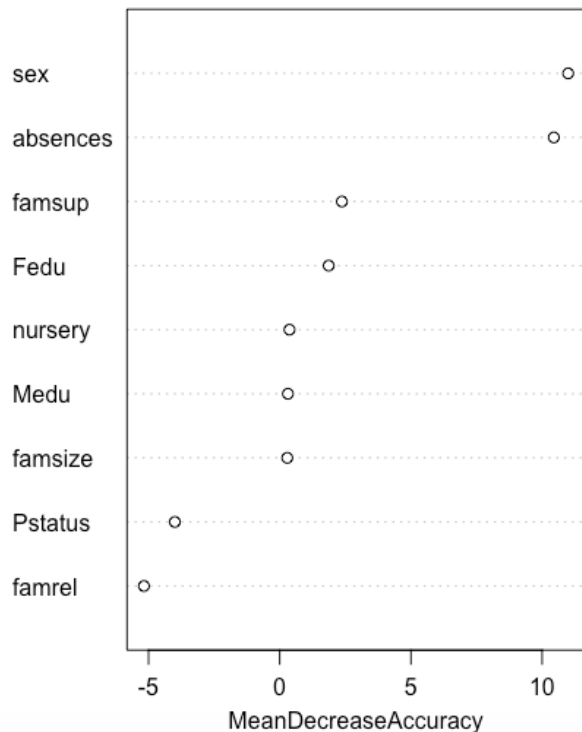
Important Variables

- absences
- Fedu
- Medu
- sex
- famrel
- nursery



Classification Tree – Interpretation (Weekend)

➔ Interpretation (Weekend)



Important Variables

- Absences
- Fedu
- Sex
- Famsup
- Medu
- famrel



Classification Tree – Evaluating the Tree Model

| | Tree | Bagging | Random Forest |
|-----------------------|-----------|-----------|---------------|
| Accuracy (Weekday) | 0.8810127 | 0.8101266 | 0.8101266 |
| Accuracy (Weekend) | 0.685544 | 0.5949367 | 0.6202532 |



Classification Tree – Key Findings



Evidence

“There is evidence suggesting that individuals who are children of alcoholics have a higher probability of becoming alcoholic or problem drinkers as a result of their unstable childhood family systems.”

– Professor Engs, Ruth C, Indiana University studying Family Background of Alcohol Abuse and Its Relationship to Alcohol Consumption among Students



| Conclusion

Logistic Regression > Decision Tree

