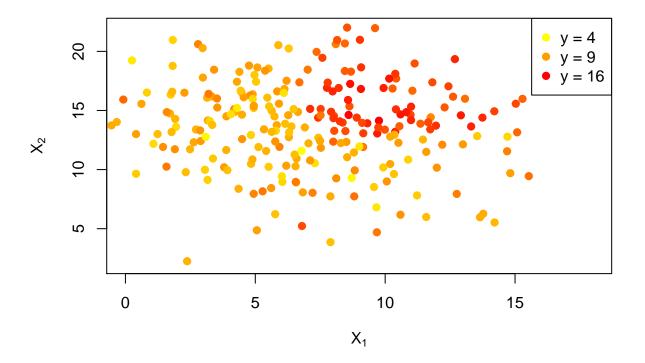
Code for recreating presentation examples

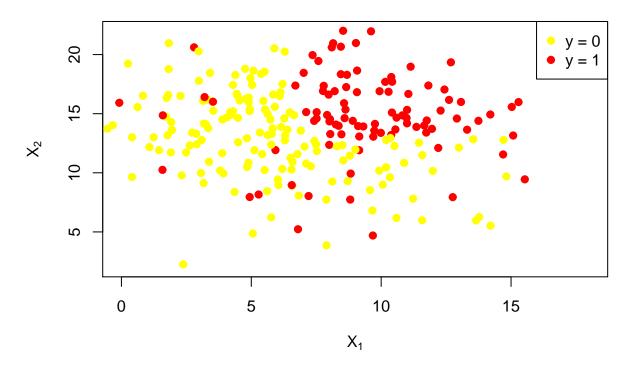
Examples involving X_1 and X_2

```
# generate data:
set.seed(12)
x1 <- rnorm(250, mean = 7, sd = 4)
x2 <- rnorm(250, mean = 14, sd = 4)
error <- rnorm(250, 0, 1.5)
y_cont <- 8 + 5*(x1 > 7 & x2 > 13) + error
y_bin <- as.factor(as.numeric(y_cont > mean(y_cont)))

# create plots:
plot(c(0, 18), c(2, 22), type = "n", xlab = expression(X[1]), ylab = expression(X[2]))
rbPal <- colorRampPalette(c('yellow', 'red'))
color <- rbPal(150)[as.numeric(cut(y_cont, breaks = 150))]
points(x1, x2, pch = 19, col = color)
cshort <- color[order(y_cont)][c(1, 125, 250)]
yshort <- round(y_cont[order(y_cont)][c(1, 125, 250)])
legend("topright", legend = paste("y =", yshort), col = cshort, pch = 19)</pre>
```

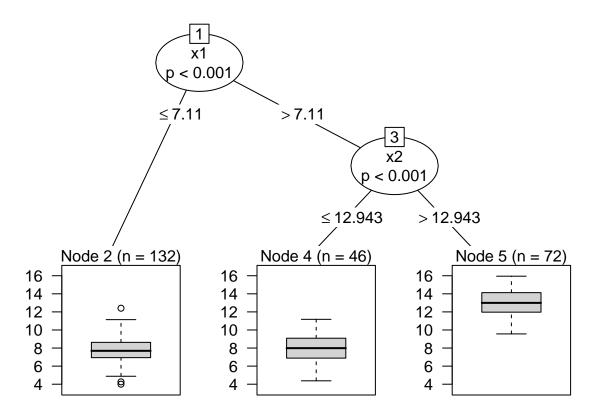


```
plot(c(0, 18), c(2, 22), type = "n", xlab = expression(X[1]), ylab = expression(X[2]))
color <- ifelse(y_bin == 0, "yellow", "red")
points(x1, x2, pch = 19, col = color)</pre>
```

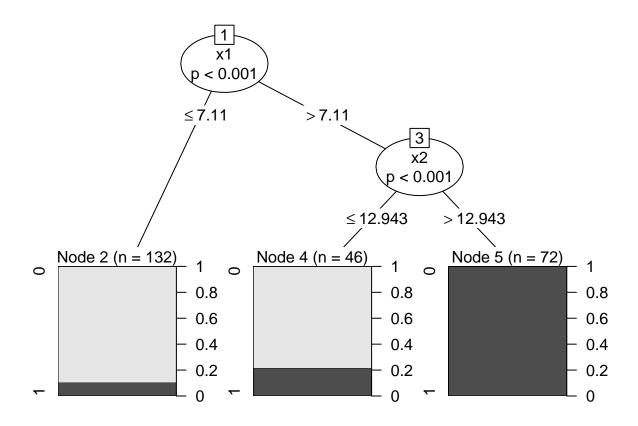


```
# fit and plot ctrees:
library("partykit")
```

```
## Loading required package: grid
ct_cont <- ctree(y_cont ~ x1 + x2)
plot(ct_cont)</pre>
```



ct_bin <- ctree(y_bin ~ x1 + x2)
plot(ct_bin)</pre>



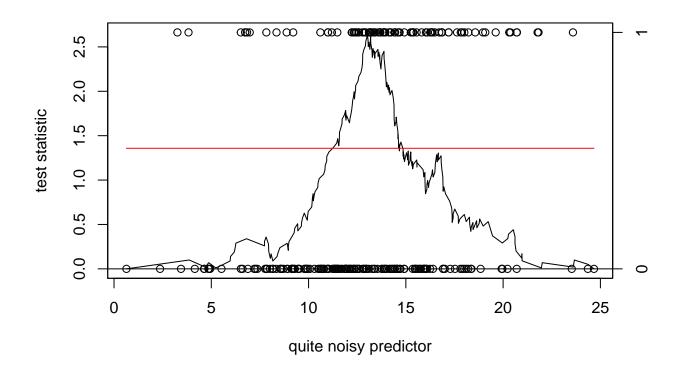
```
# perform test for variable and split point selection:
library("coin")
## Loading required package: survival
library("strucchange")
## Loading required package: zoo
## Attaching package: 'zoo'
## The following objects are masked from 'package:base':
##
       as.Date, as.Date.numeric
##
## Loading required package: sandwich
maxstat_test(y_bin ~ x1 + x2)
##
    Asymptotic Generalized Maximally Selected Statistics
##
##
## data: y_bin by x1, x2
## maxT = 9.5376, p-value < 2.2e-16
## alternative hypothesis: two.sided
## sample estimates:
##
     "best" cutpoint: <= 7.109992
          covariable: x1
##
```

```
?maxstat_test
```

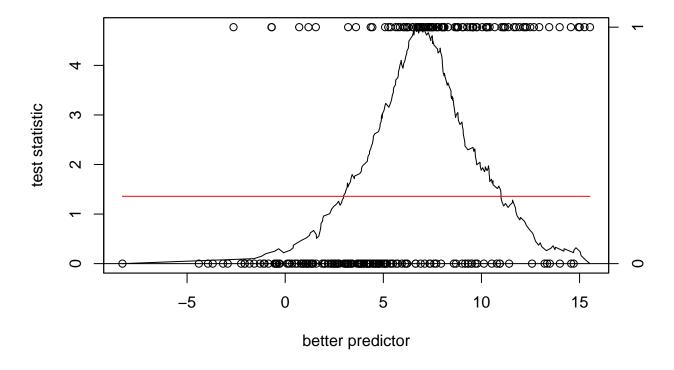
```
## starting httpd help server \dots
```

done

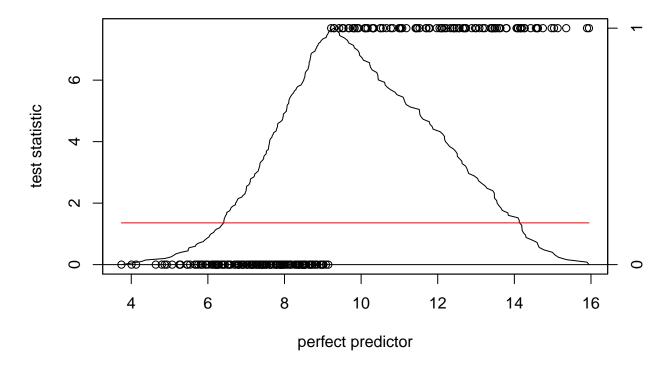
```
# Plot test statistics for X2 (noisy predictor):
gefp.x2.b <- gefp(y_bin ~ 1, family = binomial, order.by = ~ x2)
plot(gefp.x2.b, main = " ", ylab = "test statistic", xlab = "quite noisy predictor")
par(new = TRUE)
plot(x2, as.numeric(y_bin)-1, axes = FALSE, xlab = NA, ylab = NA)
mtext(side = 4, line = 3, 'class', cex = .7)
axis(side = 4, at = c(0,1))</pre>
```



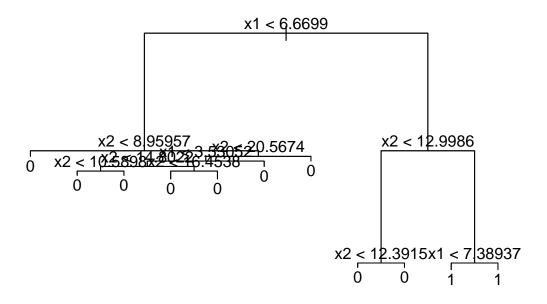
```
# Plot test statistics for X1 (better predictor):
gefp.x1.b <- gefp(y_bin ~ 1, family = binomial, order.by = ~ x1)
plot(gefp.x1.b, main = " ", ylab = "test statistic", xlab = "better predictor")
par(new = TRUE)
plot(x1, as.numeric(y_bin)-1, axes=F, xlab=NA, ylab=NA)
mtext(side = 4, line = 3, 'class', cex = .7)
axis(side = 4, at = c(0,1))</pre>
```



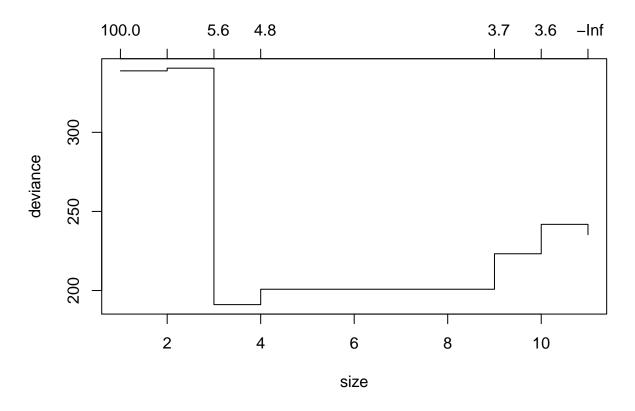
```
# Plot test statistics for the underlying continuous y variables (perfect predictor):
gefp.y_cont.b <- gefp(y_bin ~ 1, family = binomial, order.by = ~ y_cont)
plot(gefp.y_cont.b, main = " ", ylab = "test statistic", xlab = "perfect predictor")
par(new = TRUE)
plot(y_cont, as.numeric(y_bin)-1, axes=F, xlab=NA, ylab=NA)
mtext(side = 4, line = 3, 'class', cex = .7)
axis(side = 4, at = c(0,1))</pre>
```



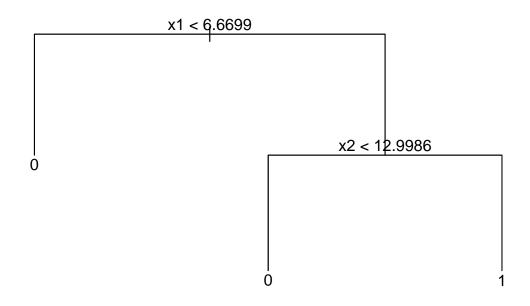
```
# fit and plot CART trees:
library("tree")
cart <- tree(y_bin ~ x1 + x2)
plot(cart)
text(cart)</pre>
```



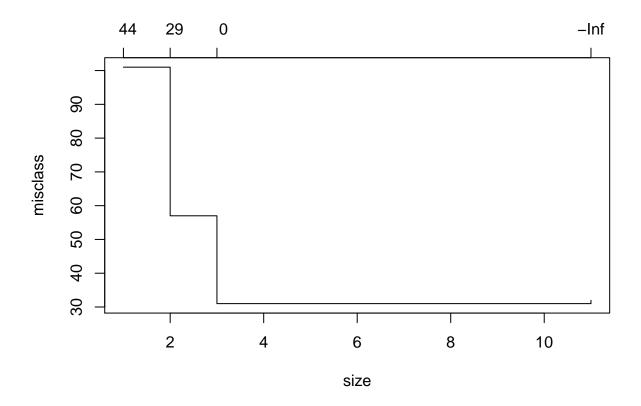
```
# Prune CART tree:
set.seed(12)
cv_cart <- cv.tree(cart)
plot(cv_cart)</pre>
```



```
pruned_cart <- prune.tree(cart, best = 3)
plot(pruned_cart)
text(pruned_cart)</pre>
```



```
# Alternatively, we could use misclassification error (instead of deviance) for pruning:
set.seed(12)
cv_cart2 <- cv.tree(cart, FUN = prune.misclass)
plot(cv_cart2)</pre>
```



In this case, deviance and misclassification error suggest trees of same size: 3 nodes

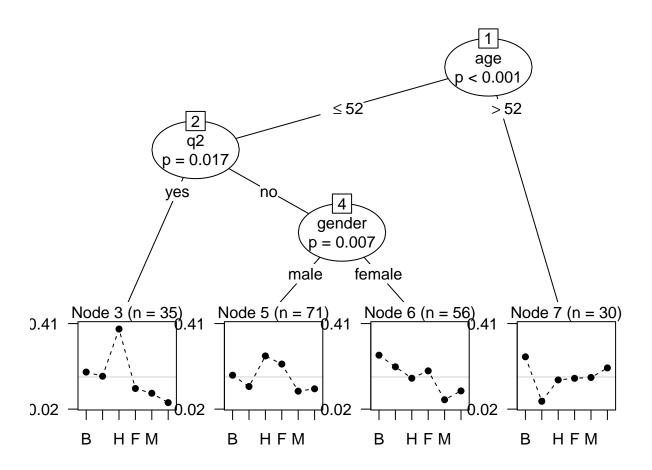
Bradley-Terry tree example

```
library("psychotree")

## Loading required package: psychotools

data("Topmodel2007", package = "psychotree")

tm_tree <- bttree(preference ~ ., data = Topmodel2007, minsplit = 5)
plot(tm_tree, abbreviate = 1)</pre>
```



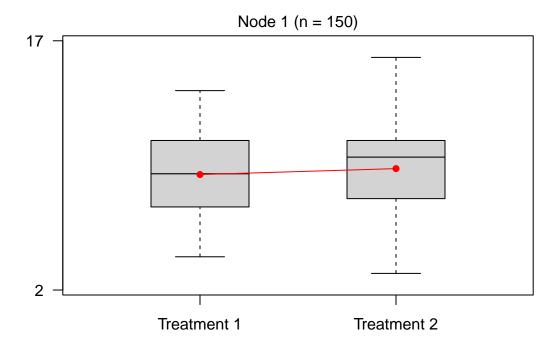
Linear (mixed-effects) model trees examples

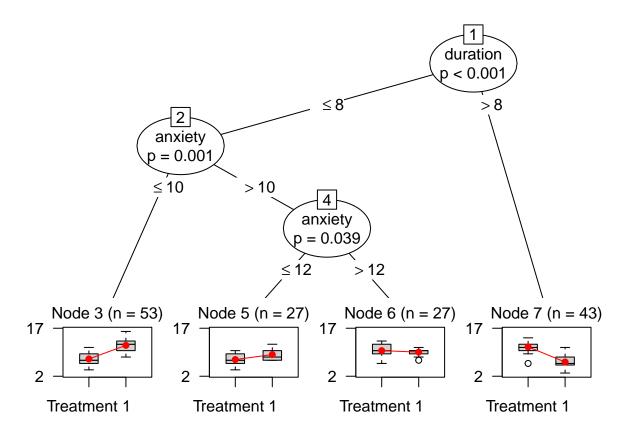
```
library("glmertree")

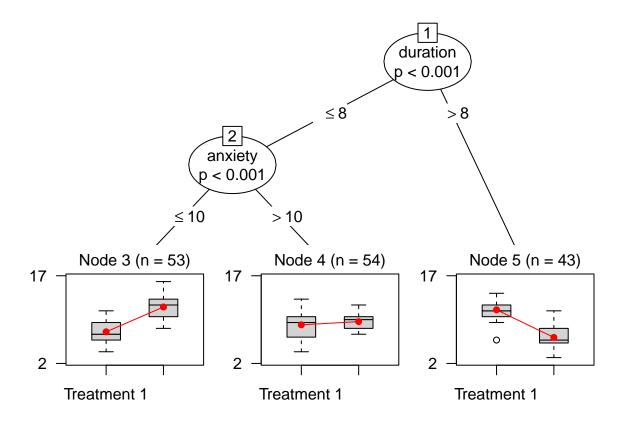
## Loading required package: lme4

## Loading required package: Matrix

# fit and plot global lm:
global_lm <- lmtree(depression ~ treatment | age, data = DepressionDemo)
plot(global_lm)</pre>
```

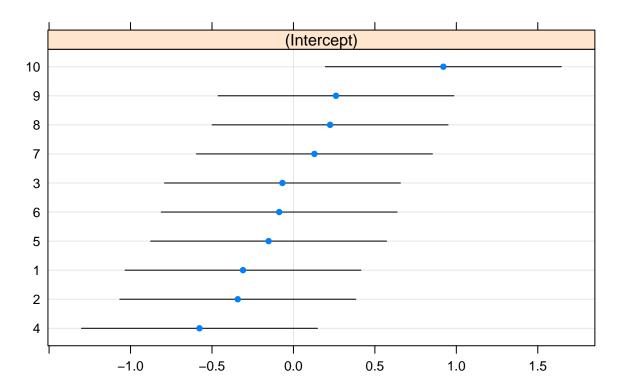




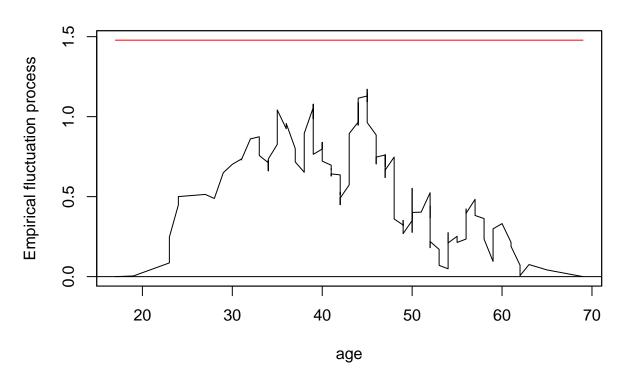


\$cluster

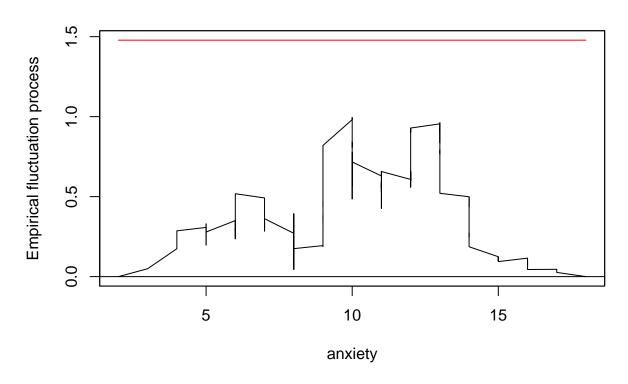
cluster



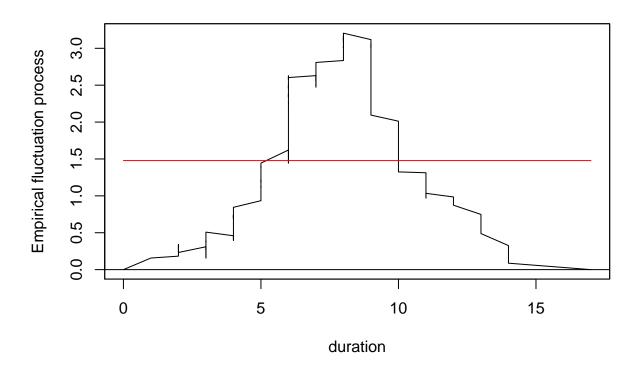
M-fluctuation test



M-fluctuation test



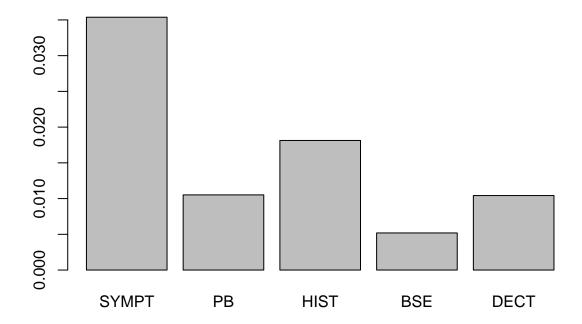
M-fluctuation test



Variable importance example

```
# Note that cforest is implemented in partykit as well as party package.
# However, only party package has varimp() function, which is needed for
# calculating importances:
library("party")
## Loading required package: mvtnorm
## Loading required package: modeltools
## Loading required package: stats4
##
## Attaching package: 'modeltools'
## The following object is masked from 'package:lme4':
##
##
       refit
##
## Attaching package: 'party'
## The following objects are masked from 'package:partykit':
##
##
       cforest, ctree, ctree_control, edge_simple, mob, mob_control,
       node_barplot, node_bivplot, node_boxplot, node_inner,
##
##
       node_surv, node_terminal
```

```
data("mammoexp", package = "TH.data")
set.seed(42) # note: different seed was used for presentation
cf_me <- cforest(ME ~ SYMPT + PB + HIST + BSE + DECT, data = mammoexp)
barplot(varimp(cf_me))</pre>
```



Conditional permutation importance example

```
set.seed(12) # note: different seed was used in presentation
mycf <- cforest(score ~ ., data = readingSkills, control = cforest_unbiased(mtry = 2))</pre>
varimp(mycf)
## nativeSpeaker
                                     shoeSize
        12.98369
##
                      78.57741
                                     17.04826
varimp(mycf, conditional = TRUE)
## nativeSpeaker
                            age
                                     shoeSize
##
        11.32044
                      43.95367
                                      1.52258
```