

# Resampling

## Manual Splitting – what could possibly go wrong?

```
library(mlr)

## Loading required package: ParamHelpers
task = makeClassifTask(data = iris, target = "Species")
learner = makeLearner("classif.randomForest")
```

### The Good

```
model = train(learner, task, subset = c(1:30, 51))
model

## Model for learner.id=classif.randomForest; learner.class=classif.randomForest
## Trained on: task.id = iris; obs = 31; features = 4
## Hyperparameters:

predictions = predict(model, task = task, subset = 31:50)
predictions

## Prediction: 20 observations
## predict.type: response
## threshold:
## time: 0.00
##   id truth response
## 31 31 setosa  setosa
## 32 32 setosa  setosa
## 33 33 setosa  setosa
## 34 34 setosa  setosa
## 35 35 setosa  setosa
## 36 36 setosa  setosa
## ... (20 rows, 3 cols)

performance(predictions, measures = acc)

## acc
## 1

calculateConfusionMatrix(predictions)

##           predicted
## true      setosa versicolor virginica -err.-
## setosa      20         0         0      0
## versicolor  0         0         0      0
## virginica   0         0         0      0
## -err.-      0         0         0      0
```

## The Bad

```
model = train(learner, task, subset = 1:100)
model

## Model for learner.id=classif.randomForest; learner.class=classif.randomForest
## Trained on: task.id = iris; obs = 100; features = 4
## Hyperparameters:

predictions = predict(model, task = task, subset = 101:150)
predictions

## Prediction: 50 observations
## predict.type: response
## threshold:
## time: 0.29
##   id      truth  response
## 101 101 virginica versicolor
## 102 102 virginica versicolor
## 103 103 virginica versicolor
## 104 104 virginica versicolor
## 105 105 virginica versicolor
## 106 106 virginica versicolor
## ... (50 rows, 3 cols)

performance(predictions, measures = acc)

## acc
##   0

calculateConfusionMatrix(predictions)

##           predicted
## true      setosa versicolor virginica -err.-
## setosa           0           0           0     0
## versicolor        0           0           0     0
## virginica         0          50           0    50
## -err.-            0          50           0    50
```

## The Ugly

```
model = train(learner, task, subset = c(1:45, 51:95, 101:110))
model

## Model for learner.id=classif.randomForest; learner.class=classif.randomForest
## Trained on: task.id = iris; obs = 100; features = 4
## Hyperparameters:

predictions = predict(model, task = task, subset = c(46:50, 96:100, 111:150))
predictions

## Prediction: 50 observations
## predict.type: response
## threshold:
## time: 0.00
##   id      truth  response
```

```
## 46 46      setosa      setosa
## 47 47      setosa      setosa
## 48 48      setosa      setosa
## 49 49      setosa      setosa
## 50 50      setosa      setosa
## 96 96 versicolor versicolor
## ... (50 rows, 3 cols)
```

```
performance(predictions, measures = acc)
```

```
## acc
## 0.84
```

```
calculateConfusionMatrix(predictions)
```

```
##           predicted
## true      setosa versicolor virginica -err.-
## setosa      5         0         0         0
## versicolor  0         5         0         0
## virginica   0         8        32         8
## -err.-      0         8         0         8
```

## Automatic Splitting

### Holdout

```
rdesc = makeResampleDesc(method = "Holdout", split = 2/3)
result = resample(learner, task, rdesc, measures = acc)
```

```
## [Resample] holdout iter 1: acc.test.mean=0.98
## [Resample] Aggr. Result: acc.test.mean=0.98
```

```
predictions = getRRPredictions(result)
performance(predictions, measures = acc)
```

```
## acc
## 0.98
```

```
calculateConfusionMatrix(predictions)
```

```
##           predicted
## true      setosa versicolor virginica -err.-
## setosa     17         0         0         0
## versicolor  0        20         1         1
## virginica   0         0        12         0
## -err.-      0         0         1         1
```

Using the holdout function:

```
result = holdout(learner, task, measures = acc, split = 2/3)
```

```
## [Resample] holdout iter 1: acc.test.mean=0.92
## [Resample] Aggr. Result: acc.test.mean=0.92
```

```
predictions = getRRPredictions(result)
performance(predictions, measures = acc)
```

```
## acc
## 0.92
```

```
calculateConfusionMatrix(predictions)
```

```
##           predicted
## true      setosa versicolor virginica -err.-
## setosa      13         0         0         0
## versicolor  0         18         1         1
## virginica   0         3         15        3
## -err.-     0         3         1         4
```

## Stratification

```
result = holdout(learner, task, measures = acc, split = 2/3, stratify = TRUE)
```

```
## [Resample] holdout iter 1: acc.test.mean=0.961
## [Resample] Aggr. Result: acc.test.mean=0.961
```

```
predictions = getRRPredictions(result)
performance(predictions, measures = acc)
```

```
## acc
## 0.9607843
```

```
calculateConfusionMatrix(predictions)
```

```
##           predicted
## true      setosa versicolor virginica -err.-
## setosa      17         0         0         0
## versicolor  0         17         0         0
## virginica   0         2         15         2
## -err.-     0         2         0         2
```

## Subsample

```
rdesc = makeResampleDesc(method = "Subsample", iters = 10, split = 2/3, predict = "both")
# or use the "subsample" function
result = resample(learner, task, rdesc, measures = acc)
```

```
## [Resample] subsampling iter 1: acc.test.mean=0.94
## [Resample] subsampling iter 2: acc.test.mean=0.94
## [Resample] subsampling iter 3: acc.test.mean= 1
## [Resample] subsampling iter 4: acc.test.mean=0.98
## [Resample] subsampling iter 5: acc.test.mean= 1
## [Resample] subsampling iter 6: acc.test.mean=0.94
## [Resample] subsampling iter 7: acc.test.mean=0.98
## [Resample] subsampling iter 8: acc.test.mean=0.94
## [Resample] subsampling iter 9: acc.test.mean=0.96
## [Resample] subsampling iter 10: acc.test.mean= 0.9
## [Resample] Aggr. Result: acc.test.mean=0.958
```

## Details for each iteration

```
getRRPredictions(result)
```

```
## Resampled Prediction for:  
## Resample description: subsampling with 10 iterations and 0.67 split rate.  
## Predict: both  
## Stratification: FALSE  
## predict.type: response  
## threshold:  
## time (mean): 0.00  
##   id      truth  response iter set  
## 1  79 versicolor versicolor   1 test  
## 2  37   setosa    setosa     1 test  
## 3 142 virginica virginica   1 test  
## 4  33   setosa    setosa     1 test  
## 5  70 versicolor versicolor   1 test  
## 6  76 versicolor versicolor   1 test  
## ... (1500 rows, 5 cols)
```

```
predictionList = getRRPredictionList(result)  
sapply(predictionList$test, performance, measures = acc)
```

```
## 1.acc 2.acc 3.acc 4.acc 5.acc 6.acc 7.acc 8.acc 9.acc 10.acc  
## 0.94 0.94 1.00 0.98 1.00 0.94 0.98 0.94 0.96 0.90
```

```
# this is also directly available in the resample result  
result$measures.test
```

```
##   iter acc  
## 1     1 0.94  
## 2     2 0.94  
## 3     3 1.00  
## 4     4 0.98  
## 5     5 1.00  
## 6     6 0.94  
## 7     7 0.98  
## 8     8 0.94  
## 9     9 0.96  
## 10    10 0.90
```

```
lapply(predictionList$test, calculateConfusionMatrix)
```

```
## $`1`  
##           predicted  
## true      setosa versicolor virginica -err.-  
## setosa      16         0         0         0  
## versicolor  0         18         1         1  
## virginica   0         2         13        2  
## -err.-      0         2         1         3  
##  
## $`2`  
##           predicted  
## true      setosa versicolor virginica -err.-  
## setosa      19         0         0         0  
## versicolor  0         16         2         2
```

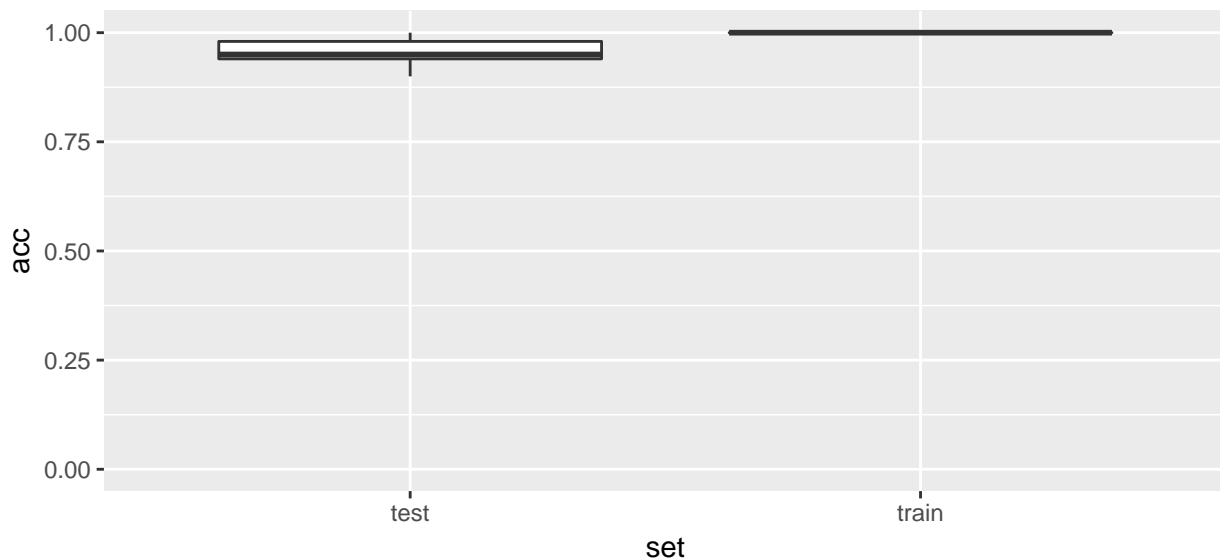
```

##  virginica      0      1     12     1
##  -err.-        0      1      2     3
##
##  $`3`
##          predicted
## true      setosa versicolor virginica -err.-
##  setosa      19      0      0      0
##  versicolor  0      19      0      0
##  virginica   0      0     12      0
##  -err.-     0      0      0      0
##
##  $`4`
##          predicted
## true      setosa versicolor virginica -err.-
##  setosa      17      0      0      0
##  versicolor  0     11      1      1
##  virginica   0      0     21      0
##  -err.-     0      0      1      1
##
##  $`5`
##          predicted
## true      setosa versicolor virginica -err.-
##  setosa      18      0      0      0
##  versicolor  0     18      0      0
##  virginica   0      0     14      0
##  -err.-     0      0      0      0
##
##  $`6`
##          predicted
## true      setosa versicolor virginica -err.-
##  setosa      15      0      0      0
##  versicolor  0     17      0      0
##  virginica   0      3     15      3
##  -err.-     0      3      0      3
##
##  $`7`
##          predicted
## true      setosa versicolor virginica -err.-
##  setosa      15      0      0      0
##  versicolor  0     21      1      1
##  virginica   0      0     13      0
##  -err.-     0      0      1      1
##
##  $`8`
##          predicted
## true      setosa versicolor virginica -err.-
##  setosa      17      0      0      0
##  versicolor  0     14      2      2
##  virginica   0      1     16      1
##  -err.-     0      1      2      3
##
##  $`9`
##          predicted
## true      setosa versicolor virginica -err.-

```

```
##   setosa      15      0      0      0
##   versicolor  0      18     1      1
##   virginica   0      1     15     1
##   -err.-     0      1      1      2
##
## $`10`
##           predicted
## true      setosa versicolor virginica -err.-
##   setosa    16      0      0      0
##   versicolor  0     15     1      1
##   virginica  0      4     14     4
##   -err.-    0      4      1      5
```

```
library(ggplot2)
ggplot(data.frame(acc = c(result$measures.train$acc, result$measures.test$acc),
                  set = rep(c("train", "test"), each = nrow(result$measures.train))),
       aes(set, acc)) + geom_boxplot() + ylim(0, 1)
```



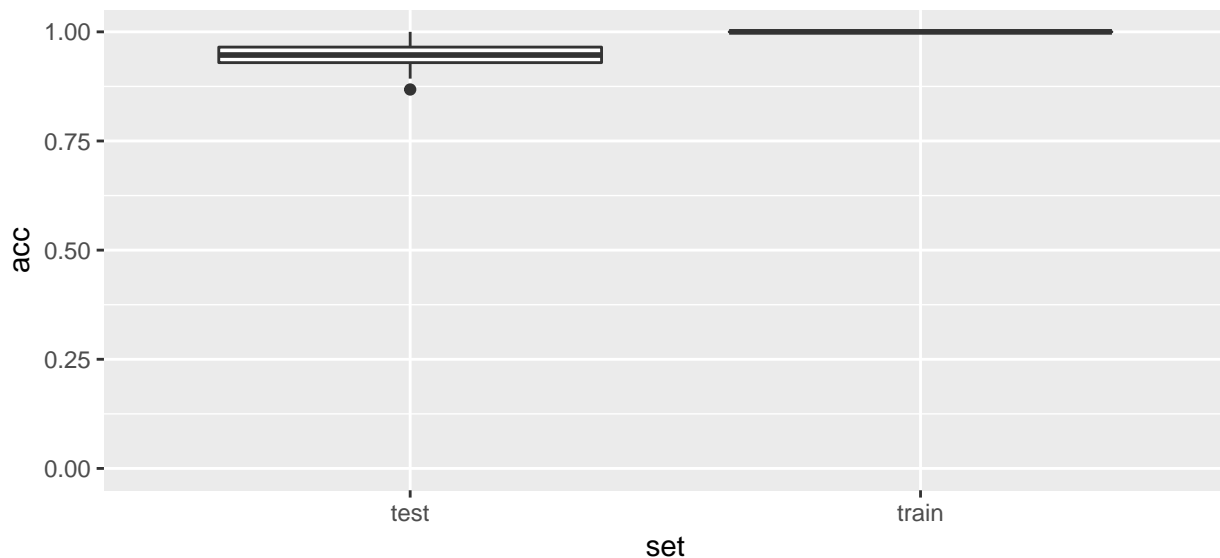
## Bootstrap

```
rdesc = makeResampleDesc(method = "Bootstrap", predict = "both")
# or use the "bootstrapOOB" function
result = resample(learner, task, rdesc, measures = acc)
```

```
## [Resample] OOB bootstrapping iter 1: acc.test.mean=0.946
## [Resample] OOB bootstrapping iter 2: acc.test.mean=0.979
## [Resample] OOB bootstrapping iter 3: acc.test.mean= 1
## [Resample] OOB bootstrapping iter 4: acc.test.mean=0.929
## [Resample] OOB bootstrapping iter 5: acc.test.mean=0.966
## [Resample] OOB bootstrapping iter 6: acc.test.mean=0.929
## [Resample] OOB bootstrapping iter 7: acc.test.mean=0.926
## [Resample] OOB bootstrapping iter 8: acc.test.mean=0.868
## [Resample] OOB bootstrapping iter 9: acc.test.mean=0.961
## [Resample] OOB bootstrapping iter 10: acc.test.mean=0.964
## [Resample] OOB bootstrapping iter 11: acc.test.mean=0.925
```

```
## [Resample] OOB bootstrapping iter 12: acc.test.mean=0.951
## [Resample] OOB bootstrapping iter 13: acc.test.mean=0.923
## [Resample] OOB bootstrapping iter 14: acc.test.mean=0.978
## [Resample] OOB bootstrapping iter 15: acc.test.mean=0.966
## [Resample] OOB bootstrapping iter 16: acc.test.mean=0.946
## [Resample] OOB bootstrapping iter 17: acc.test.mean=0.946
## [Resample] OOB bootstrapping iter 18: acc.test.mean=0.947
## [Resample] OOB bootstrapping iter 19: acc.test.mean=0.982
## [Resample] OOB bootstrapping iter 20: acc.test.mean=0.98
## [Resample] OOB bootstrapping iter 21: acc.test.mean= 1
## [Resample] OOB bootstrapping iter 22: acc.test.mean=0.962
## [Resample] OOB bootstrapping iter 23: acc.test.mean=0.944
## [Resample] OOB bootstrapping iter 24: acc.test.mean=0.932
## [Resample] OOB bootstrapping iter 25: acc.test.mean=0.893
## [Resample] OOB bootstrapping iter 26: acc.test.mean=0.919
## [Resample] OOB bootstrapping iter 27: acc.test.mean=0.944
## [Resample] OOB bootstrapping iter 28: acc.test.mean=0.947
## [Resample] OOB bootstrapping iter 29: acc.test.mean=0.943
## [Resample] OOB bootstrapping iter 30: acc.test.mean=0.964
## [Resample] Aggr. Result: acc.test.mean=0.949
```

```
ggplot(data.frame(acc = c(result$measures.train$acc, result$measures.test$acc),
  set = rep(c("train", "test"), each = nrow(result$measures.train))),
  aes(set, acc)) + geom_boxplot() + ylim(0, 1)
```



## Cross-Validation

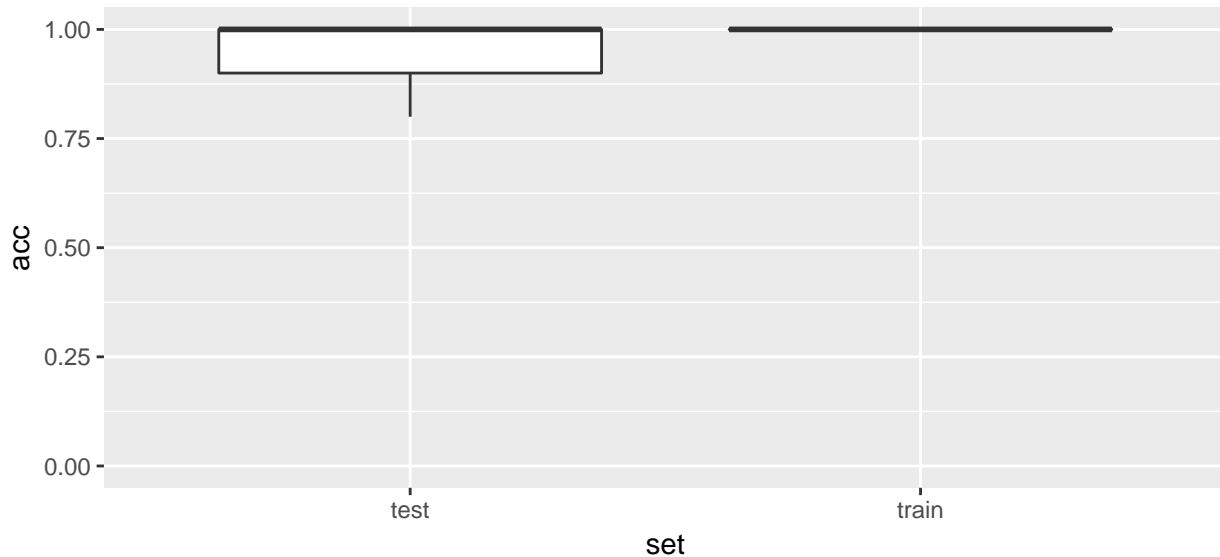
```
rdesc = makeResampleDesc(method = "CV", predict = "both")
# or use the "crossval" function
result = resample(learner, task, rdesc, measures = acc)
```

```
## [Resample] cross-validation iter 1: acc.test.mean= 1
## [Resample] cross-validation iter 2: acc.test.mean= 1
## [Resample] cross-validation iter 3: acc.test.mean=0.867
## [Resample] cross-validation iter 4: acc.test.mean= 0.8
```



```
## [Resample] cross-validation iter 5: acc.test.mean= 1
## [Resample] cross-validation iter 6: acc.test.mean=0.867
## [Resample] cross-validation iter 7: acc.test.mean= 1
## [Resample] cross-validation iter 8: acc.test.mean= 1
## [Resample] cross-validation iter 9: acc.test.mean= 1
## [Resample] cross-validation iter 10: acc.test.mean= 1
## [Resample] Aggr. Result: acc.test.mean=0.953
```

```
ggplot(data.frame(acc = c(result$measures.train$acc, result$measures.test$acc),
                    set = rep(c("train", "test"), each = nrow(result$measures.train))),
       aes(set, acc)) + geom_boxplot() + ylim(0, 1)
```



## Leave-One-Out Cross-Validation

```
# or set number of folds to n
rdesc = makeResampleDesc(method = "LOO", predict = "both")
result = resample(learner, task, rdesc, measures = acc)
```

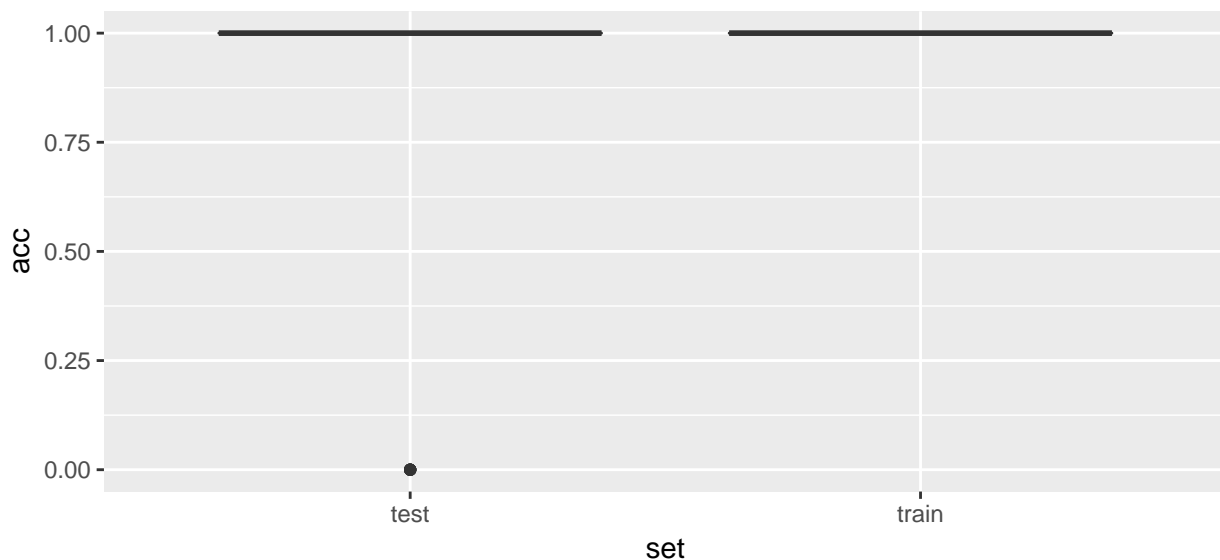
```
## [Resample] LOO iter 1: acc.test.mean= 1
## [Resample] LOO iter 2: acc.test.mean= 1
## [Resample] LOO iter 3: acc.test.mean= 1
## [Resample] LOO iter 4: acc.test.mean= 1
## [Resample] LOO iter 5: acc.test.mean= 1
## [Resample] LOO iter 6: acc.test.mean= 1
## [Resample] LOO iter 7: acc.test.mean= 1
## [Resample] LOO iter 8: acc.test.mean= 1
## [Resample] LOO iter 9: acc.test.mean= 1
## [Resample] LOO iter 10: acc.test.mean= 1
## [Resample] LOO iter 11: acc.test.mean= 1
## [Resample] LOO iter 12: acc.test.mean= 1
## [Resample] LOO iter 13: acc.test.mean= 1
## [Resample] LOO iter 14: acc.test.mean= 1
## [Resample] LOO iter 15: acc.test.mean= 1
## [Resample] LOO iter 16: acc.test.mean= 1
## [Resample] LOO iter 17: acc.test.mean= 1
```





```
## [Resample] L00 iter 126: acc.test.mean= 1
## [Resample] L00 iter 127: acc.test.mean= 1
## [Resample] L00 iter 128: acc.test.mean= 1
## [Resample] L00 iter 129: acc.test.mean= 1
## [Resample] L00 iter 130: acc.test.mean= 1
## [Resample] L00 iter 131: acc.test.mean= 1
## [Resample] L00 iter 132: acc.test.mean= 1
## [Resample] L00 iter 133: acc.test.mean= 1
## [Resample] L00 iter 134: acc.test.mean= 0
## [Resample] L00 iter 135: acc.test.mean= 1
## [Resample] L00 iter 136: acc.test.mean= 1
## [Resample] L00 iter 137: acc.test.mean= 1
## [Resample] L00 iter 138: acc.test.mean= 1
## [Resample] L00 iter 139: acc.test.mean= 1
## [Resample] L00 iter 140: acc.test.mean= 1
## [Resample] L00 iter 141: acc.test.mean= 1
## [Resample] L00 iter 142: acc.test.mean= 1
## [Resample] L00 iter 143: acc.test.mean= 1
## [Resample] L00 iter 144: acc.test.mean= 1
## [Resample] L00 iter 145: acc.test.mean= 1
## [Resample] L00 iter 146: acc.test.mean= 1
## [Resample] L00 iter 147: acc.test.mean= 1
## [Resample] L00 iter 148: acc.test.mean= 1
## [Resample] L00 iter 149: acc.test.mean= 1
## [Resample] L00 iter 150: acc.test.mean= 1
## [Resample] Aggr. Result: acc.test.mean=0.96
```

```
ggplot(data.frame(acc = c(result$measures.train$acc, result$measures.test$acc),
                    set = rep(c("train", "test"), each = nrow(result$measures.train))),
       aes(set, acc)) + geom_boxplot() + ylim(0, 1)
```



## Blocking

```
task = makeClassifTask(data = iris, target = "Species", blocking = iris$Species)
rdesc = makeResampleDesc(method = "CV", iters = 3)
result = resample(learner, task, rdesc, measures = acc)
```

```
## [Resample] cross-validation iter 1: acc.test.mean= 0
## [Resample] cross-validation iter 2: acc.test.mean= 0
## [Resample] cross-validation iter 3: acc.test.mean= 0
## [Resample] Aggr. Result: acc.test.mean= 0
```