

Deploying Predictive Models in the Cloud using Yhat

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NKN Tech Talk

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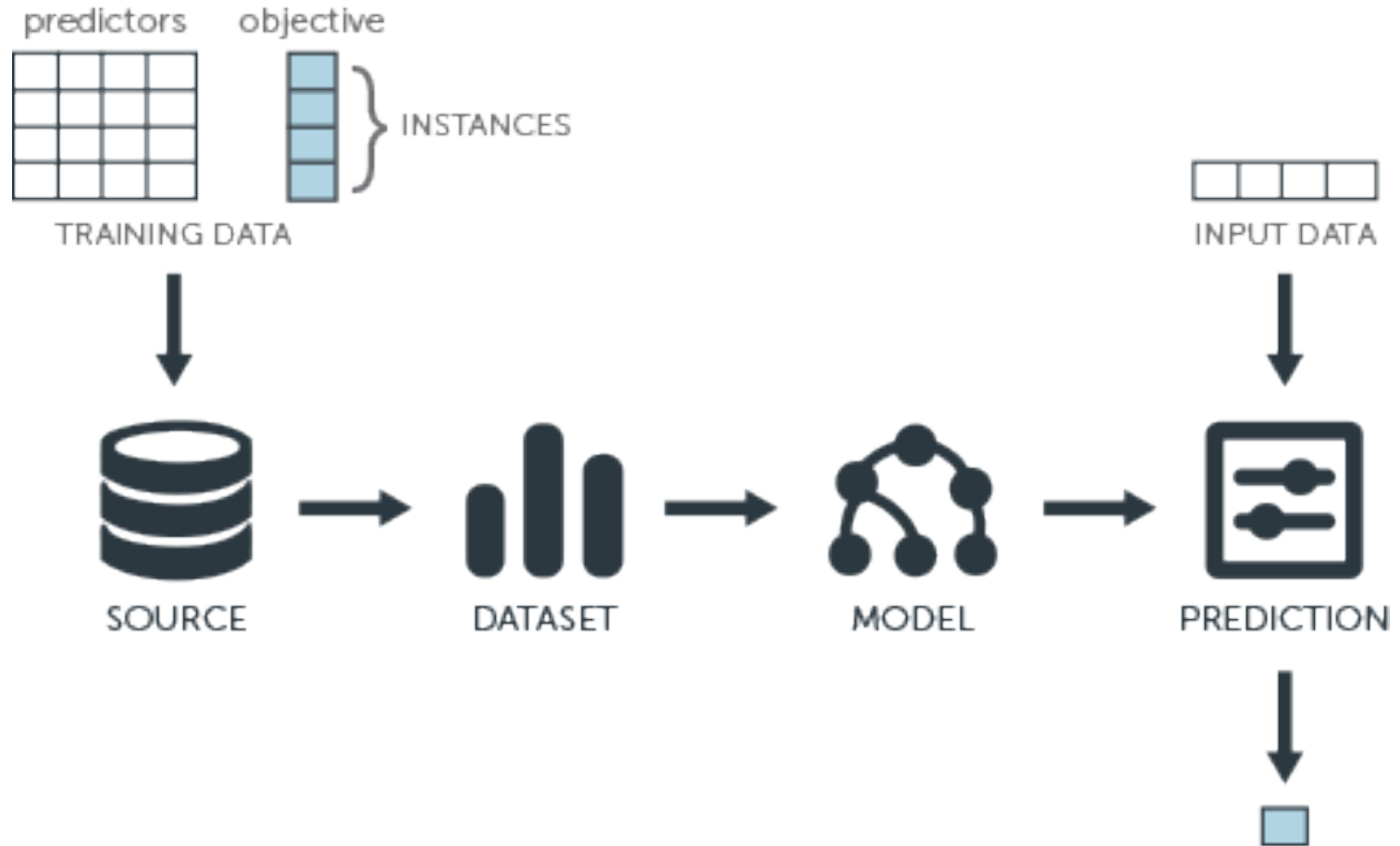
Machine Learning: Training Statistical Predictive Models in R



topepo.github.io/caret/

- *caret (Classification And Regression Training)* is a powerful R package for machine learning
 - Data Splitting
 - Training
 - Cross Validation
 - Prediction

Model Training and Prediction



Real Example – Wearable Sensor Data

github.com/sheneman/machinelearning

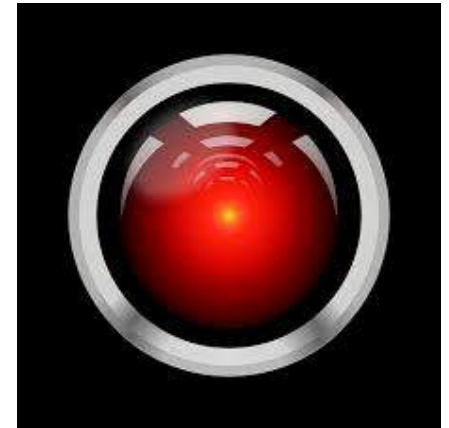
- Use *quantitative* raw data from wearable sensors to *qualitatively* classify human motion.
 - Velloso, E.; Bulling, A.; Gellersen, H.; Ugulino, W.; Fuks, H. Qualitative Activity Recognition of Weight Lifting Exercises. Proceedings of 4th International Conference in Cooperation with SIGCHI (Augmented Human '13) . Stuttgart, Germany: ACM SIGCHI, 2013. <http://groupware.les.inf.puc-rio.br/har>
- 160 variables, 20K observations
 - Predictors:
 - Subject ID, roll, pitch, yaw, acceleration, time
 - mean, stdev, kurtosis, skewness
 - Outcomes/objectives:
 - Classification: A,B,C,D,E



Training the Models

Method #1: Recursive Partitioning and Regression Trees (RPART):

```
control <- trainControl(method="cv", number=5)  
modelfit <- train(classe ~ ., data=new_training_data,  
                  trControl=control, method="rpart")
```

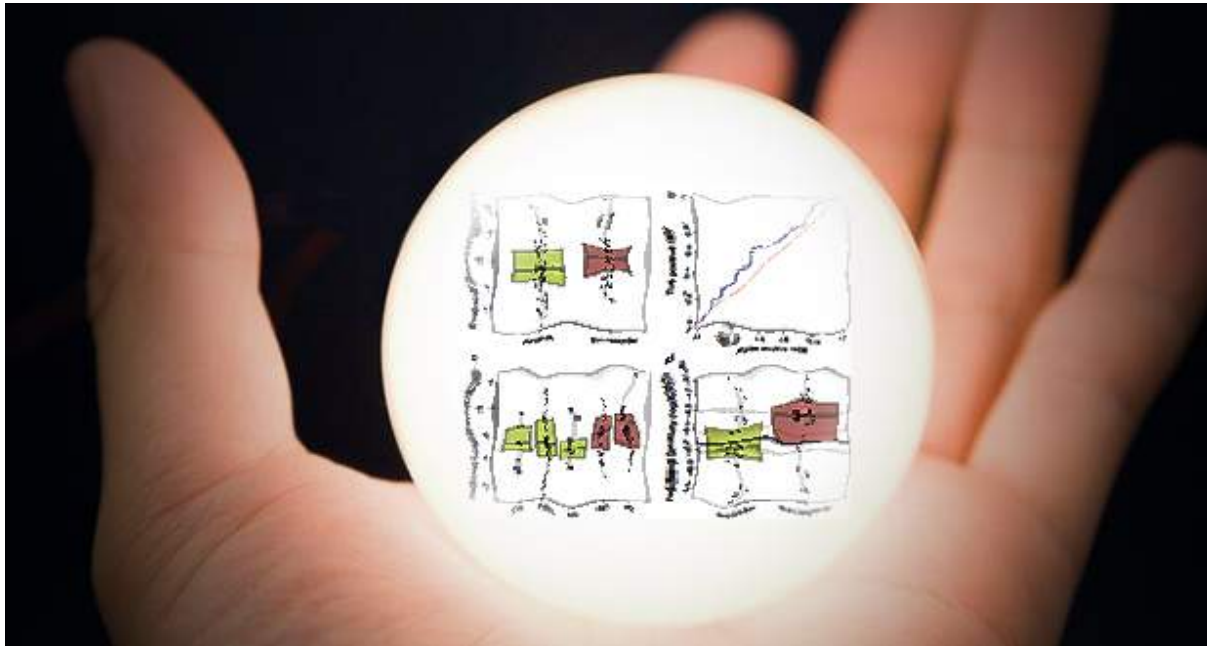


Method #2: Bootstrap Aggregation (BAGGING):

```
control <- trainControl(method="cv", number=5, verboseIter=TRUE)  
modelfit <- train(classe ~ ., data=new_training_data,  
                  trControl=control, method="treebag")
```

Prediction from Trained Models

```
predictions <- predict(modelfit, new_training_data)
```



Results

RPART Method

Aggregating results
Selecting tuning parameters
Fitting $cp = 0.0395$ on full training set
Confusion Matrix and Statistics

Reference

Prediction	A	B	C	D	E
A	5080	1581	1587	1449	524
B	81	1286	108	568	486
C	405	930	1727	1199	966
D	0	0	0	0	0
E	14	0	0	0	1631

Overall Statistics

Accuracy : **0.4956**

95% CI : (0.4885, 0.5026)

No Information Rate : 0.2844

P-Value [Acc > NIR] : < 2.2e-16

BAGGING Method

Aggregating results
Fitting final model on full training set
Confusion Matrix and Statistics

Reference

Prediction	A	B	C	D	E
A	5580	0	0	0	0
B	0	3797	1	0	0
C	0	0	3421	0	0
D	0	0	0	3216	0
E	0	0	0	0	3607

Overall Statistics

Accuracy : **0.9999**

95% CI : (0.9997, 1)

No Information Rate : 0.2844

P-Value [Acc > NIR] : < 2.2e-16

yhat



Rodeo

Python IDE for Data Science



scienceOPS

Publish Predictive Models



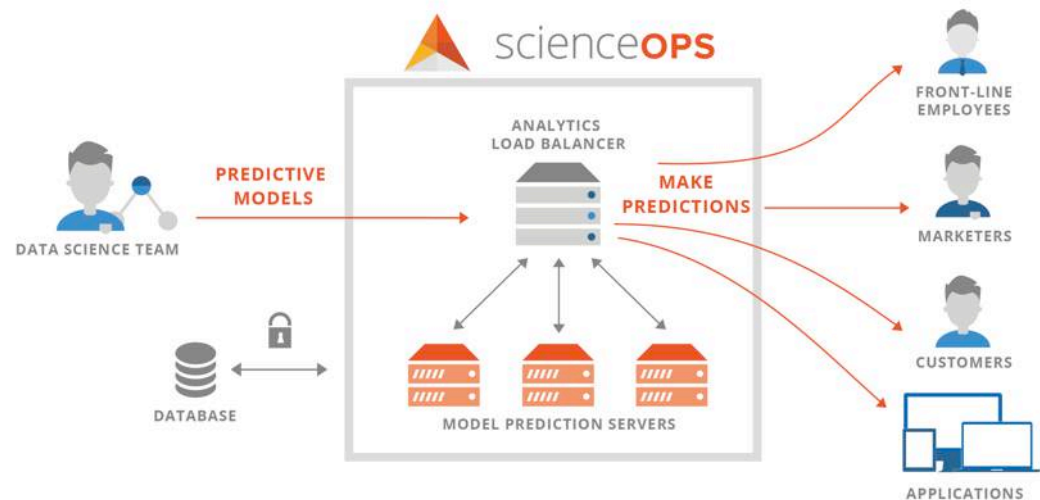
scienceCLUSTER

Distributed Computing Foo



WHAT IS SCIENCEOPS?

ScienceOps is a platform for deploying, managing and scaling predictive models and advanced decision-making algorithms into production. The platform is designed for production-level settings and utilizes a distributed architecture to allocate and optimize requests to models across the cluster.



yhatr



- <https://github.com/yhat/yhatr>
 - implicitly requires *rjson* package
- `model.require()`
 - specify library dependencies
- `model.transform()`
 - transform input data
- `model.predict()`
 - call the prediction function

Hello World in

R

```
library(yhatr)

model.predict <- function(request) {
  me <- request$name
  greeting<-paste ("Hello", me, "!")
  data.frame(greeting)
}

yhat.config <- c(
  username="YOUR_USERNAME",
  apikey="YOUR_APIKEY",
  env="https://sandbox.yhathq.com/"
)
yhat.deploy("HelloWorld")
```

Python

```
from yhat import Yhat, YhatModel , preprocess

class HelloWorld(YhatModel):
    @preprocess(in_type=dict, out_type=dict)
    def execute(self, data):
        me = data['name']
        greeting = "Hello " + str(me) + "!"
        return { "greeting": greeting }

yh = Yhat("YOUR_USERNAME", "YOUR_APIKEY", "https://sandbox.yhathq.com/")
yh.deploy("HelloWorld", HelloWorld, globals())
```

* In R, all input and output to model.predict() must be a *Data Frame* type

Examples:

- Hello World!
- Custom NKN Example
- Beer Selector



Applications

- [MaaS](#) – *Model as a Service* paradigm
- REST API
 - Build applications against standard API
 - Simulate 3rd party app development from your model
 - Deploy to one or millions of users
 - A Form of Data/Model Interoperability:
 - JSON + REST Web Service
 - Build interesting composite apps using predictions from multiple models!
- Auto re-deploy model as additional data streams in:
 - Weather, Climate, Precision Agriculture, Hazard Prediction, Climate Hot Spots, Social Media
- Horizontally Scalable on Cloud Infrastructure

*“Even if NKN does not use Yhat specifically, this kind of idea is a wonderful example of the **NKN Data Observatory** concept in action. Namely, Yhat can leverage existing data to do **new science via data/model interoperability**.”*

Exposing standard web service APIs are the key.”

-- Famous Olde Quote

Thank You!

