

Single function CRUG
Meetup
Pmml() and onnx()

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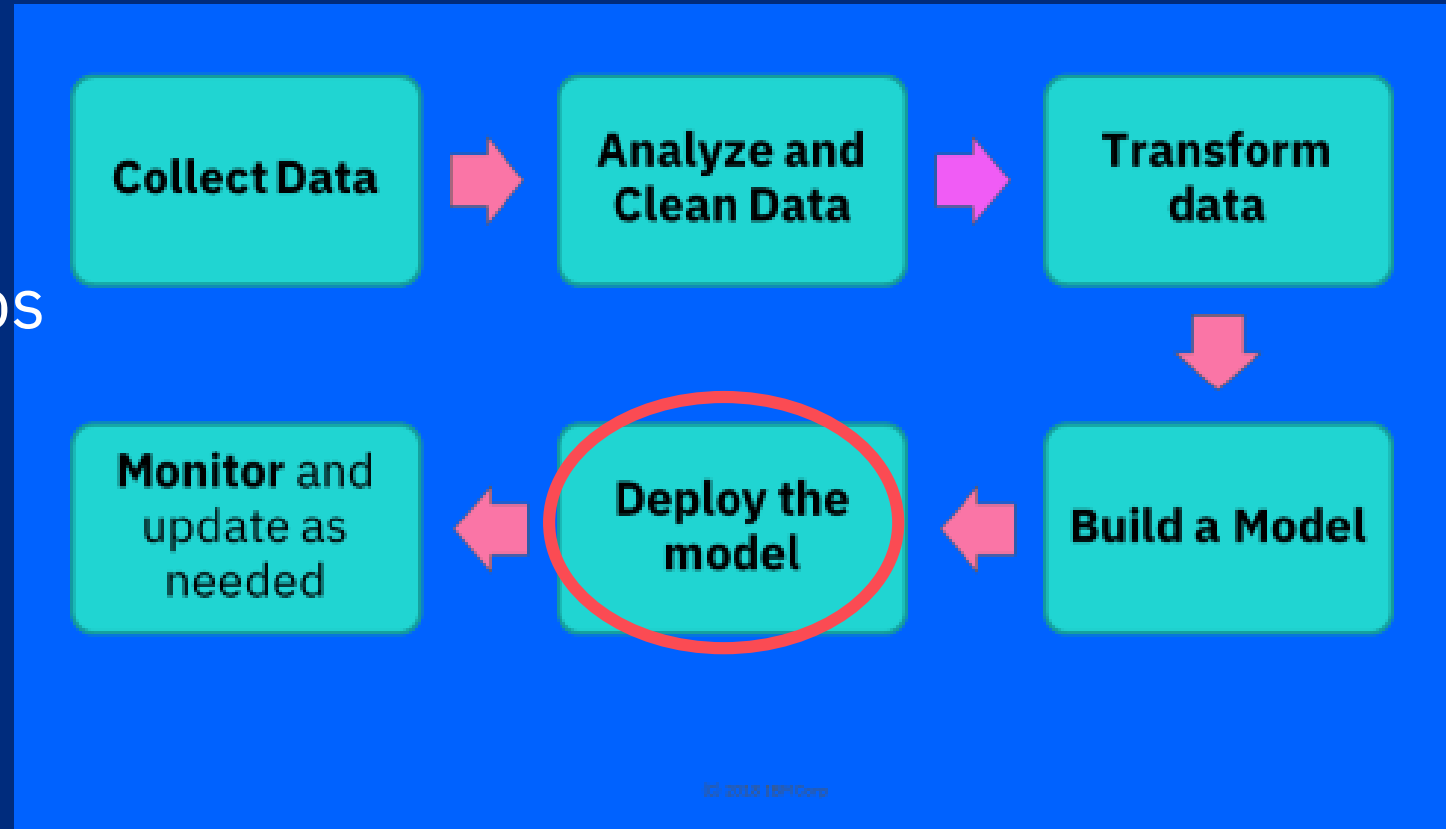
ML Model deployment

Challenges:

- Different Teams
- Different environments
- Need to keep data prep steps

A Solution: open standards

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DMG and PMML



Data Mining Group (dmg.org) since 1990's

Predictive Model Markup Language

- An Open Standard for **XML** Representation
- Over 30 vendors and organizations
- 17 models + combinations + transforms + ...
- dmg.org/pmml



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Main Components of PMML



Header

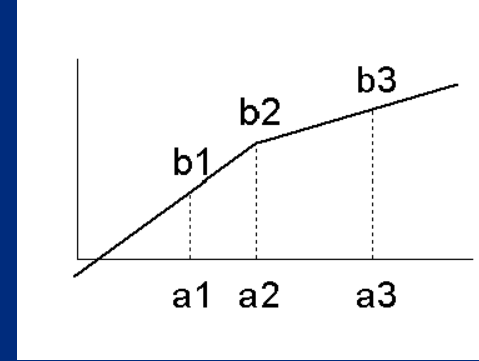
Data Dictionary

Transformation Dictionary

Model(s)

Transformations

- **NormContinuous:** piece-wise linear transform
- **NormDiscrete:** map a categorical field to a set of dummy fields
- **Discretize:** binning
- **MapValues:** map one or more categorical fields into another categorical one
- **Functions:** built-in and user-defined
- Other transformations



*favorite_pets.sav [DataSet2] - IBM SPSS Statistics

pet_d2 is 1 if (pet = 2), 0 otherwise.

	pet	pet_d1	pet_d2	pet_d3	pet_d4
1	1	1	0	0	0
2	2	0	1	0	0
3	3	0	0	1	0
4	4	0	0	0	1

PMML 4.4 Models

- **Anomaly Detection (new)**
- **Association Rules Model**
- **Clustering Model**
- **General Regression**
- **Naïve Bayes**
- **Nearest Neighbor Model**
- **Neural Network**
- **Regression**
- **Tree Model**
- **Baseline Model**
- **Bayesian Network**
- **Gaussian Process**
- **Ruleset**
- **Scorecard**
- **Sequence Model**
- **Support Vector Machine**
- **Time Series**
- **Mining Model:** composition or ensemble (or both) of models



R package pmml

There was also “**pmmlTransformations**”, now merged into pmml
<https://cran.r-project.org/package=pmml>

Recommends: `ada`, `amap`, `arules`, `caret`, `clue`, `data.table`, `gbm`,
`glmnet`, `neighbr`, `nnet`, `rpart`, `randomForest`, `kernlab`, `e1071`,
`testthat`, `survival`, `xgboost`, `knitr`

Maintained by Dmitriy Bolotov and others from Software AG

Build and save a decision tree (C&RT) model predicting Species class:

```
> irisTree <- rpart( Species~., iris )  
> saveXML( pmml( irisTree ), "IrisTree.xml" )
```

More details for pmml()

Usage

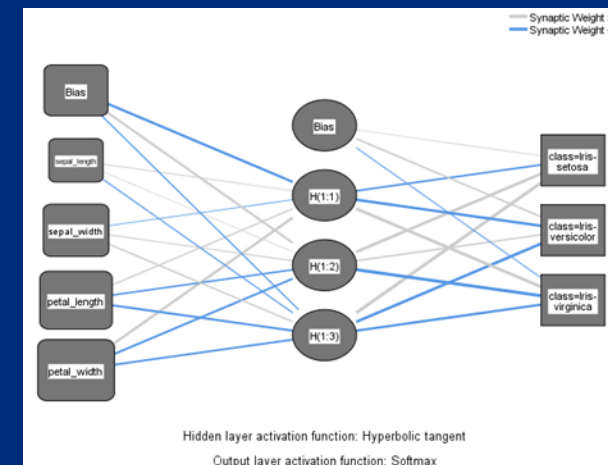
```
pmml(  
  model = NULL,  
  model_name = "R_Model",  
  app_name = "SoftwareAG PMML Generator",  
  description = NULL,  
  copyright = NULL,  
  transforms = NULL,  
  ...  
)
```

Method **pmml()** can be used to do any of the following:

- Export a model in PMML
- Export data transforms in PMML
- Merge data transforms into a PMML model

An example PMML – Data Dictionary, Transformations

```
▼<DataDictionary numberOfFields="5">
  ▼<DataField name="class" optype="categorical" dataType="string">
    <Value value="Iris-setosa"/>
    <Value value="Iris-versicolor"/>
    <Value value="Iris-virginica"/>
  </DataField>
  <DataField name="sepal_length" optype="continuous" dataType="double"/>
  <DataField name="sepal_width" optype="continuous" dataType="double"/>
  <DataField name="petal_length" optype="continuous" dataType="double"/>
  <DataField name="petal_width" optype="continuous" dataType="double"/>
</DataDictionary>
```

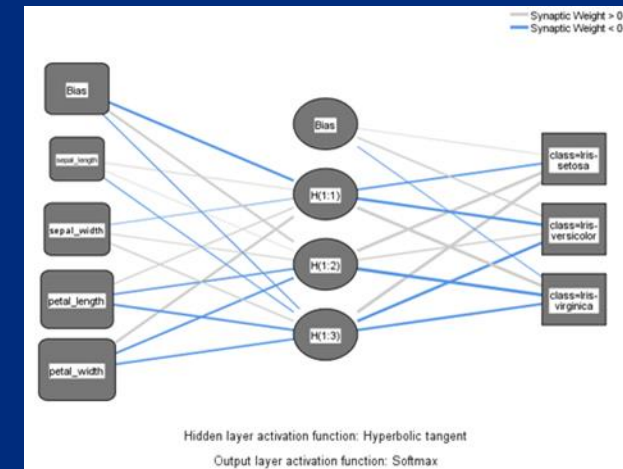


```
▼<DerivedField optype="categorical" dataType="double" name="classValue2">
  <NormDiscrete field="class" value="Iris-virginica"/>
</DerivedField>
▼<DerivedField optype="continuous" dataType="double" name="sepal_lengthNorm">
  ▼<NormContinuous field="sepal_length">
    <LinearNorm orig="4.3" norm="-1.84285714285714"/>
    <LinearNorm orig="7.7" norm="2.3204081632653"/>
  </NormContinuous>
</DerivedField>
▼<DerivedField optype="continuous" dataType="double" name="sepal_widthNorm">
  ▼<NormContinuous field="sepal_width">
    <LinearNorm orig="2" norm="-2.48539690378995"/>
    <LinearNorm orig="4.4" norm="3.13131926296699"/>
  </NormContinuous>
</DerivedField>
```

Example PMML – Neural Network MiningSchema and inputs

```
▼<NeuralNetwork functionName="classification" activationFunction="tanh">
  ▼<MiningSchema>
    <MiningField name="sepal_length"/>
    <MiningField name="sepal_width"/>
    <MiningField name="petal_length"/>
    <MiningField name="petal_width"/>
    <MiningField name="class" usageType="predicted"/>
  </MiningSchema>
  ▼<NeuralInputs>
    ▼<NeuralInput id="0">
      ▼<DerivedField optype="continuous" dataType="double">
        <FieldRef field="sepal_lengthNorm"/>
      </DerivedField>
    </NeuralInput>
    ▼<NeuralInput id="1">
      ▼<DerivedField optype="continuous" dataType="double">
        <FieldRef field="sepal_widthNorm"/>
      </DerivedField>
    </NeuralInput>
```

Predictors



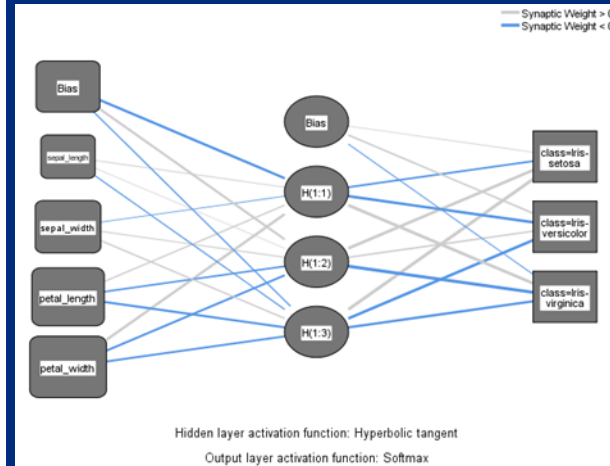
Example PMML - Neural Network hidden layer and outputs

```
▼<Neuron id="6" bias="-0.69138649428932">
  <Con from="0" weight="-0.57324998362272"/>
  <Con from="1" weight="0.892806772564007"/>
  <Con from="2" weight="-1.23192787546061"/>
  <Con from="3" weight="-1.19705013526962"/>
</Neuron>
</NeuralLayer>
▼<NeuralLayer numberOfNeurons="3" activationFunction="identity" normalizationMethod="softmax">
  ▼<Neuron id="7" bias="0.101922887283541">
    <Con from="4" weight="-1.05690948855012"/>
    <Con from="5" weight="2.00228899161664"/>
    <Con from="6" weight="3.31278374396491"/>
  </Neuron>
  ▼<Neuron id="8" bias="0.917636281284728">
    <Con from="4" weight="-1.47230776836775"/>
    <Con from="5" weight="0.905795272070893"/>
    <Con from="6" weight="-1.60793177845373"/>
  </Neuron>
  ▼<Neuron id="9" bias="-0.2772471777484">
    <Con from="4" weight="2.22290439134024"/>
    <Con from="5" weight="-2.43960637239511"/>
    <Con from="6" weight="-1.32214182019044"/>
  </Neuron>
</NeuralLayer>
▼<NeuralOutputs>
  ▼<NeuralOutput outputNeuron="7">
    ▼<DerivedField optype="categorical" dataType="double">
      <FieldRef field="classValue0"/>
    </DerivedField>
  </NeuralOutput>
  ▼<NeuralOutput outputNeuron="8">
    ▼<DerivedField optype="categorical" dataType="double">
      <FieldRef field="classValue1"/>
    </DerivedField>
  </NeuralOutput>
  ▼<NeuralOutput outputNeuron="9">
    ▼<DerivedField optype="categorical" dataType="double">
      <FieldRef field="classValue2"/>
    </DerivedField>
  </NeuralOutput>
</NeuralOutputs>
```

Hidden layer neuron

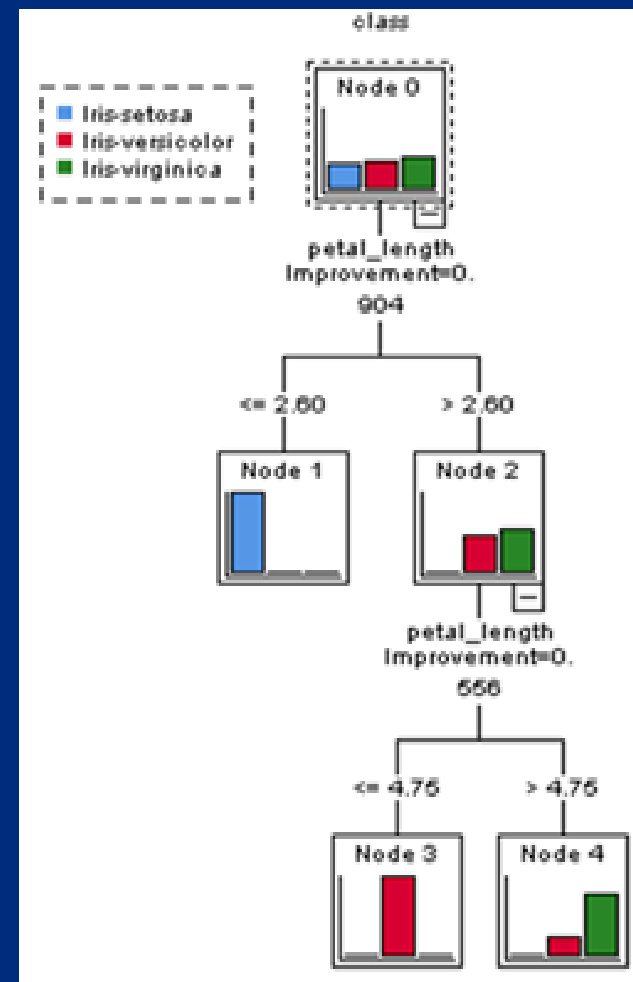
*Output
Layer
Neurons*

*Connecting
target to the
neurons*



Example PMML for a Tree Model

```
<Node id="0"> <True/>
  <Node id="1" score="Iris-setosa" recordCount="50.0">
    <SimplePredicate field="petal_length" operator="lessOrEqual"
      value="2.6"/>
    <ScoreDistribution value="Iris-setosa" recordCount="50.0"/>
    <ScoreDistribution value="Iris-versicolor" recordCount="0.0"/>
    <ScoreDistribution value="Iris-virginica" recordCount="0.0"/>
  </Node>
  <Node id="2">
    <SimplePredicate field="petal_length" operator="greaterThan"
      value="2.6"/>
    <Node id="3" score="Iris-versicolor" recordCount="40.0">
      <SimplePredicate field="petal_length"
        operator="lessOrEqual" value="4.75"/>
```



PMML Powered

From

<http://dmg.org/pmml/products.html>:

Alpine Data

Angoss

BigML

Equifax

Experian

FICO

Fiserv

Frontline Solvers

GDS Link

IBM (Includes SPSS)

IBM Developer

JPMML

KNIME

KXEN

Liga Data

Microsoft

MicroStrategy

NG Data

Open Data

Opera

Pega

Pervasive Data Rush

Predixion Software

Rapid I

R

Salford Systems (Minitab)

SAND

SAS

Software AG (incl. Zementis)

Spark

Sparkling Logic

Teradata

TIBCO

WEKA



ONNX: Open Neural Network eXchange – onnx.ai



Since 2017; Protobuf format

Covers DL and traditional ML

Active work by many companies

@SvetaLevitan



```
library(onnx)

graph_def <- make_graph(
  nodes = list(
    make_node("FC", list("X", "W1", "B1"), list("H1")),
    make_node("Relu", list("H1"), list("R1")),
    make_node("FC", list("R1", "W2", "B2"), list("Y"))
  ),
  name = "MLP",
  inputs = list(
    make_tensor_value_info('X' , onnx$TensorProto$FLOAT, list(1L)),
    make_tensor_value_info('W1', onnx$TensorProto$FLOAT, list(1L)),
    make_tensor_value_info('B1', onnx$TensorProto$FLOAT, list(1L)),
    make_tensor_value_info('W2', onnx$TensorProto$FLOAT, list(1L)),
    make_tensor_value_info('B2', onnx$TensorProto$FLOAT, list(1L))
  ),
  outputs = list(
    make_tensor_value_info('Y', onnx$TensorProto$FLOAT, list(1L))
  )
)

print_readable(graph_def)
```


Conclusions

Model deployment is an important part of ML lifecycle

Data Mining Group works on open standards for model deployment

PMML eases deployment for supported models and data prep

ONNX is becoming a de-facto standard for Deep Learning

Links and contact

[Dmg.org/pmml](https://dmg.org/pmml) dmg.org/pfa onnx.ai codait.org
[@SvetaLevitan](https://twitter.com/SvetaLevitan) slevitan@us.ibm.com