

# Vowpal Wabbit on Spark

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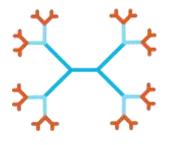




# Microsoft Machine Learning for Apache Spark









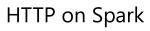
Vowpal Wabbit

Cognitive Service

LightGBM

Spark Serving







CNTK



Lime



Binding Autogeneration

# **Vowpal Wabbit on Spark | Goals**

Easy to use 

Seamless integration w/ Spark ML

Blazing fast 

Linear out-of-core learning

Distributed 

Data parallel w/o manual cluster configuration

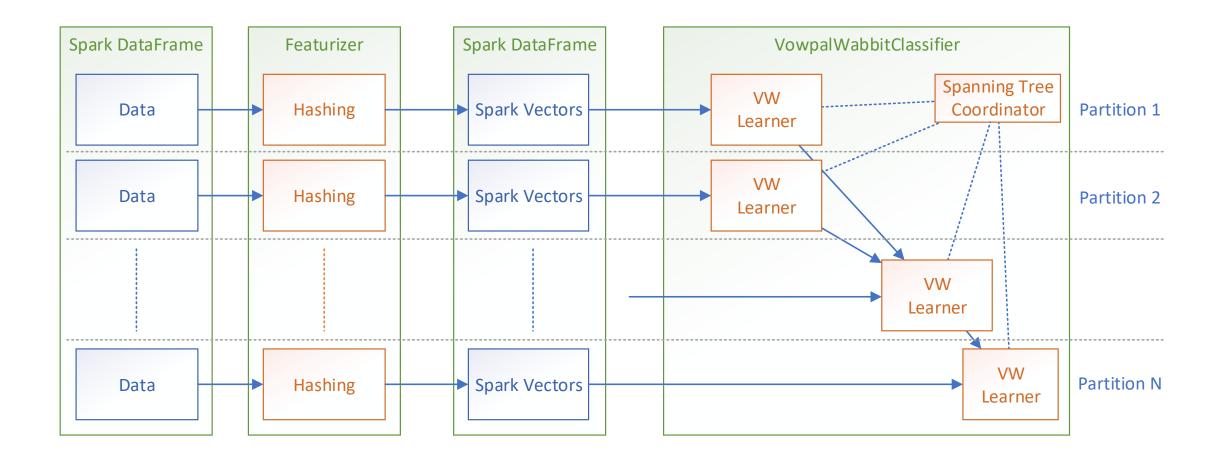
Scalable → 10TB experiments







# **Vowpal Wabbit on Spark | Architecture**









# **Vowpal Wabbit on PySpark | Featurize**

from mmlspark.vw import VowpalWabbitFeaturizer

```
featurizer = VowpalWabbitFeaturizer(
  inputCols=['sepal length', 'sepal width', 'petal length'],
  outputCol='features')

df_train = featurizer.transform(df_input)
```







# Vowpal Wabbit on Spark | Feature Engineering

| Туре                        | Sample Data                          | Vowpal Wabbit Format             | Feature Inde[x ices]                | Feature<br>Value(s) |
|-----------------------------|--------------------------------------|----------------------------------|-------------------------------------|---------------------|
| Numeric                     | 3.2                                  | col1 col1:3.2                    | hash(col1)                          | 3.2                 |
| Boolean                     | True                                 | col1 col1:1                      | hash(col1)                          | 1                   |
| String                      | NewYork                              | col1 NewYork:1                   | hash(New York)                      | 1                   |
| Array of String             | ["Hello","world"]                    | col1 Hello:1 World:1             | hash(Hello), hash(World)            | 1, 1                |
| Map of<br>String to String  | {"Income":"High",<br>"Age":"Middle"} | col1 IncomeHigh:1<br>AgeMiddle:1 | hash(IncomeHigh)<br>hash(AgeMiddle) | 1, 1                |
| Map of<br>String to Numeric | {"Speed":12.3, "Velocity":3}         | col1 Speed:12.3<br>Velocity:3    | hash(Speed)<br>hash(Velocity)       | 12.3, 3             |

- 1<sup>st</sup> namespace character = feature group used for interactions, n-grams, ...
- · Numeric: Double, Float, Integer, Long, Short, Byte







# Vowpal Wabbit on PySpark | Train & Predict

from mmlspark.vw import VowpalWabbitClassifier

```
estimator = VowpalWabbitClassifier(
    args="--holdout_off --loss_function logistic --progress 1000000",
    learningRate=1.0,
    powerT=0.0,
    l1=1e-8)

model = estimator.fit(df_train)

predictions = model.transform(df_test)
```

# Vowpal Wabbit on Spark | Hyper-parameter sweeps

Spark ML integration

#### Frequent parameters

· learningRate, powerT, I1, I2, interactions

Parallel cross-validation

```
val vw = new VowpalWabbitClassifier()
    .setArgs("--link=logistic")
val paramGrid = new ParamGridBuilder()
   .addGrid(vw.learningRate, Array(0.5, 0.05, 0.001))
   .addGrid(vw.numPasses, Array(1, 3, 5))
   .build()
val cv = new CrossValidator()
  .setEstimator(vw)
  .setEvaluator(new BinaryClassificationEvaluator)
  .setEstimatorParamMaps(paramGrid)
  .setNumFolds(3)
  .setParallelism(2)
val model = cv.fit(dataset)
model.transform(dataset)
val auc = model.avgMetrics(0)
```







# **Vowpal Wabbit on Spark | Bandit Support**

#### Coming soon...

#### Multi-line examples

- Actions
- Multi-class

#### Label support

· action:cost:probability

Nested structures







# Thanks!







# **Vowpal Wabbit on Spark | Features**

#### Distributed AllReduce training

- Data partitions → compute jobs
- · Limited by number of available executors

Multi-pass using VW cache file on disk

#### Performance metrics exposed as data frame

· average loss, total number of features,...

#### Native Model Export







# **Vowpal Wabbit on Spark | Features**

### Machine learning tasks

- Binary classification
- Regression

#### Data input

- Sparse and dense features using Spark Vector
- Example weights
- · Namespace support by using multiple columns









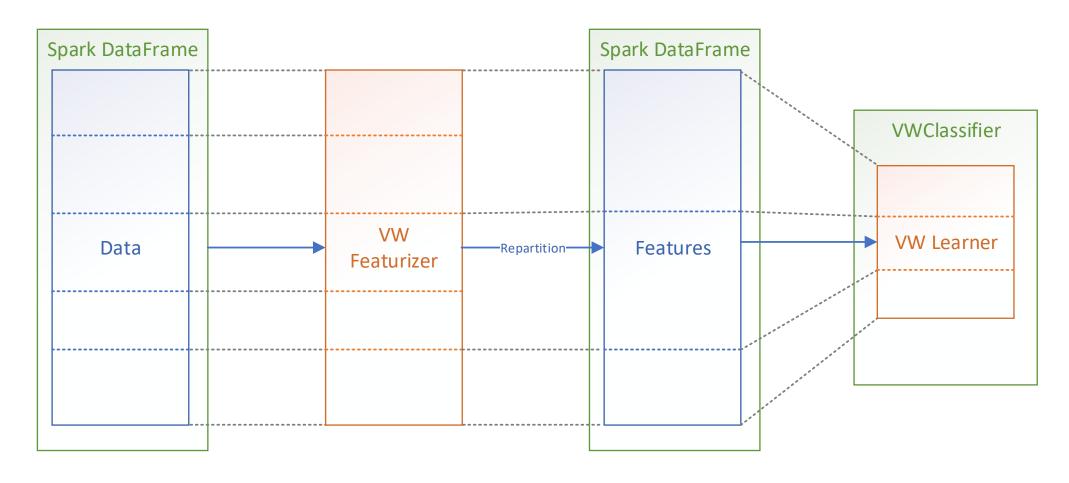
#### Suite of algorithms

- Supervised
- Unsupervised
- Reinforcement learning
  - Azure Cognitive Service Personalizer

#### Strengths

- Speed
- Memory efficiency
- Sparse features

# **Vowpal Wabbit on Spark | Optimizations**



Repartitioning featurized data can reduce parallel training jobs







# Vowpal Wabbit on Spark | Feature Engineering

```
new VowpalWabbitFeaturizer()
    .setInputCols(Array("title", "body"))
    .setOutputCol("features")
    .transform(df)
```

| Option                      | Description  |  |  |
|-----------------------------|--|--|--|
| SumCollisions               | Summed if true, otherwise the first is kept.                                   |  |  |
| Seed                        | Seed used for hashing.   |  |  |
| StringSplitInputCols        | String column values split on whitespace.                                      |  |  |
| PreserveOrderNumBits        | Bits used to preserve order. If >0 inserts position at the highest order bits. |  |  |
| PrefixStringsWithColumnName | Prefixes string features with column name (default: true).                     |  |  |







# **Vowpal Wabbit on Spark | N-Grams**

#### Vowpal Wabbit supports n-gram generation at learning time

Reduces memory footprint significantly

#### **Obstacle**

- · Features must be presented in order of occurrence (e.g. text) for n-gramming
- · Spark Sparse Vectors require **feature indices** to be sorted
- After hashing input text order is lost (indices = hash of word)

#### **Solution**

- Create composite feature indices (position + hash)
- Strip position for learning







## **Vowpal Wabbit on Spark | N-Grams**

```
new VowpalWabbitFeaturizer()
    .setStringSplitInputCols(Array("title", "body"))
    .setPreserveOrderNumBits(4)
    .setNumBits(20)
    .setPrefixStringsWithColumnName(false)
    .transform(df)
```

Title: "Hello World"

0000 0000 0000 0011 0001 0010 1000 0110

0000 0100 0000 0001 0100 0000 0100 0101

Position (4 bits)

Hash (20 bits)

Position (4 bits)

Hash (20 bits)





