

Big Data with ADAMS

What the heck is ADAMS?





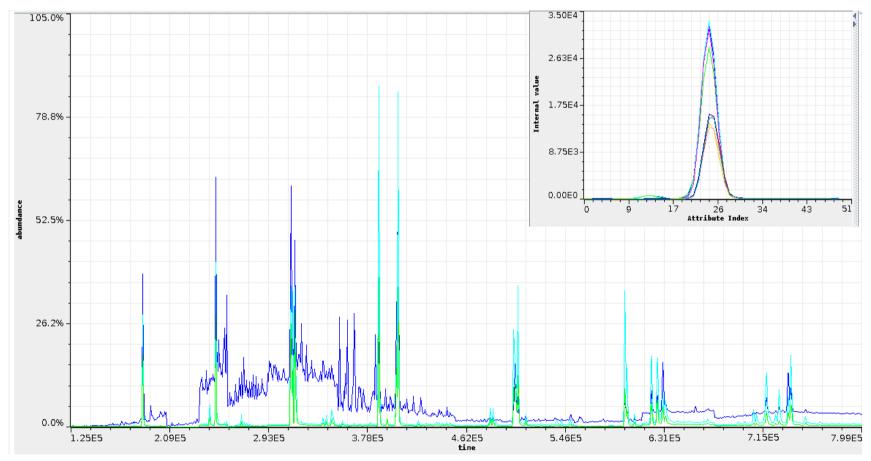
- University of Waikato, NZ since 2003
- Senior Research Programmer
- Lead developer of ADAMS
- Other projects
 - WEKA
 - WEKA MOOC
 - python-weka-wrapper
 - MEKA
- More info

http://www.cms.waikato.ac.nz/~fracpete/



How things started

• PAH domain (poly-aromatic hydrocarbons)





GC-MS Challenges

- Multi-dimensional (each GC point has 2-dim MS data attached)
- Retention time shifts (x-axis, linear and non-linear)
- Baseline shifts (y-axis)
- Noise (impurities in samples)
- Coelution (shoulder peaks)
- Varying sample rates (x-axis)
- Hardware problems (detector saturation)
- Multiple compounds to predict



Solution



- Extensive pre-processing
- Pre-processing/prediction in parallel for various compounds
 - → Workflow engine, but which one?



THE UNIVERSITY OF WAIKATO Te Whare Wānanga o Waikato

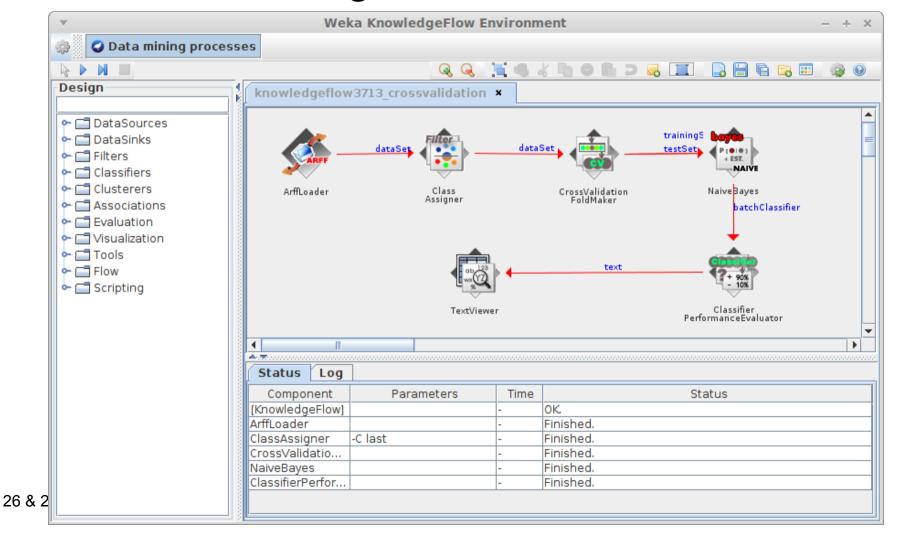
Workflow systems

- most workflow systems are "canvas-based"
 - user places operators on canvas
 - manually connects operators
- older system were DAG-based
 - eg using a tree-layout
- following examples
 - cross-validation of classifier on single dataset
 - output of results



Workflow systems (2)

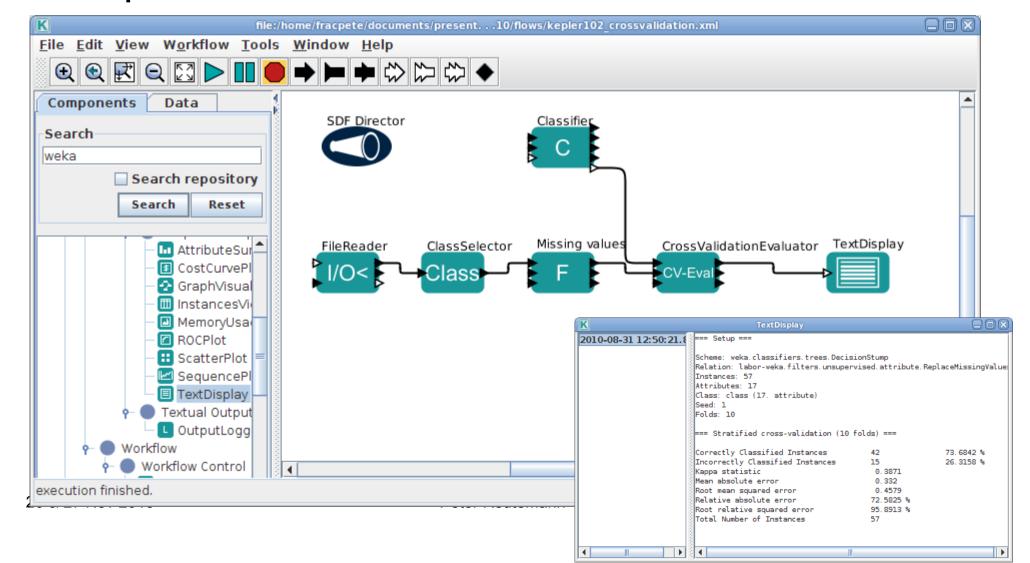
WEKA KnowledgeFlow 3.7.13





Workflow systems (3)

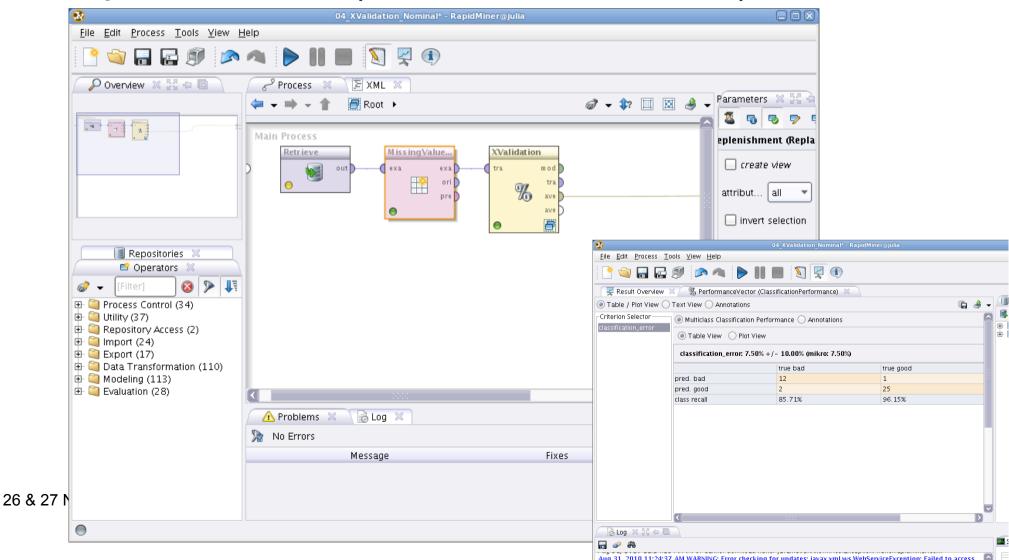
Kepler 1.0.2





Workflow systems (4)

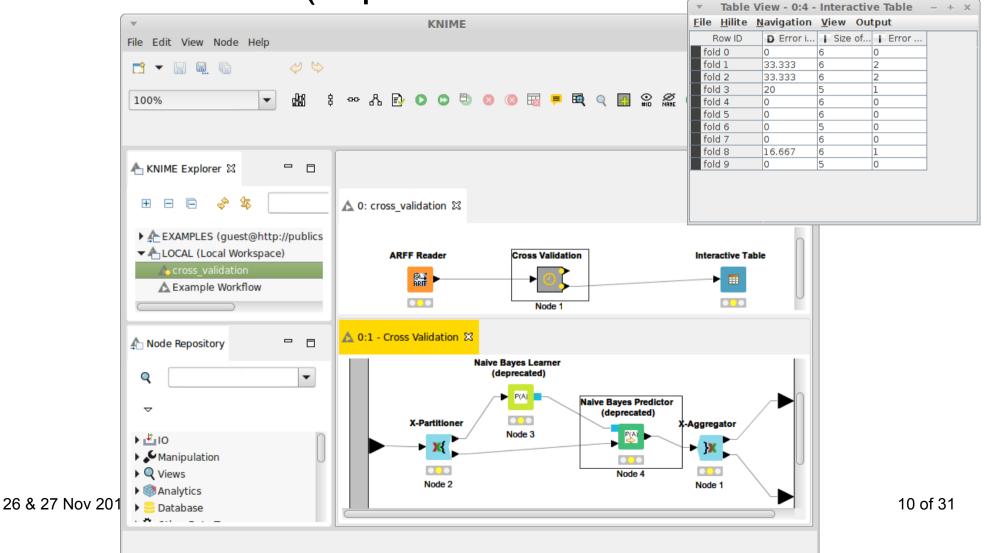
RapidMiner 5.0 ("free" 6.5 is useless)





Workflow systems (5)

KNIME 3.0.1 (imported from 2.3.1)





Why another workflow?

- KnowledgeFlow too WEKA-centric
- Initial development with Kepler
- Canvas-based set up is tedious
- Only minimal subset of functionality necessary (e.g., no grid computing)
- Primary connection types: 1-to-1 and 1-to-n
 - → prototype implementation

What is ADAMS?



- Java, GPLv3
- Base modules

access, core, compress, event, excel, gnuplot, groovy, imaging (+boofcv, imagej, imagemagick, openimaj), jython, latex, maps, meta, moa, net, odf, osm, pdf, r, random, spreadsheet, timeseries, twitter, visualstats, weka

Add-ons modules

heatmap, image-webservice, jooq, meka, rats, video, webservice, weka-webservice

Incubator modules

nlp, jclouds, openstack, openml, javacv, ...

Flow



- Operators are called "actors"
- Actors arranged in tree, no connections
- Actor "handlers" nest other actors
 - e.g., sequence of actors
- Control actors control data flow
 - e.g., branch 🖵, tee 🔳, if-then-else 🗐, switch 🖰
- Input/output defines
 - standalone CA, source CA, transformer CA, sink

Flow (2)

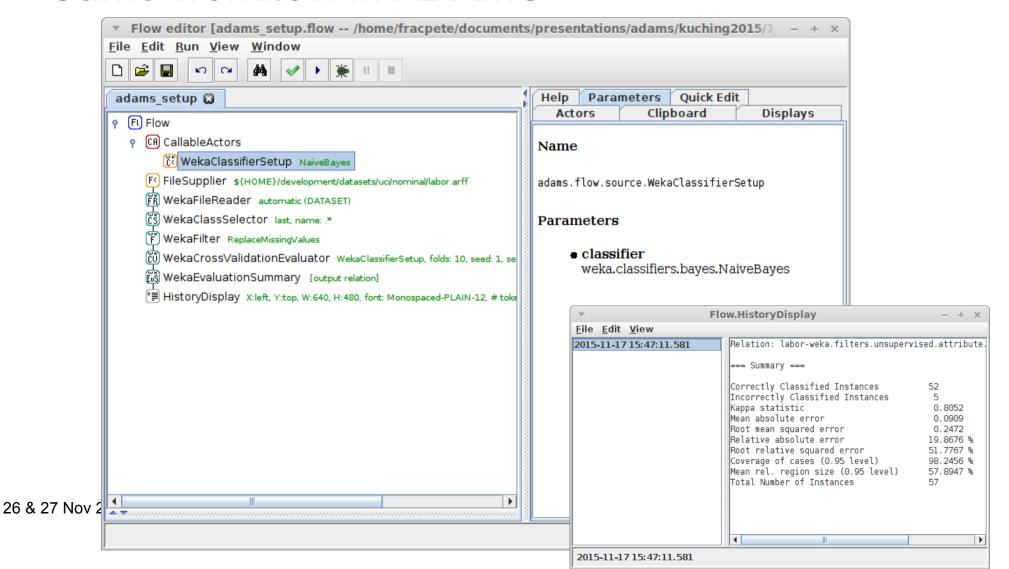


- Data-driven system, but events possible
- Tree only supports 1-to-n connections
- Simulating n-to-m semantics
 - Containers
 - Variables
 - Internal storage
 - Callable actors



Previous example

Same workflow in ADAMS

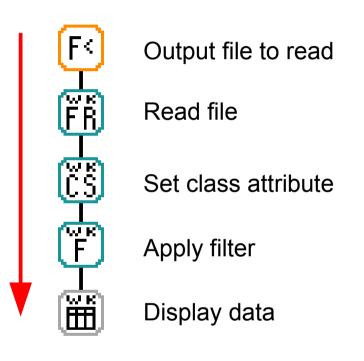








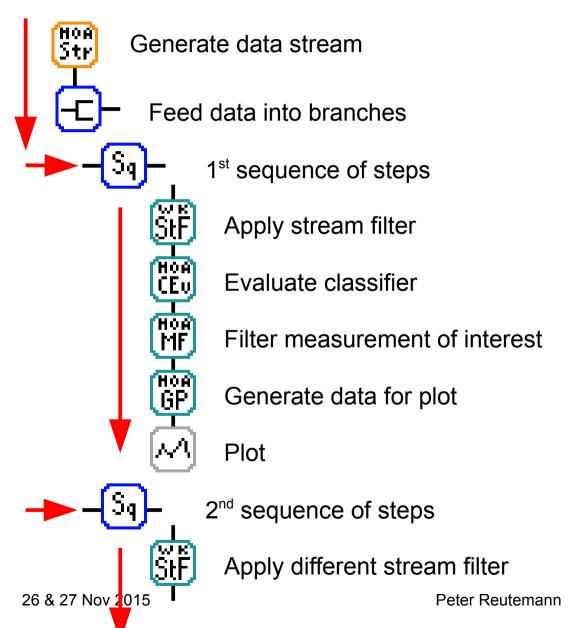
Execute nested actors one after the other



Load dataset, apply filter and display dataset







Filter data stream in two separate branches with different filters, evaluate classifier and plot metric





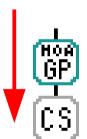


groups actors accessible via their name ("callable actors")



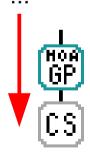
combined plot





1st evaluation: create plotting data

Pump data into referenced plot



2nd evaluation: create plotting data

Pump data into referenced plot

Generate combined plot of two evaluations by using "callable actors" functionality

Tensorflow



Feature	ADAMS	Tensorflow
Bindings	Java	Python and C++
GUI	✓	
Scripting	Groovy/Jython	Python
GPU		✓
Deep learning		✓
Non-deep learning	✓	
Multi-processor	✓	✓
Parallel execution	✓	✓
Remote execution	✓	
Data flow programming	kind of	✓
Platforms	Linux, Mac, Windows	Linux, Mac
License	GPLv3	Apache 2.0

TensorFlow Disappoints – Google Deep Learning falls shallow [source: KDNuggets]



Research (demos)

- Compare two MOA classifiers (drift)
- Compare MOA classifier on different streams
- MOA cluster visualization
- Track mouse in video

MOA - Drift

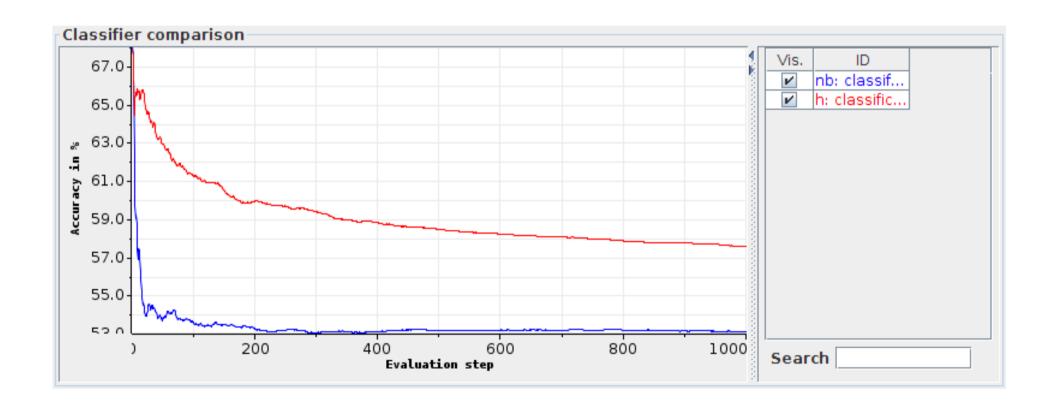


Flow

- FI Compares the accuracy of a Hoeffding tree classifier with NaiveBayes on a RandomRBFGenerator (with drift) data stream.
 - CA CallableActors
 - Hoeffding HoeffdingTree
 - NaiveBayes NaiveBayes
 - SequencePlotter x:left, Y:top, W:800, H:350
 - MOAStream RandomRBFGeneratorDrift/1000000, chunks of 1000
 - P -C-Branch parallel, threads: #cores
 - ր -[Տգ]-hoeffding
 - MOAClassifierEvaluation Hoeffding, BasicClassificationPerformanceEvaluator/1000
 - MOAMeasurementsFilter .*classifications.*
 - MOAMeasurementPlotGenerator h:
 - (S) CallableSink SequencePlotter
 - ণ জি-naivebayes
 - MOAClassifierEvaluation NaiveBayes, BasicClassificationPerformanceEvaluator/1000
 - MOAMeasurementsFilter .*classifications.*
 - MOAMeasurementPlotGenerator nb:
 - (S) CallableSink SequencePlotter

MOA - Drift







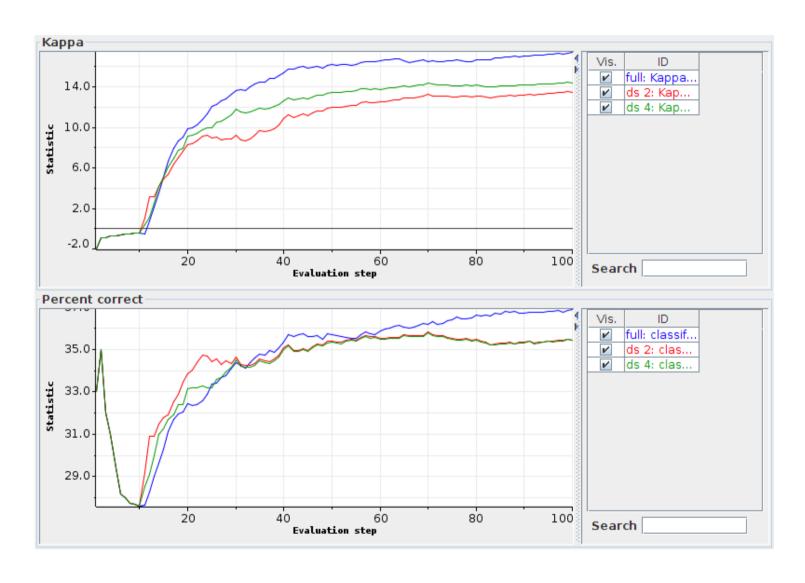
MOA - different streams

Compares the performance of a classifier on different data streams. The DownSample stream filter is used to reduce the number of attributes for two of the three streams. GridView X:left, Y:top, W:800, H:600, Rows: 2, Cols: 1 Kappa X:right, Y:top, W:800, H:350 Percent correct X:right, Y:top, W:800, H:350 CA CallableActors MOAClassifier DecisionStump MOAStream RandomRBFGenerator/10000, one-by-one P -(Sq)-full MOAClassifierEvaluation MOAClassifier, BasicClassificationPerformanceEvaluator/100 9 (-C)-Branch sequential P -{Sq}-kappa พี่ฅี MOAMeasurementsFilter карра St.* MOAMeasurementPlotGenerator full: (S) CallableSink Kappa 9 - [Sq]-percent correct MF MOAMeasurementsFilter .*correct.* MOAMeasurementPlotGenerator full: (S) CallableSink Percent correct ণ - জি-downsample 2 SF WekaStreamFilter DownSample ใช้ฟี MOAClassifierEvaluation MOAClassifier, BasicClassificationPerformanceEvaluator/100 -C-Branch sequential P - Sq-kappa

MOAMassuramentsFilter Kanna St.*



MOA - different streams





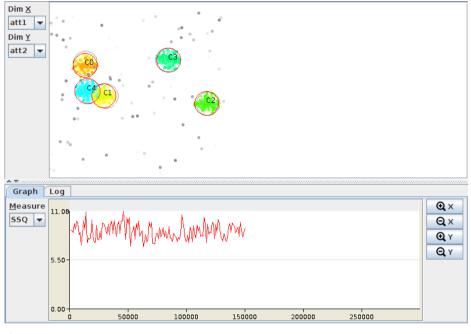
MOA - Cluster visualization

Flow

- Visualizes the clustering of the data stream. One on the raw data, the other one with filtered data.
 - CA CallableActors
 - 📆 MOAClustererSetup ClusterGenerator
 - MOAStream RandomRBFGeneratorEvents/500000, chunks of 1000
 - - ף -S₄-raw
 - MOAClusterVisualization (raw) MOAClustererSetup
 - 9 -(Sq)-filtered
 - SF WekaStreamFilter MultiFilter
 - MOAClusterVisualization (filtered) MOAClustererSetup

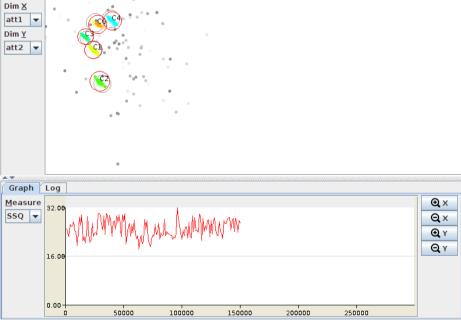


MOA - Cluster visualization



Stream 1

Stream 2







```
Flow

FI Let's the user select an initial object to be tracked.
In this case, select the head of the person.
   Frame X:left, Y:top, W:800, H:600, short title
         a background background
         Trail X:right, Y:top, W:1000, H:600, short title, paintlet: Circles, zoom: -1.0
      SetVariable @{hm_factor} = 16
      ▶ Start
   P (Tr) process video
         SelectFile directory: ${HOME}/documents/presentations/adams/bigmine-2015/flows/data
       BaseName remove extension
             $ SetVariable @{file} [REPLACE]
         MovielmageSequence 100ms
         SetStorageValue frame

    T-get timestamp

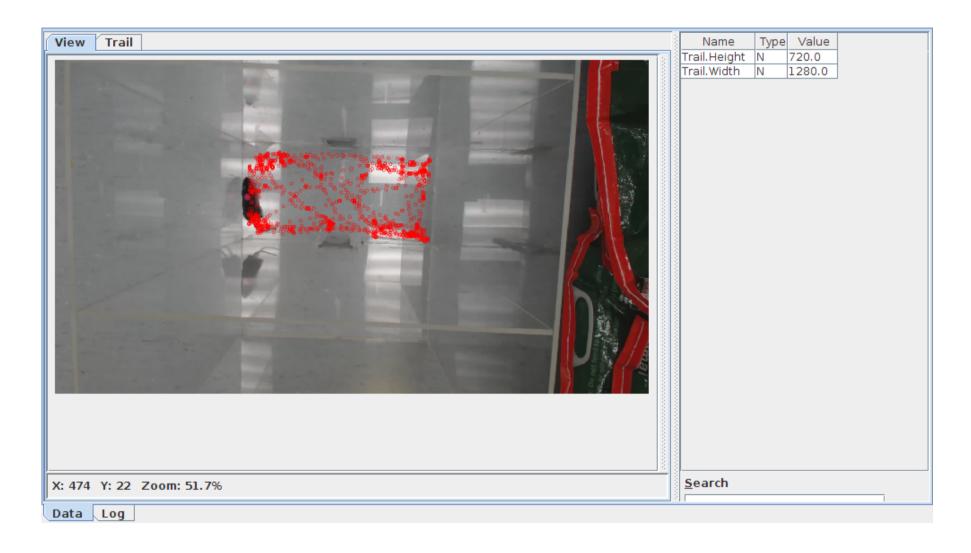
             (BI) GetReportValue Timestamp[S]
             (PB) Convert StringToDateTimeType
         P (T)-short
                PB Convert-1 DateTimeTypeToString
                $v SetVariable @{timestamp} [REPLACE]
                RB Convert-1 DateTimeTypeToString
                $ SetVariable @{timestamp_long} [REPLACE]
       9 (1x)-select object @{variable}
            [66] Inspect X:left, Y:top, W:800, H:600, provider: ImageViewer (X:left, Y:top, W:640, H:480), skip: _Skip, accept: _Accept
             SetStorageValue frame
       oreate copy for map @{variable}
             SetStorageValue map

    create copy for background @{variable}

             SetStorageValue background
       P (1x)-create heatmap/trail @{variable}
         ∘ Twidth
                ImageInfo width
                SetVariable @{width} [REPLACE]
         γ (T)-height
                ImageInfo HEIGHT
                $v SetVariable @{height} [REPLACE]
```

Track mouse





Industry





- BLGG environmental lab in NL
- Spectral analysis
 - XRF: 10,000, MIR: 2,000, NIR: 1,500
- In operation since 2006
- Predictive modelling: soil, plant (~250 models)
- 1,000 to 3,000 samples per day
- Savings due to less wet chemistry
 - USD 18 million to USD 33 million per year

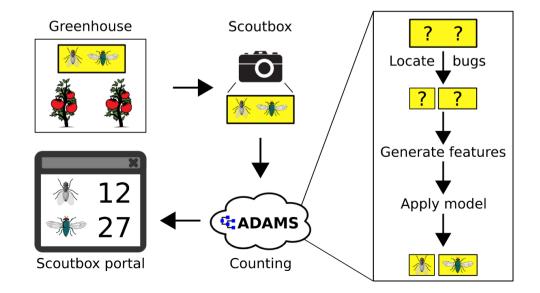
Industry (2)



Cropwatch BV



- monitoring trends in insect populations (whitefly, macrolophus, thrips)
- currently used in greenhouses
- analyzing images of sticky plates



Questions?



