An Introduction to WEKA

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Lecture at National Yang Ming University, June 2006





- What is WEKA
- Knowledge Flow
- Explorer
- Why Knowledge Flow
- Cross Validation
- Reference

What is WEKA



- Developed at Univ of Waikato in New Zealand
- A collection of state-of-art machine learning algorithms and data preprocessing tools
- Provide implementation of
 - Regression
 - Classification
 - Clustering
 - Association rules
 - Feature selection



What is WEKA

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5.0_05\lib\ext		1.gui.GUIChooser
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	Com and	
	GUI	
	Simple CLI Explorer Experimenter KnowledgeFlow	





- What is WEKA
- Knowledge Flow
- Explorer
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- Experiment 1:
 - Type: Classification
 - Feature selection: GainRatio; Ranker (top 3)
 - Algorithm: ID3
 - Training: Weather_nominal.arff
 - Test: Weather_nominal.arff



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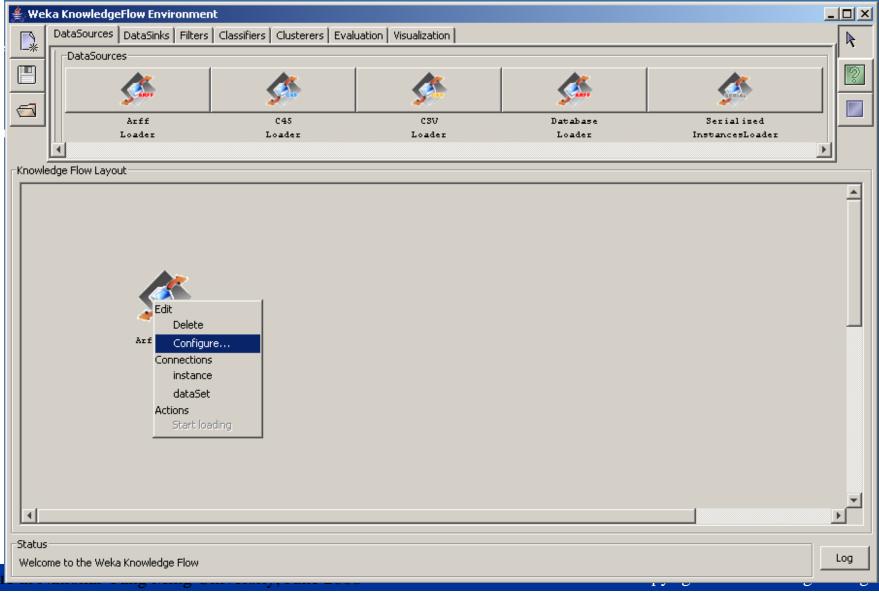
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  4 @attribute temperature {hot, mild, cool}
  5 @attribute humidity {high, normal}
  6 @attribute windy {TRUE, FALSE}
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     @attribute play {yes, no}
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  Q ...
     Adata
 10 sunny, hot, high, FALSE, no
 11 sunny, hot, high, TRUE, no
 12 overcast, hot, high, FALSE, yes
 13 rainy,mild,high,FALSE,yes
 14 rainy, cool, normal, FALSE, yes
 15 rainy,cool,normal,TRUE,no
 16 overcast, cool, normal, TRUE, yes
 17 sunny,mild,high,FALSE,no
 18 sunny,cool,normal,FALSE,ves
 19 rainy,mild,normal,FALSE,yes
 20 sunny,mild,normal,TRUE,yes
 21 overcast, mild, high, TRUE, yes
 22 overcast, hot, normal, FALSE, yes
 23 rainy,mild,high,TRUE,no
 24
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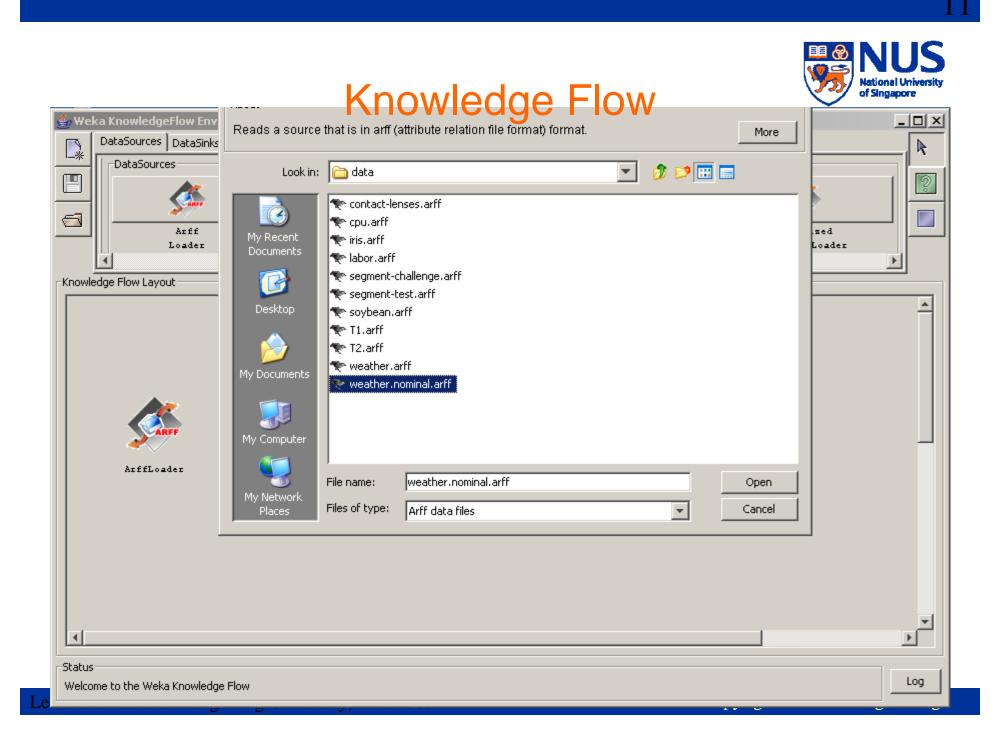


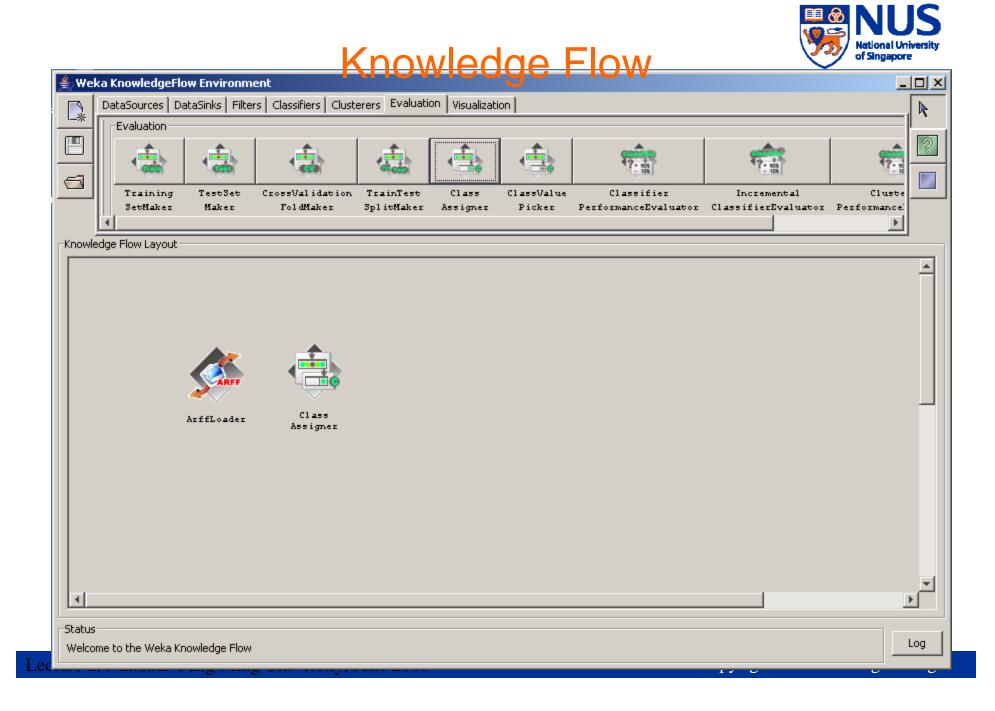
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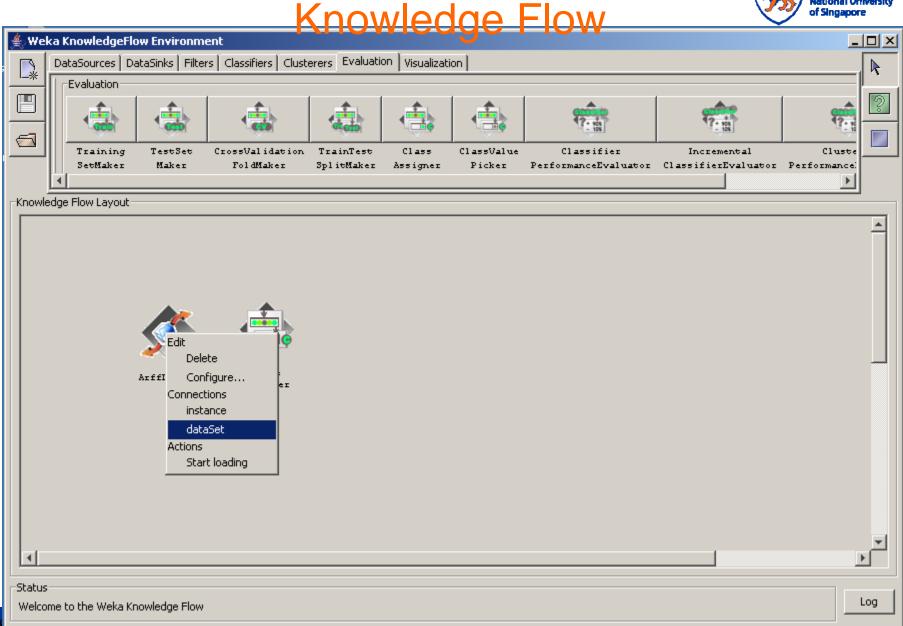






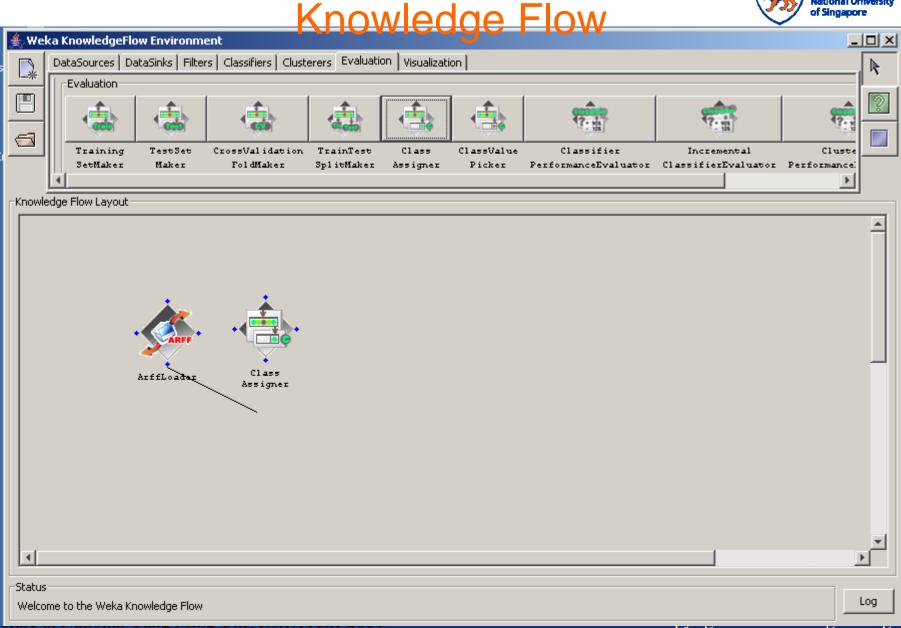


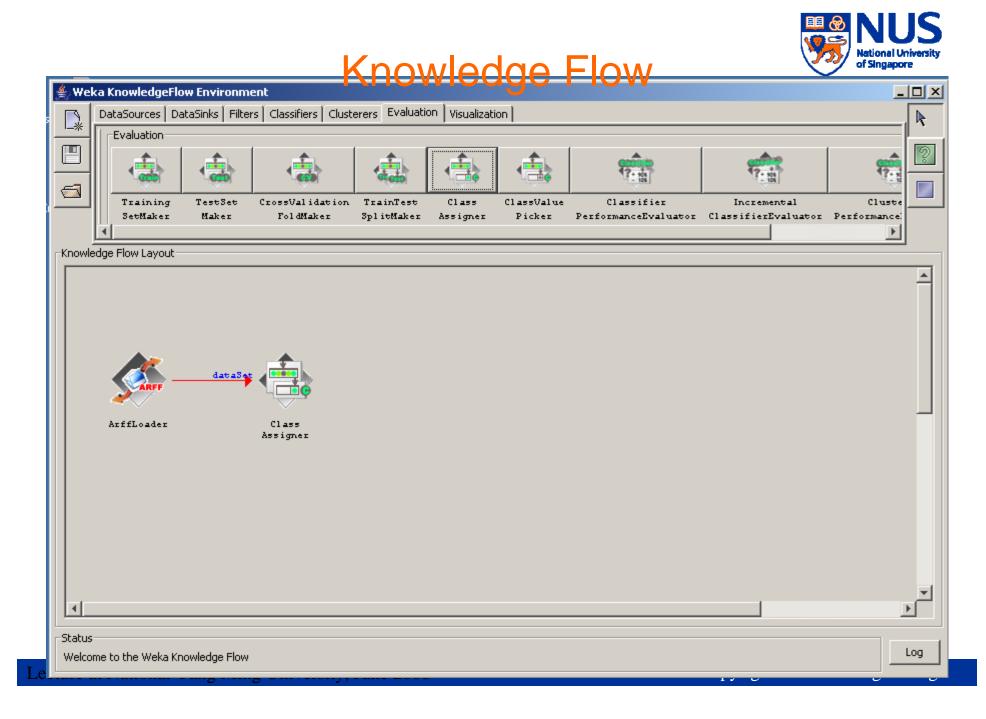


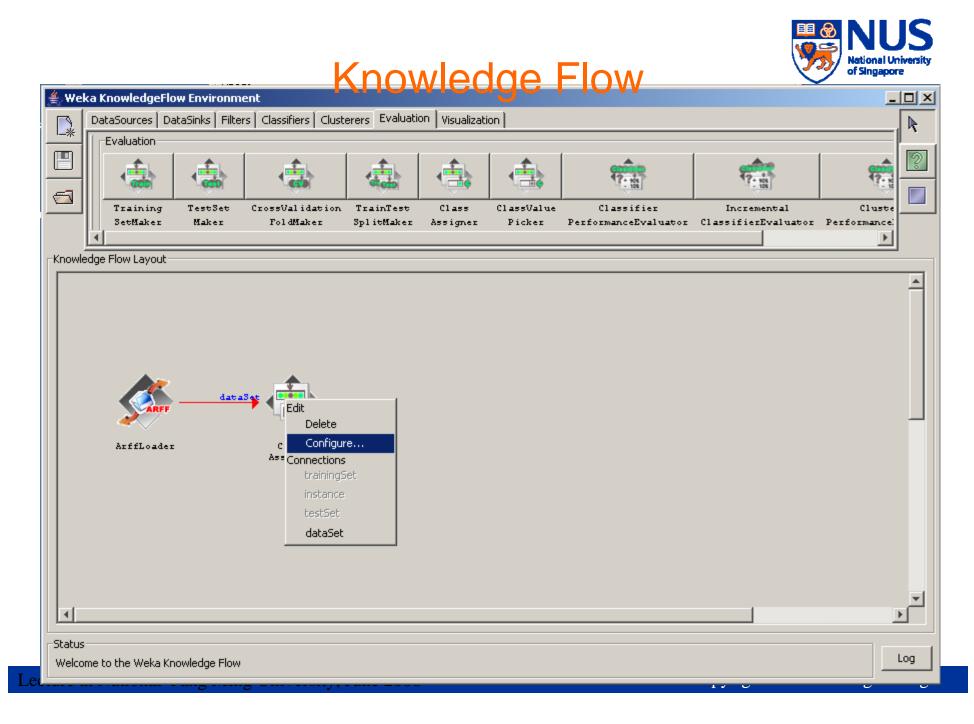


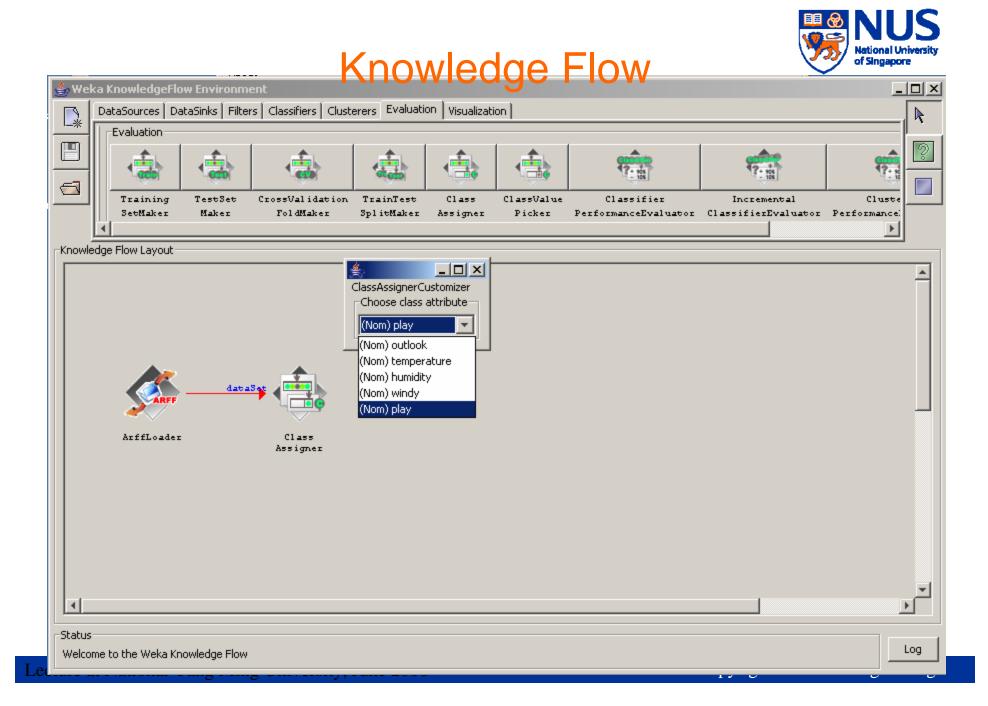
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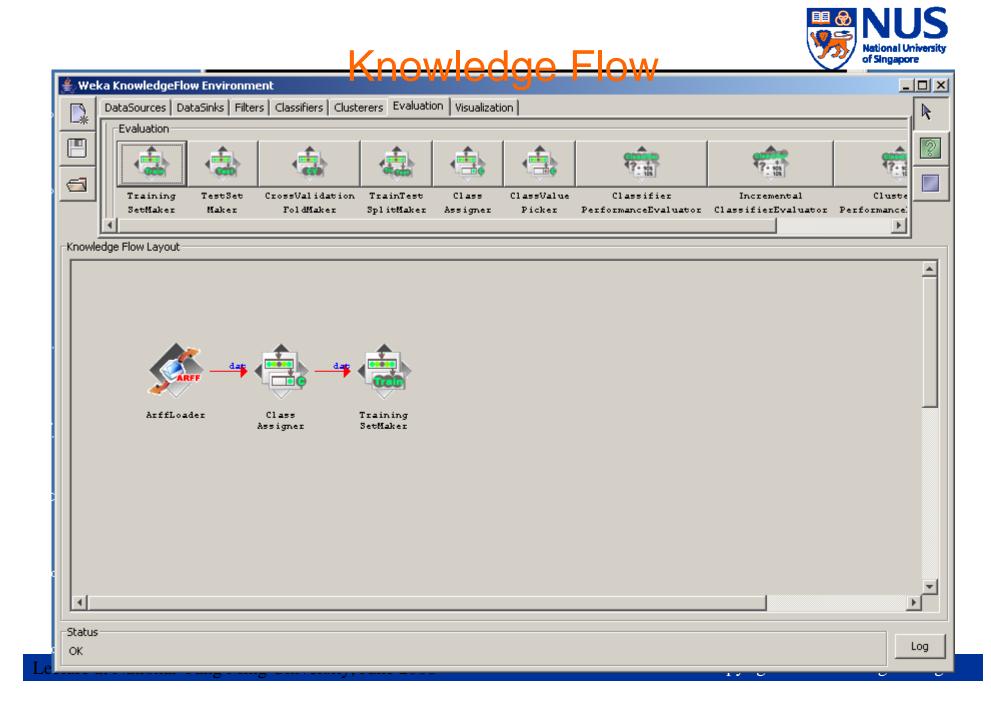




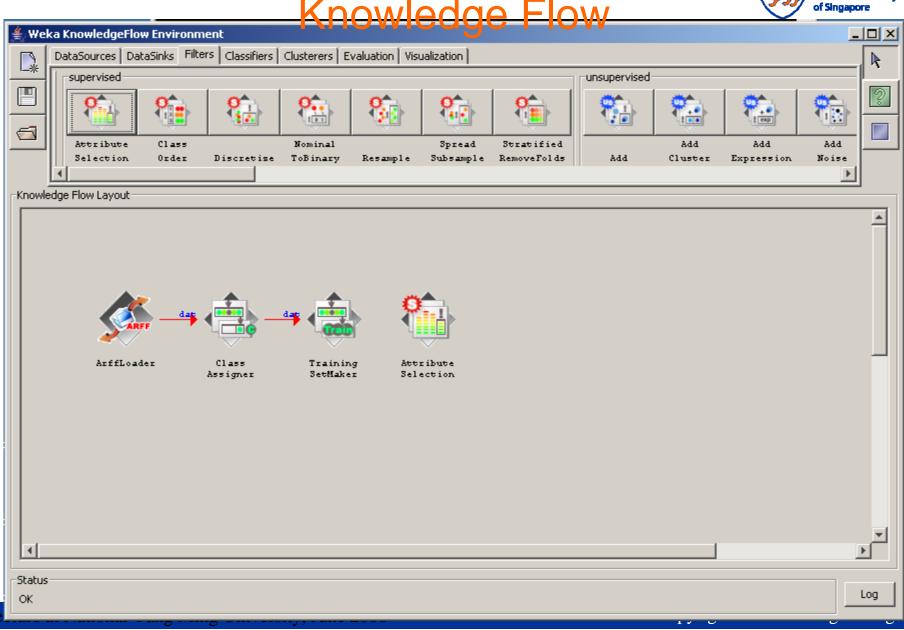


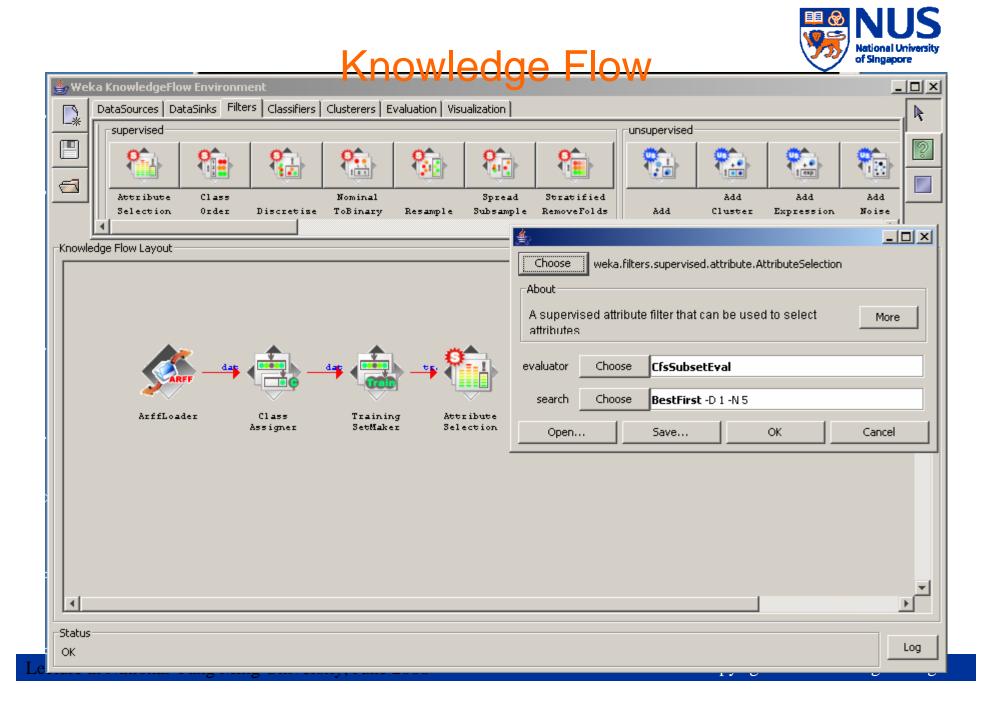


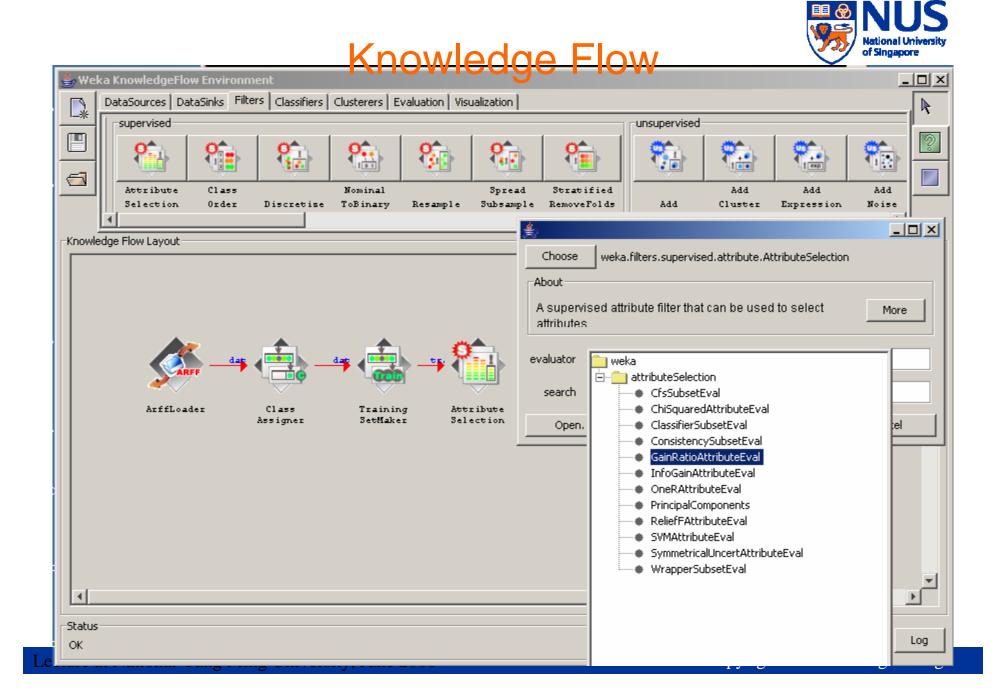


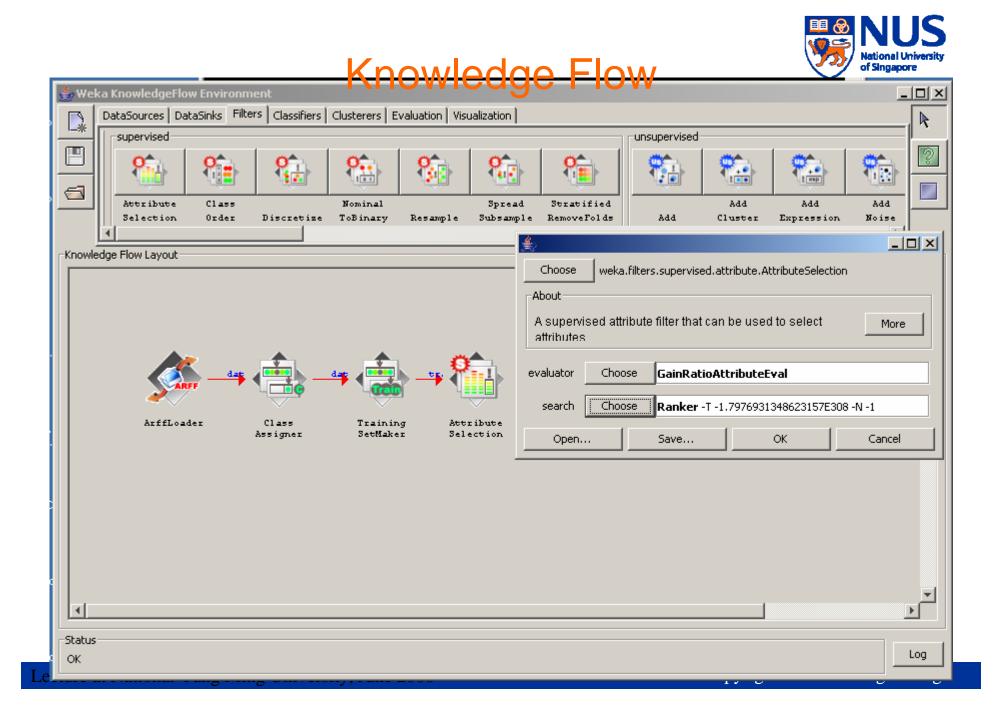


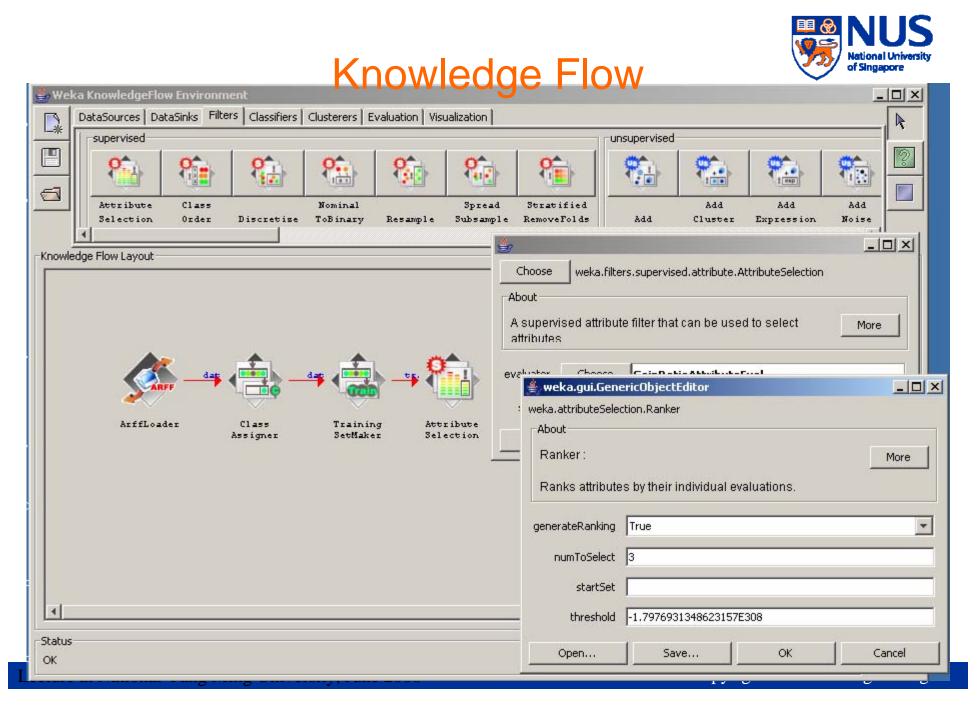




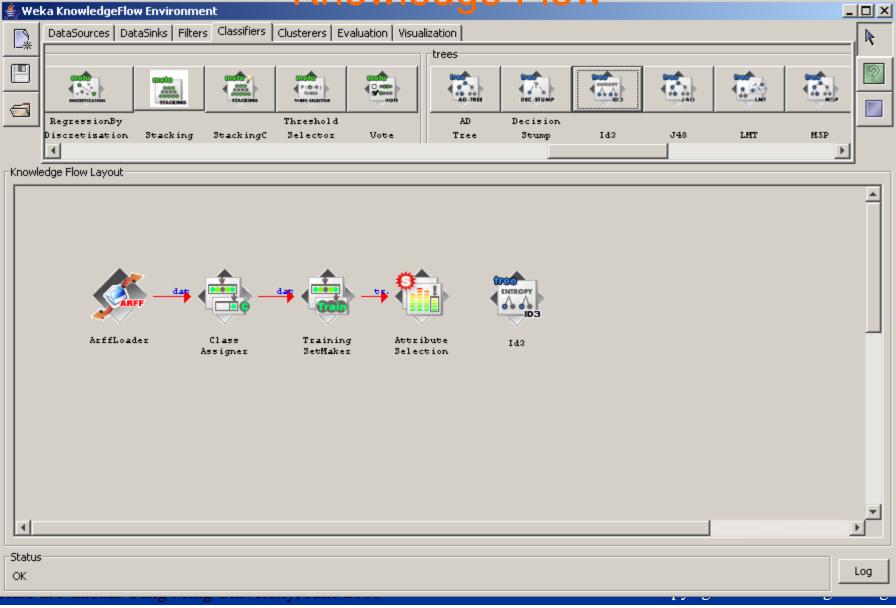


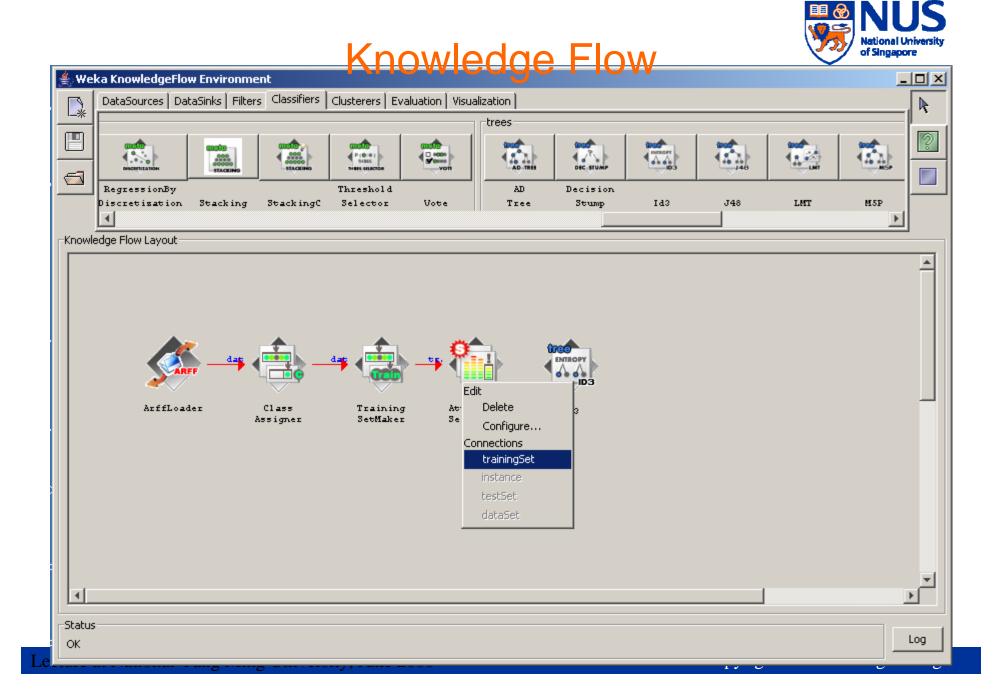


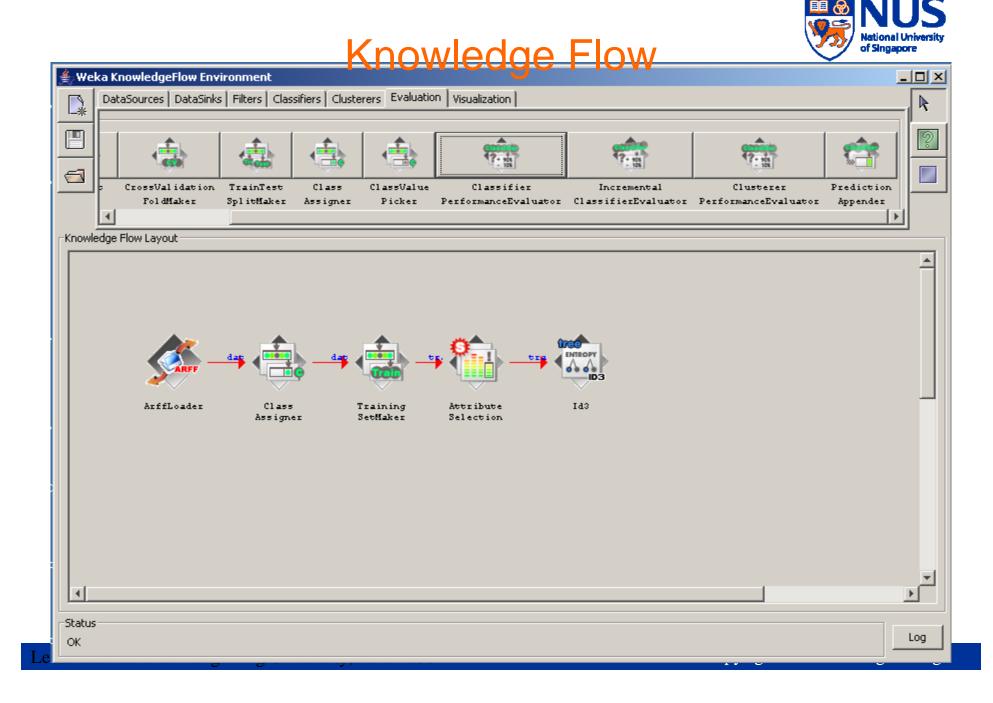


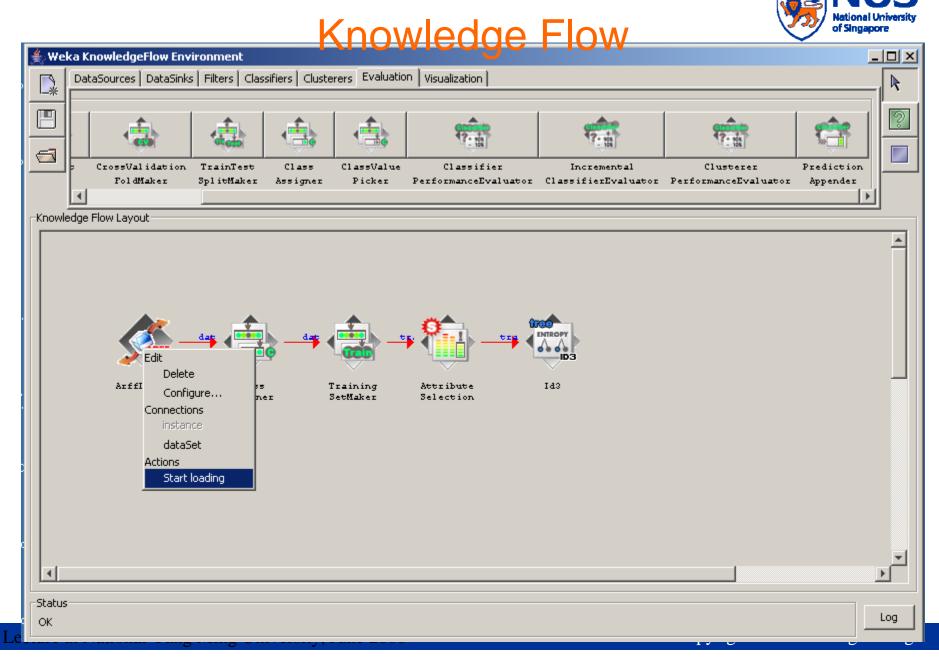


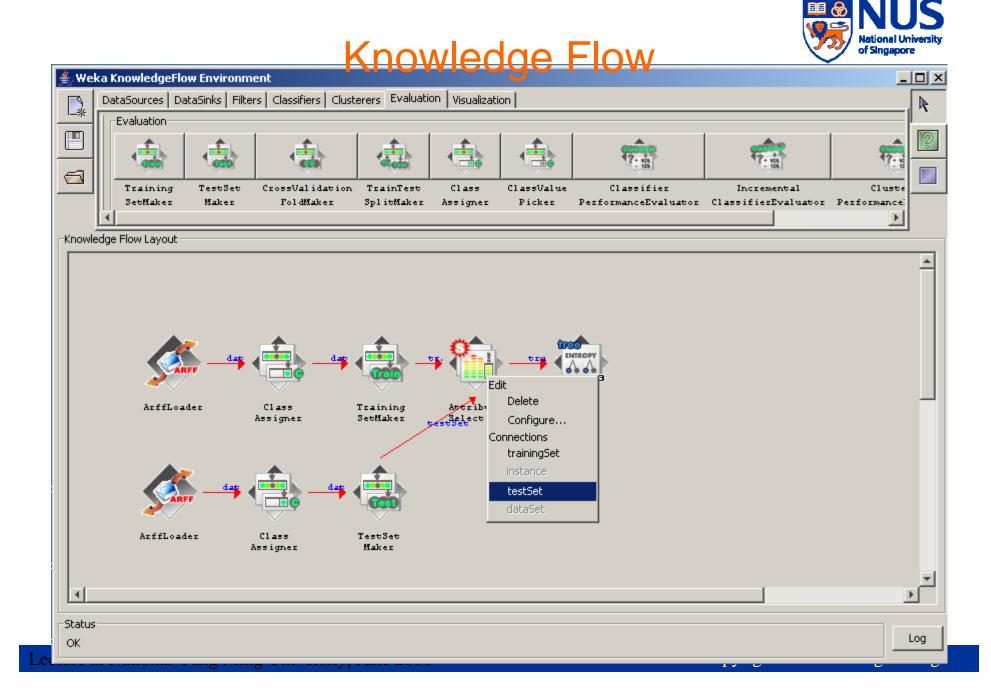


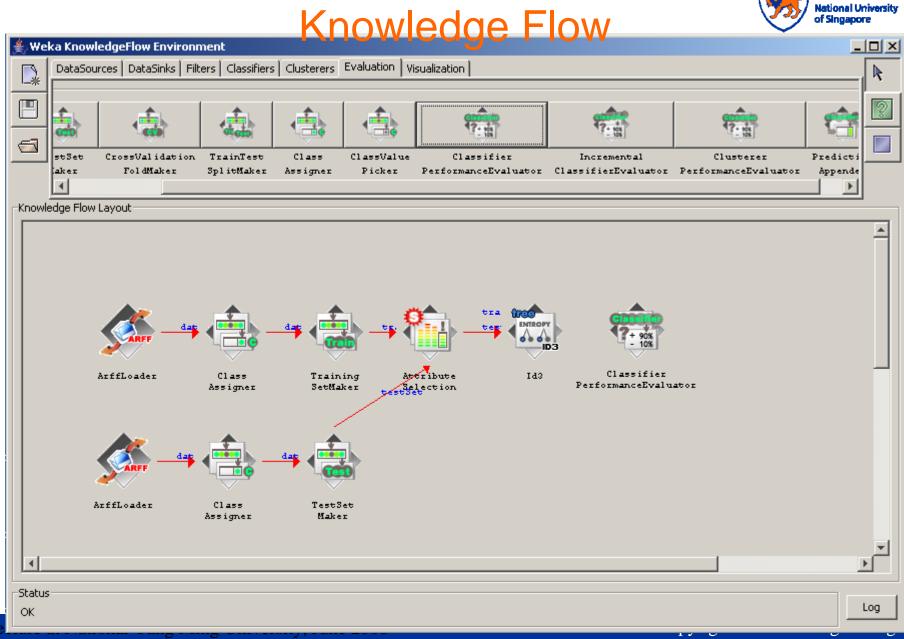




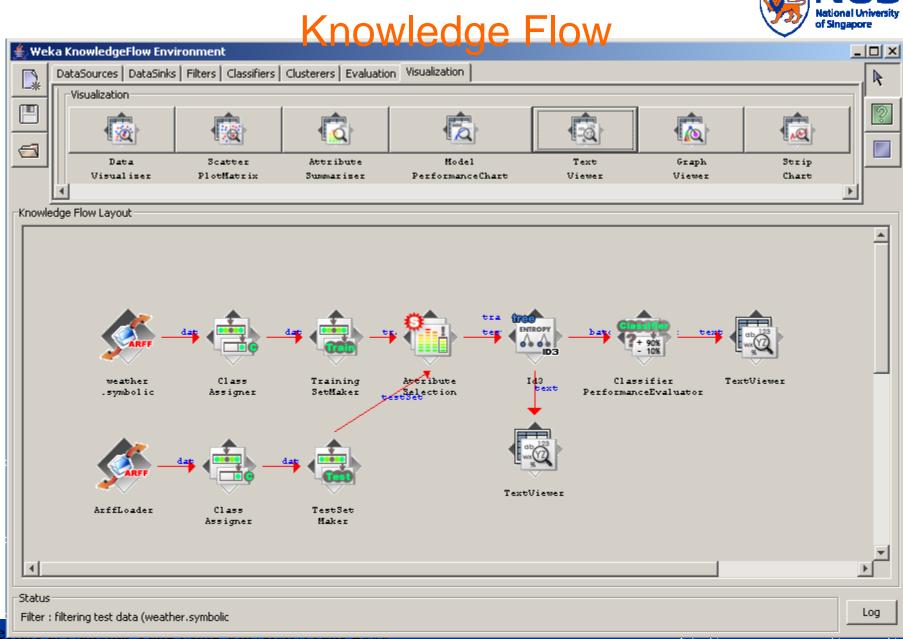












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- What is WEKA
- Knowledge Flow
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- Cross Validation
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- Do the same experiment...
- Experiment 1:
 - Type: Classification
 - Feature selection: GainRatio; Ranker top 3
 - Algorithm: ID3
 - Training: Weather_nominal.arff
 - Test: Weather_nominal.arff





🚖 Weka Explorer					
Preprocess Classify Clus	ster 🛛 Associate 🗍 Select att	ributes Visualize			
Open file	Open URL	Open DB	Undo	Edit	Save
Filter					
Choose None					Apply
Current relation			Selected attribute		
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Filter				
Choose None				Apply
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3 humidity		/070104002013/L000 4V 0		
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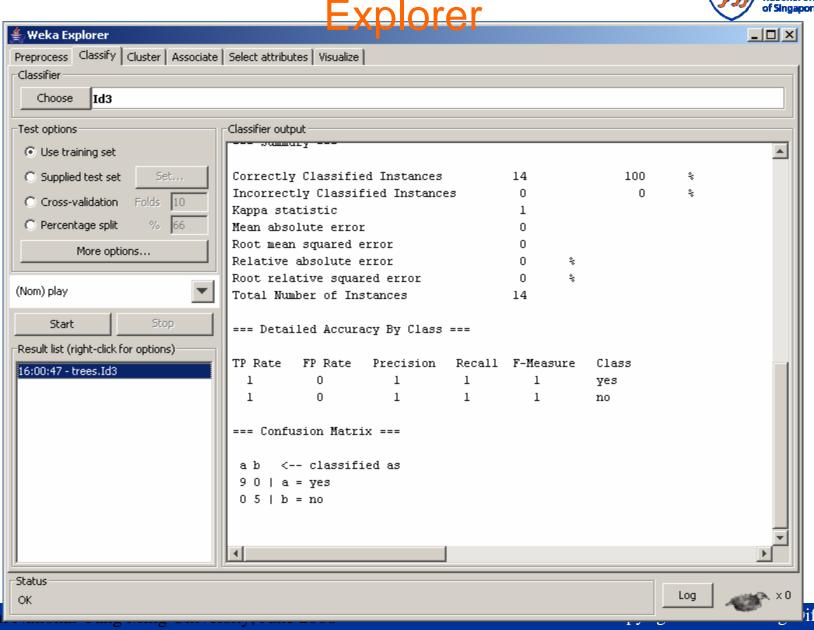
40



🚖 Weka Explorer		
Preprocess Classify Cluster Associate Select attrib	butes Visualize	
Classifier		
Choose Id3		
Test options	Classifier output	
 Use training set 		
O Supplied test set Set		
C Cross-validation Folds 10		
C Percentage split % 66		
More options		
(Nom) play		
Start Stop		
Result list (right-click for options)		
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Status OK	Log	
OK		

Lecture









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Why Knowledge Flow

- There are some jobs we cannot do in explorer ...
 - Combine feature selection
 - Build more complicated systems
- KF describes the process more clearly
 - Never regard the training and test data to be separate in the previous example in explorer
- KF help us to access some mid-process info of the machine learning method
 - Cross Validation





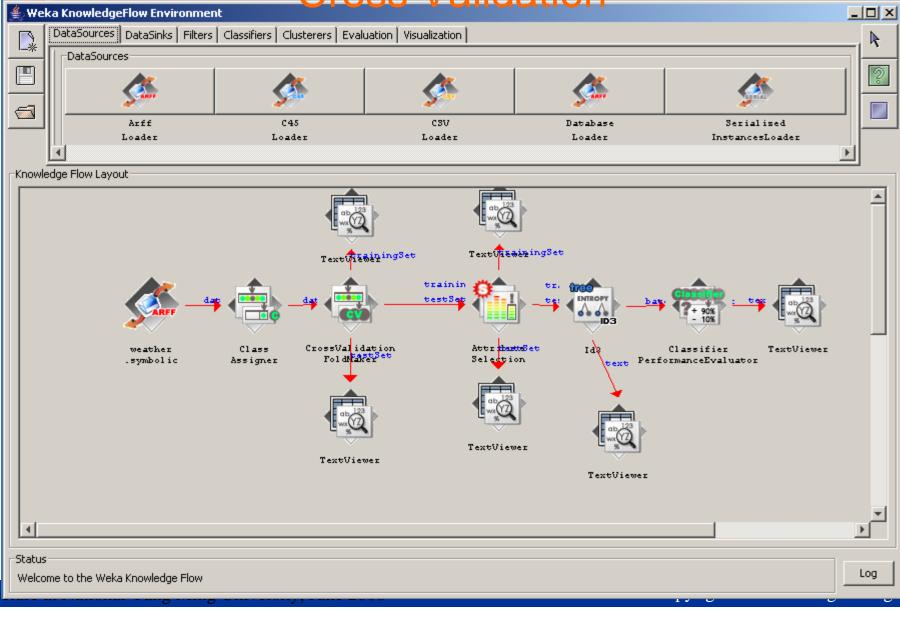
- What is WEKA
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- Experiment 2:
 - Type: Classification
 - Feature selection: GainRatio; Ranker top 3
 - Algorithm: ID3
 - Training: Weather_nominal.arff (CV)
 - Test: Weather_nominal.arff (CV)
 - CV type: 3-folder CV



Cross Validation





What do we view in this case? Text1 VS. Text2 (1)

👙 Text Viewer		擒 Text Viewer		
Fext Viewer Result list 16:39:31 - weather.symbolic 16:39:35 - weather.symbolic 16:39:35 - weather.symbolic2 16:39:35 - weather.symbolic3	Text @relation weather.symbolic @attribute outlook {sunny,overc @attribute temperature {hot,mil @attribute humidity {high,norma @attribute windy {TRUE,FALSE} @attribute play {yes,no}	Event Viewer Result list 16:39:31 - weather.symbolic 16:39:35 - weather.symbolic 16:39:35 - weather.symbolic2 16:39:35 - weather.symbolic3	Text @relation weather.symbolic @attribute outlook {sunny,overcast,rainy} @attribute temperature {hot,mild,cool} @attribute humidity {high,normal} @attribute windy {TRUE,FALSE} @attribute play {yes,no}	
	<pre>@data sunny,hot,high,FALSE,no overcast,hot,normal,FALSE,yes overcast,mild,high,TRUE,yes sunny,cool,normal,FALSE,yes sunny,hot,high,TRUE,no sunny,mild,normal,TRUE,yes overcast,hot,high,FALSE,yes rainy,mild,normal,FALSE,yes rainy,mild,high,TRUE,no</pre>		<pre>@data rainy,cool,normal,TRUE,no sunny,mild,high,FALSE,no overcast,cool,normal,TRUE,yes rainy,cool,normal,FALSE,yes rainy,mild,high,FALSE,yes</pre>	



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Text1 VS. Text2 (2)

🖆 Text Viewer		套 Text Viewer	
Result list 16:39:31 - weather.symbolic 16:39:35 - weather.symbolic 16:39:35 - weather.symbolic3 16:39:35 - weather.symbolic3	Text @relation weather.symbolic @attribute outlook (sunny,overc @attribute temperature (hot,mil @attribute humidity (high,norma @attribute windy (TRUE,FALSE) @attribute play (yes,no) @data sunny,mild,high,FALSE,no rainy,cool,normal,TRUE,no sunny,mild,normal,TRUE,yes sunny,hot,high,TRUE,no rainy,mild,high,FALSE,yes rainy,mild,normal,FALSE,yes rainy,cool,normal,FALSE,yes overcast,cool,normal,FALSE,yes	Result list 16:39:31 - weather.symbolic 16:39:35 - weather.symbolic 16:39:35 - weather.symbolic2 16:39:35 - weather.symbolic3	Text @relation weather.symbolic @attribute outlook {sunny,overcast,rainy} @attribute temperature {hot,mild,cool} @attribute humidity {high,normal} @attribute windy {TRUE,FALSE} @attribute play {yes,no} @data rainy,mild,high,TRUE,no sunny,hot,high,FALSE,no overcast,hot,high,FALSE,yes sunny,cool,normal,FALSE,yes overcast,mild,high,TRUE,yes



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Text1 VS. Text2 (3)

🚖 Text Viewer		🛓 Text Viewer	
Result list 16:39:31 - weather.symbolic 16:39:35 - weather.symbolic 16:39:35 - weather.symbolic2 16:39:35 - weather.symbolic3	Text @relation weather.symbolic @attribute outlook {sunny,overc @attribute temperature {hot,mil @attribute humidity {high,norma @attribute windy {TRUE,FALSE} @attribute play {yes,no}	Text Viewer Result list 16:39:31 - weather.symbolic 16:39:35 - weather.symbolic 16:39:35 - weather.symbolic2 16:39:35 - weather.symbolic3	Text @relation weather.symbolic @attribute outlook {sunny,overcast,rainy} @attribute temperature {hot,mild,cool} @attribute humidity {high,normal} @attribute windy {TRUE,FALSE} @attribute play {yes,no}
	<pre>@data sunny,hot,high,FALSE,no sunny,mild,high,FALSE,no rainy,cool,normal,TRUE,no rainy,mild,high,TRUE,no rainy,cool,normal,FALSE,yes sunny,cool,normal,FALSE,yes overcast,mild,high,TRUE,yes rainy,mild,high,FALSE,yes overcast,hot,high,FALSE,yes overcast,cool,normal,TRUE,yes</pre>		<pre>@data sunny,hot,high,TRUE,no sunny,mild,normal,TRUE,yes rainy,mild,normal,FALSE,yes overcast,hot,normal,FALSE,yes</pre>



Text3 VS. Text4 (1)

🖕 Text Viewer		套 Text Viewer	
Result list	Text	Result list	Text
16:39:31 - weather.symbolic 16:39:35 - weather.symbolic-we 16:39:35 - weather.symbolic-we2 16:39:35 - weather.symbolic-we3 16:51:27 - weather.symbolic-we 16:51:27 - weather.symbolic-we2 16:51:28 - weather.symbolic-we	<pre>@relation 'weather.symbolic-weka.filters. @attribute humidity {high,normal} @attribute outlook {sunny,overcast,rainy} @attribute temperature {hot,mild,cool} @attribute play {yes,no}</pre>	16:39:31 - weather.symbolic 16:39:35 - weather.symbolic-we 16:39:35 - weather.symbolic-we2 16:39:35 - weather.symbolic-we3 16:51:27 - weather.symbolic-we 16:51:27 - weather.symbolic-we2 16:51:28 - weather.symbolic-we	<pre>@relation 'weather.symbolic-weka.filte @attribute humidity {high,normal} @attribute outlook {sunny,overcast,rai @attribute temperature {hot,mild,cool} @attribute play {yes,no}</pre>
	<pre>@data high, sunny, hot, no normal, overcast, hot, yes high, overcast, mild, yes normal, sunny, cool, yes high, sunny, hot, no normal, sunny, mild, yes high, overcast, hot, yes normal, rainy, mild, yes high, rainy, mild, no</pre>		<pre>@data normal,rainy,cool,no high,sunny,mild,no normal,overcast,cool,yes normal,rainy,cool,yes high,rainy,mild,yes</pre>



Text3 VS. Text4 (2)

16:39:31 - weather.symbolic	ext	Describelish	
16:39:35 - weather.symbolic-we3 0 16:51:27 - weather.symbolic-we2 0 16:51:27 - weather.symbolic-we2 0 16:51:28 - weather.symbolic-we 0 0 0	Grelation 'weather.symbolic-weka.filters. Gattribute humidity (high,normal) Gattribute outlook (sunny,overcast,rainy) Gattribute windy (TRUE,FALSE) Gattribute play (yes,no)	Result list 16:39:31 - weather.symbolic 16:39:35 - weather.symbolic-we 16:39:35 - weather.symbolic-we2 16:39:35 - weather.symbolic-we3 16:51:27 - weather.symbolic-we 16:51:28 - weather.symbolic-we	Text @relation 'weather.symbolic-weka.filte @attribute humidity {high,normal} @attribute outlook {sunny,overcast,rai: @attribute windy {TRUE,FALSE} @attribute play {yes,no} @data bigh mains TDUE and
h n h h h n n n n	high, sunny, FALSE, no hormal, rainy, TRUE, no hormal, sunny, TRUE, yes high, sunny, TRUE, no high, rainy, FALSE, yes hormal, rainy, FALSE, yes hormal, overcast, TRUE, yes hormal, overcast, FALSE, yes		<pre>@data high,rainy,TRUE,no high,sunny,FALSE,no high,overcast,FALSE,yes normal,sunny,FALSE,yes high,overcast,TRUE,yes</pre>



Text3 VS. Text4 (3)

👙 Text Viewer		粪 Text Viewer	
Result list	Text	Result list	Text
16:39:31 - weather.symbolic 16:39:35 - weather.symbolic-we 16:39:35 - weather.symbolic-we2 16:39:35 - weather.symbolic-we3 16:51:27 - weather.symbolic-we 16:51:27 - weather.symbolic-we2 16:51:28 - weather.symbolic-we	<pre>@relation 'weather.symbolic-weka.filters. @attribute outlook {sunny,overcast,rainy} @attribute humidity {high,normal} @attribute temperature {hot,mild,cool} @attribute play {yes,no}</pre>	16:39:31 - weather.symbolic 16:39:35 - weather.symbolic-we 16:39:35 - weather.symbolic-we2 16:39:35 - weather.symbolic-we3 16:51:27 - weather.symbolic-we 16:51:27 - weather.symbolic-we2 16:51:28 - weather.symbolic-we	<pre>@relation 'weather.symbolic-weka.filte @attribute outlook {sunny,overcast,rai @attribute humidity {high,normal} @attribute temperature {hot,mild,cool} @attribute play {yes,no}</pre>
	<pre>@data sunny,high,hot,no sunny,high,mild,no rainy,normal,cool,no rainy,normal,cool,yes sunny,normal,cool,yes overcast,high,mild,yes rainy,high,mild,yes overcast,high,hot,yes overcast,normal,cool,yes</pre>		<pre>@data sunny,high,hot,no sunny,normal,mild,yes rainy,normal,mild,yes overcast,normal,hot,yes</pre>



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Cross Validation

Trees

🚔 16:51:27 - Model: Id3	👙 16:51:27 - Model: Id32	≜ 16:51:28 - Model: Id3
=== Classifier model ===	=== Classifier model ===	=== Classifier model ===
Scheme: Id3 Relation: weather.symbolic- Training Fold: 1	Scheme: Id3 Relation: weather.symbolic- Training Fold: 2	Scheme: Id3 Relation: weather.symbolic-weka.filters.supervised.attribute.A Training Fold: 3
Id3	Id3	Id3
humidity = high outlook = sunny: no outlook = overcast: yes outlook = rainy: no humidity = normal: yes	outlook = sunny humidity = high: no humidity = normal: yes outlook = overcast: yes outlook = rainy windy = TRUE: no windy = FALSE: yes	outlook = sunny humidity = high: no humidity = normal: yes outlook = overcast: yes outlook = rainy: yes



Cross Validation Evaluation of result

攁 Text Viewer							
Result list	_Text						
16:39:35 - Id3	Incorrect	tly Classif	fied Instanc	es	2	14.2857 %	
16:51:28 - Id3	Kappa statistic			0.6889			
	Mean abso	olute errom	<u>-</u>		0.1786		
	Root mean	n squared e	error		0.4009		
	Relative absolute error				37.234 %		
	Root rela	ative squar	ced error		82.7516 %		
	Total Num	uber of Ins	stances		14		
	=== Detai	iled Accura	acy By Class	===			
	TP Rate	FP Rate	Precision	Recall	F-Measure	Class	
	0.889	0.2	0.889	0.889	0.889	yes	
	0.8	0.111	0.8	0.8	0.8	no	
	=== Confu	usion Matri	ix ===				
	ab <-	classifi	ied as				
	81 a	= ves					
	14 b	_					
	1 1 1 2	110					_
	<u> </u>						



- Conclusion:
 - Source data are separated into several folders for cross validation
 - Feature selection is done for each training folder (only training) folder separately
 - Different trees are build in different cases
 - The evaluation of classification is by overall results



- Experiment 3:
 - Type: Classification
 - Feature selection: GainRatio; Ranker top 2
 - Algorithm: ID3
 - Training: Weather_nominal.arff
 - Test: Weather_nominal.arff



Ranker top 3 VS. Ranker top 2

👙 16:51:27 - Model: Id32	🚔 17:08:31 - Model: Id32
=== Classifier model ===	=== Classifier model ===
	Coheren T10
Scheme: Id3	Scheme: Id3
Relation: weather.symbolic-weka.filt	Relation: weather.symbolic-weka.filters.supervised.attribute.A
Training Fold: 2	Training Fold: 2
Id3	Id3
outlook = sunny	outlook = sunny
humidity = high: no	humidity = high: no
humidity = normal: yes	humidity = normal: yes
outlook = overcast: yes	outlook = overcast: yes
outlook = rainy	outlook = rainy
windy = TRUE: no	humidity = high: yes
windy = FALSE: yes	humidity = normal: yes





- Conclusion:
 - Attribute "windy" was ignored. In this case, the classifier only consider the attribute that was kept



Reference

- <u>http://www.cs.waikato.ac.nz/~ml/</u>
- Ian H. Witten, Eibe Frank. *Data Mining: Practical Machine Learning Tools and Techniques* (Second Edition)