



Creating a Digital Twin of an Insider Threat Detection Enterprise using Model Based Systems Engineering

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Agenda

Motivation

Benefits of Using a Digital Twin (DT)

Insider Threat Detection Evaluation Challenges

Inference Enterprise Modeling as a DT Technique

Solution Method – Model Based Digital Twin Generation

Discussion



Motivation

Constant threats -> need for continuous monitoring

Operational testing is a challenge

- Lack of ground truth
- Data Privacy
- Blackbox models (COTS)
- Sensitive to disruptions

Solution: Use digital twin techniques to create real-time virtual counterpart of the physical system

Digital Twin Benefits

Provides virtual and dynamic digital representation of the system

Enables real-time monitoring of the system

Reduces cost associated with testing and verification

Integration with MBSE reduces ambiguity through unifying system structures and behaviors

Insider Threat

Malicious threat that comes from people within the organization

Examples:

- 2019 Capital One: \$150M
 - Data breach via former employee of vendor
- 2018 Google and Uber: \$245M
 - IP theft via former employee

2016 DoD 5220.22-M:

- Insider Threat Program Mandate



Limitations with Insider Threat Detection

Limited understanding of detection accuracy / performance

Confusion Matrix

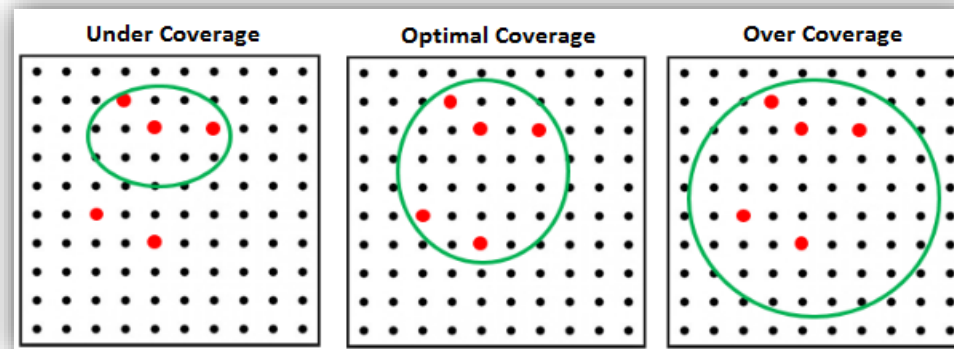
	Positive	negative
Predicted positive	TP	FP
Predicted negative	FN	TN

Recall = $\frac{TP}{TP+FN}$

Precision = $\frac{TP}{TP+FP}$

False Positive Rate = $\frac{FP}{FP+TN}$

Overwhelming number of false alarms / limited ability to reduce the population to manageable subpopulations for more detailed analysis



Inference Enterprise (IE): enterprises that make inferences based on incomplete information (lack of ground truth)

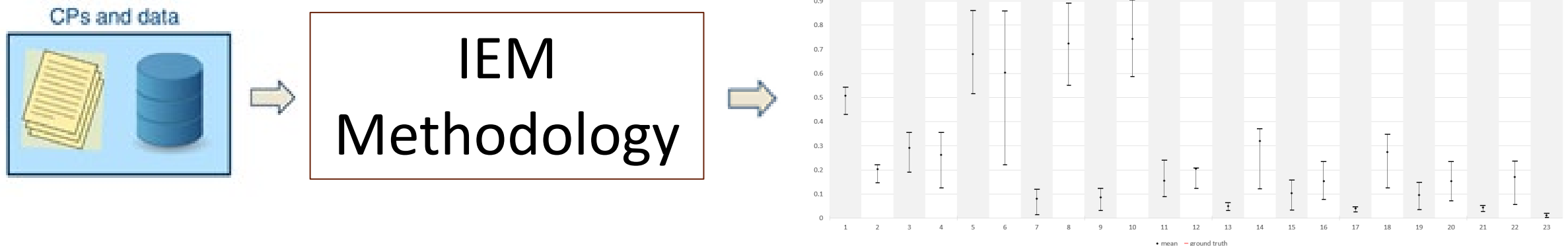
Methodology Development

19 Challenge Problems: description of Inference Enterprise (IE) and data

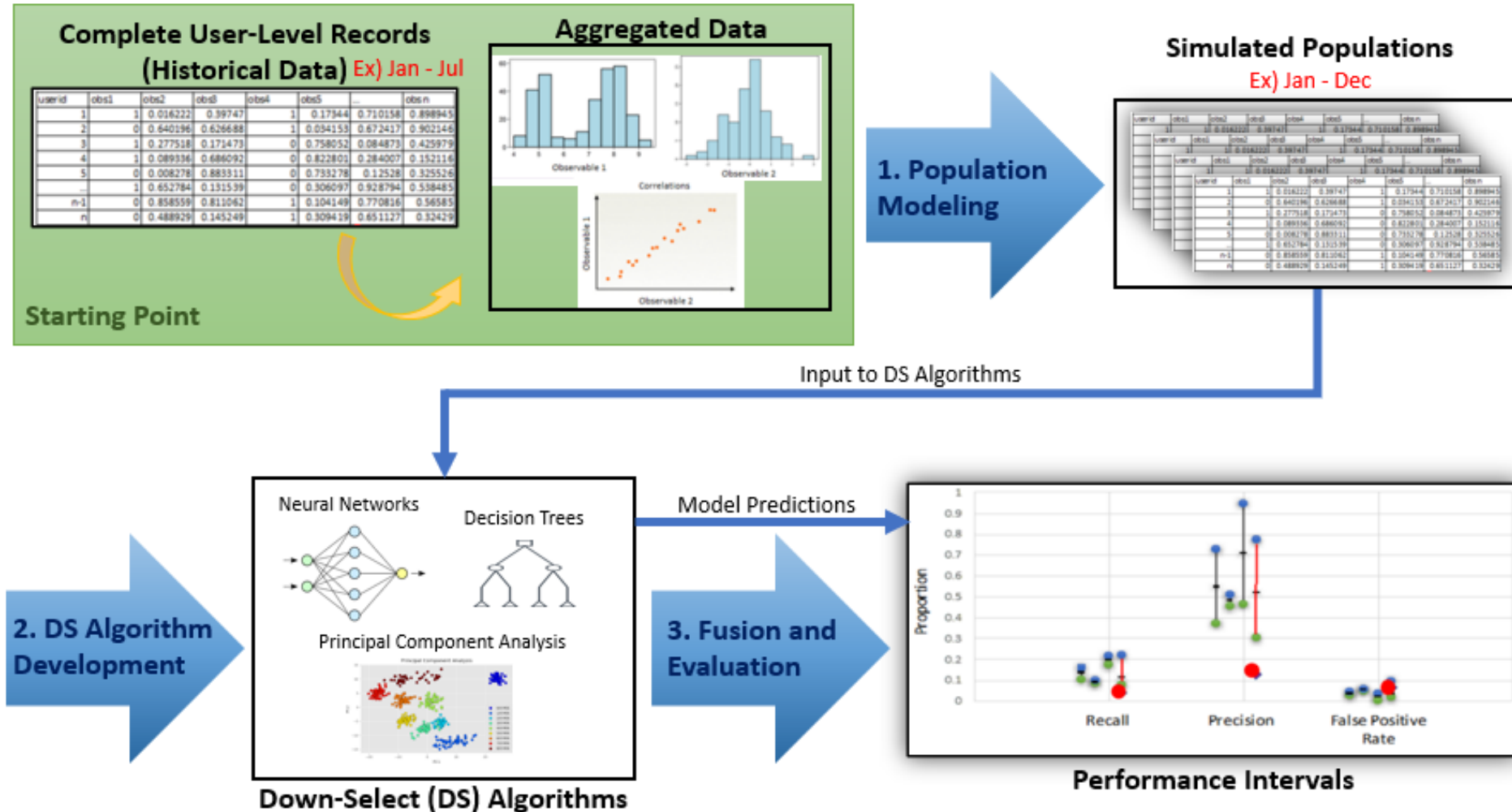
- Range of threat behavior: accidental compromise to deliberate sabotage
- Missing and/or incomplete data
- Down-select algorithms used: classification / clustering algorithms
- Forecast future performance, evaluate hypothetical upgrades to system

Inference Enterprise Modeling (IEM) methodology

- Build models and evaluate performance



Inference Enterprise Modeling (IEM)

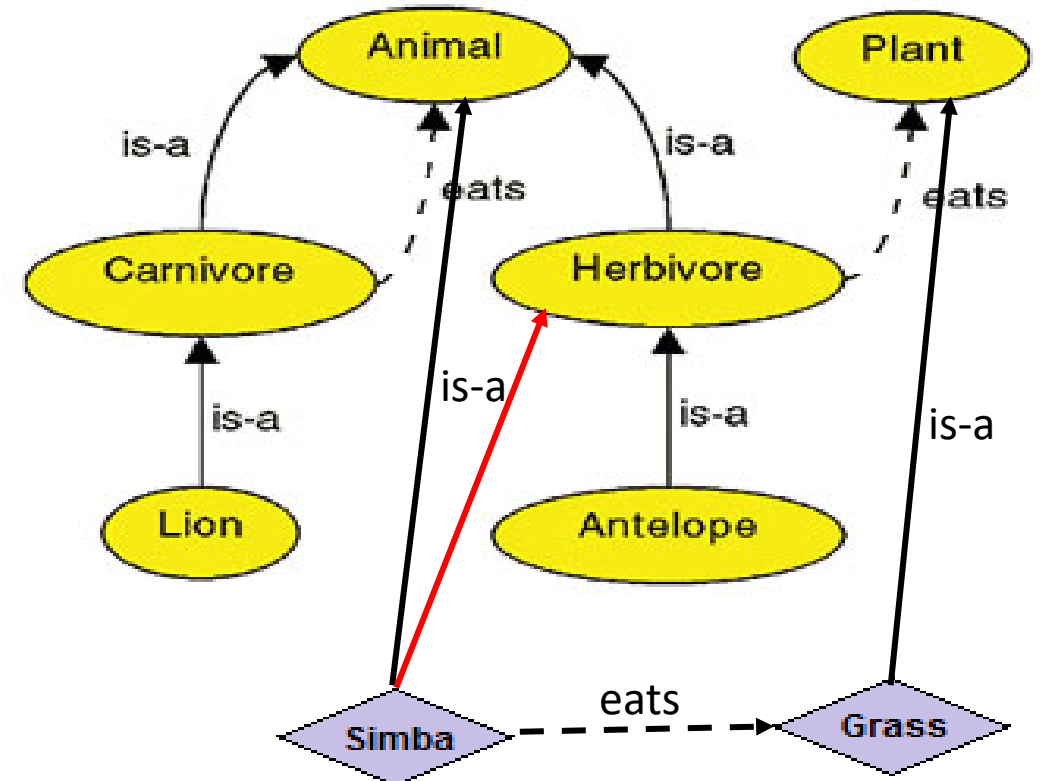


Relevant Methods - Ontology

Formal description of knowledge as a set of concepts within a domain and the relationships between those concepts

Benefits:

- Provides coherent and easy navigation between one concept to another - classes and instances
- By having essential relationships between concepts, enables automated reasoning
- Easy to connect with other ontologies
- Can represent any data format, enabling smoother data integration

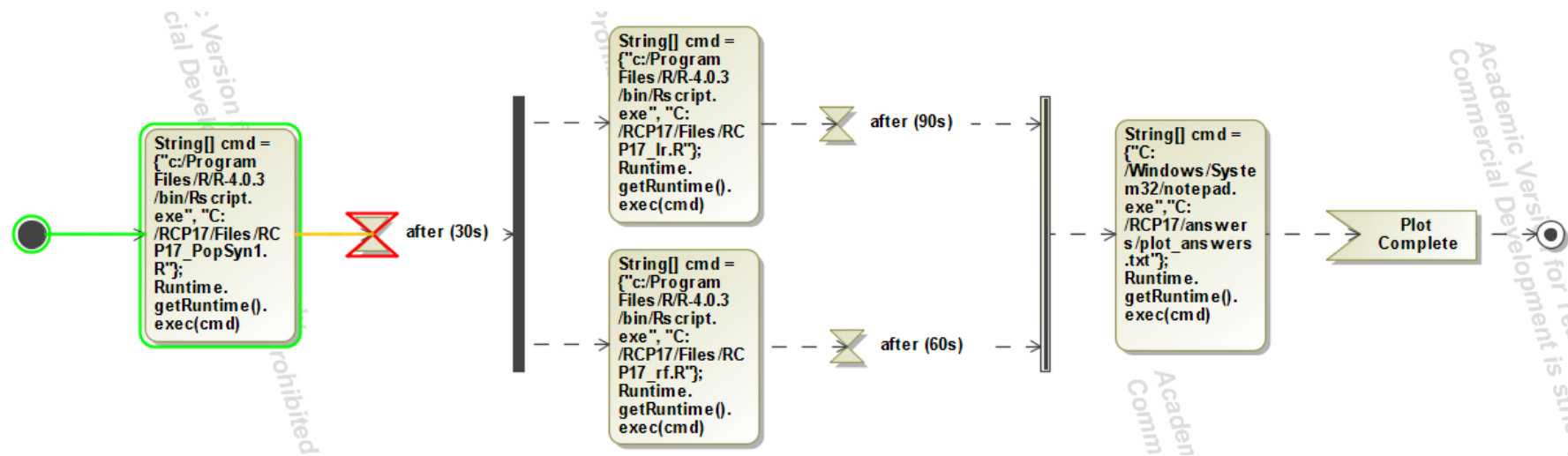


Relevant Methods – Process Modeling

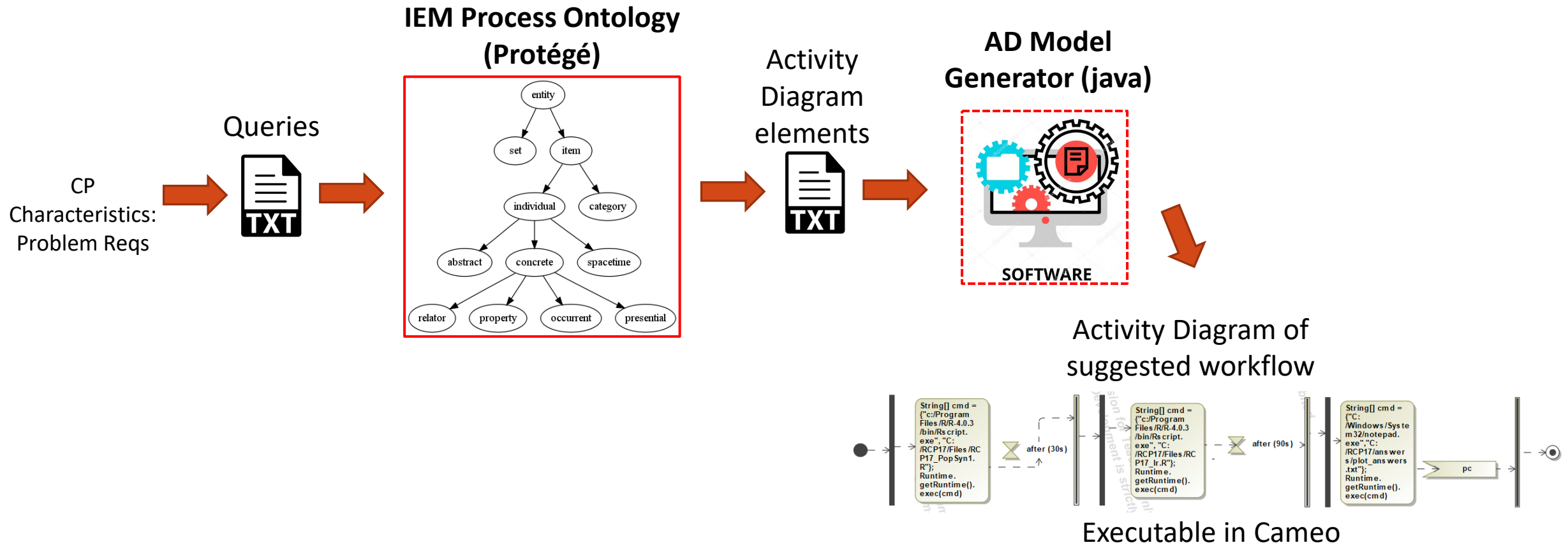
Graphical representation of business processes or workflows

Business process: collection of tasks an organization performs to create products, reach goals, provide value

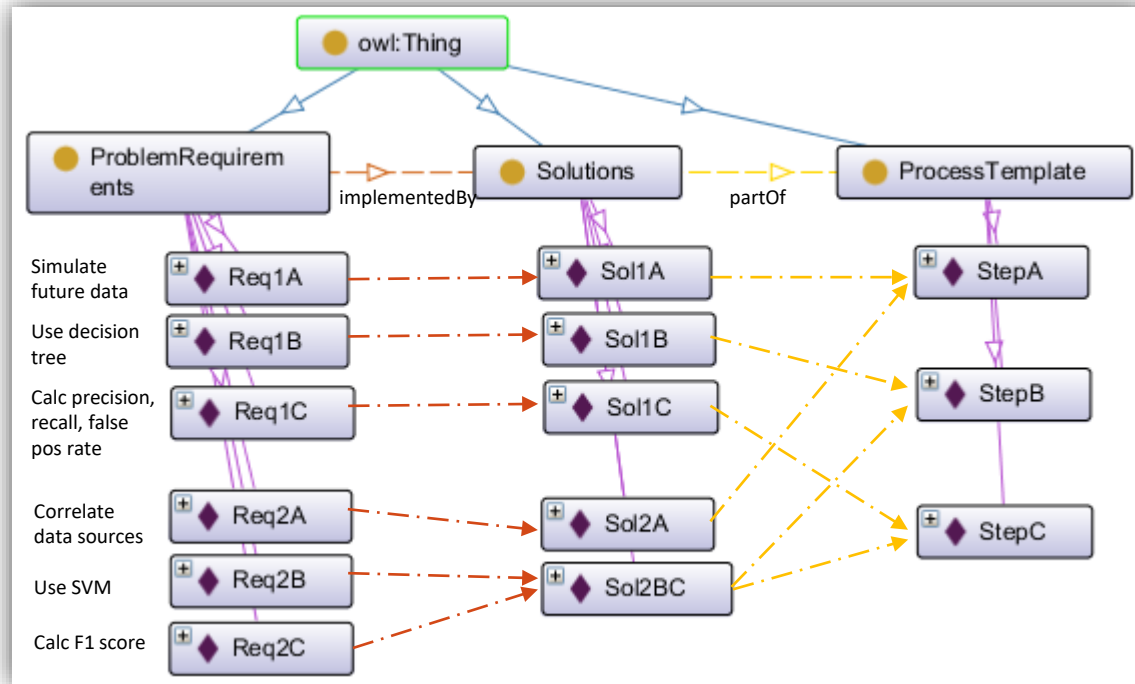
Process Modeling Languages: Business Process Modeling Notation (BPMN), Event Driven Process Chain (EPC), UML Activity Diagrams



Solution Components



IEM Process Ontology Overview



ProblemRequirements: characteristics that define problem

- Simulate future data based on data provided
- Use certain ML algorithm to predict
- Calculate precision, recall, and false pos. rate

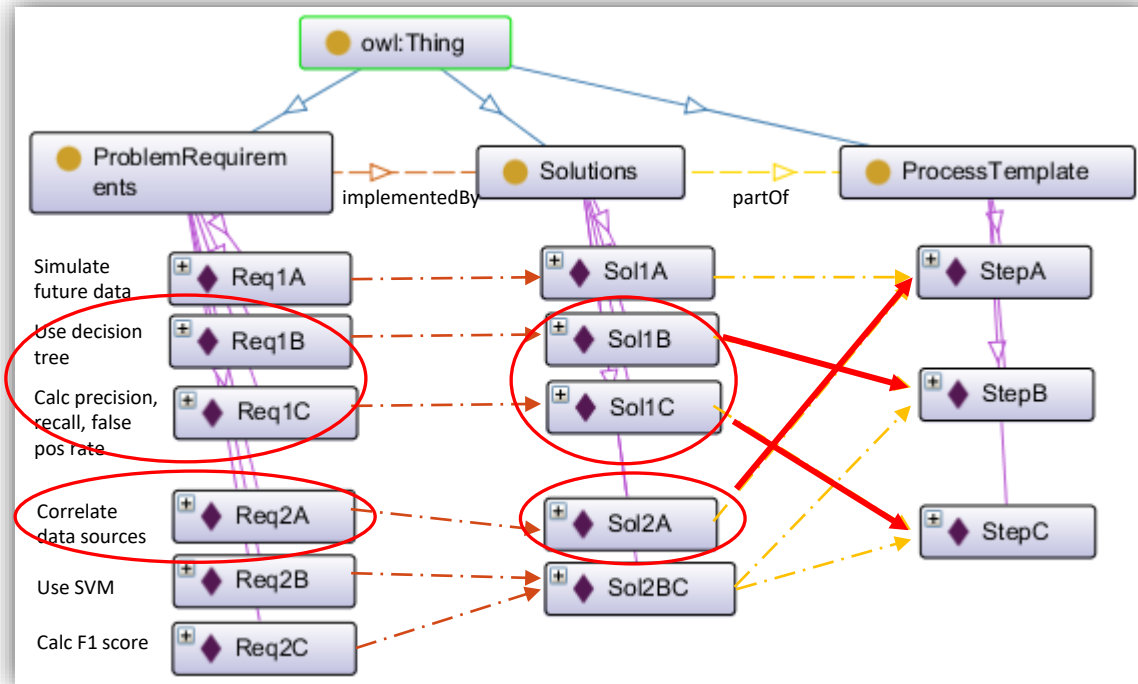
Solutions: software assets and/or manual activities that fulfill requirements

- Population assumptions based on SME input
- Population simulation and/or ML algorithms

ProcessTemplate

- Step A: Simulate Populations
- Step B: Down-select
- Step C: Calculate estimates

Small Example



Input New Problem:

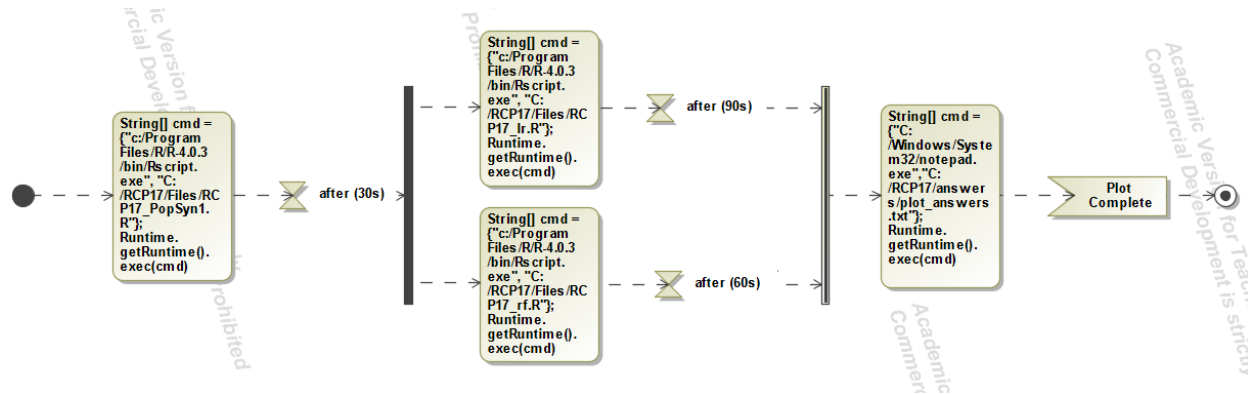
- Correlate data sources
- Use decision tree to predict
- Calculate precision, recall, and false pos. rate

Output ProcessTemplate:

StepA	StepB	StepC
Sol2A	Sol1B	Sol1C

IEM Process Ontology Development

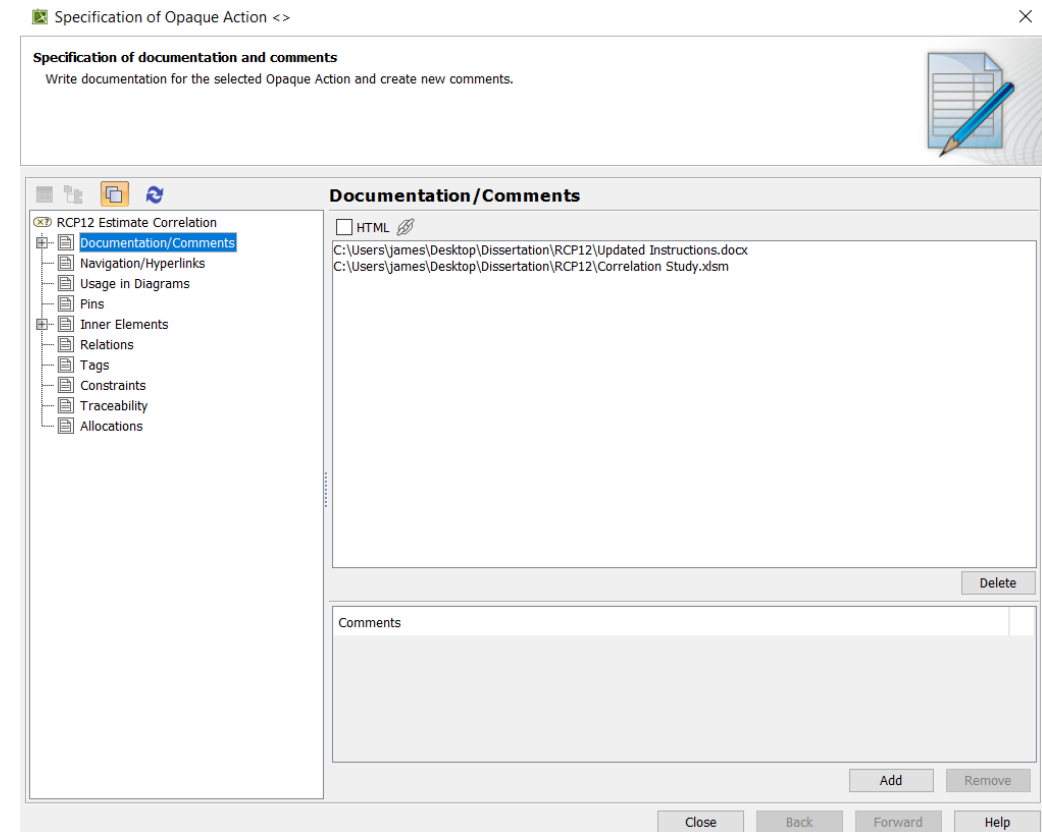
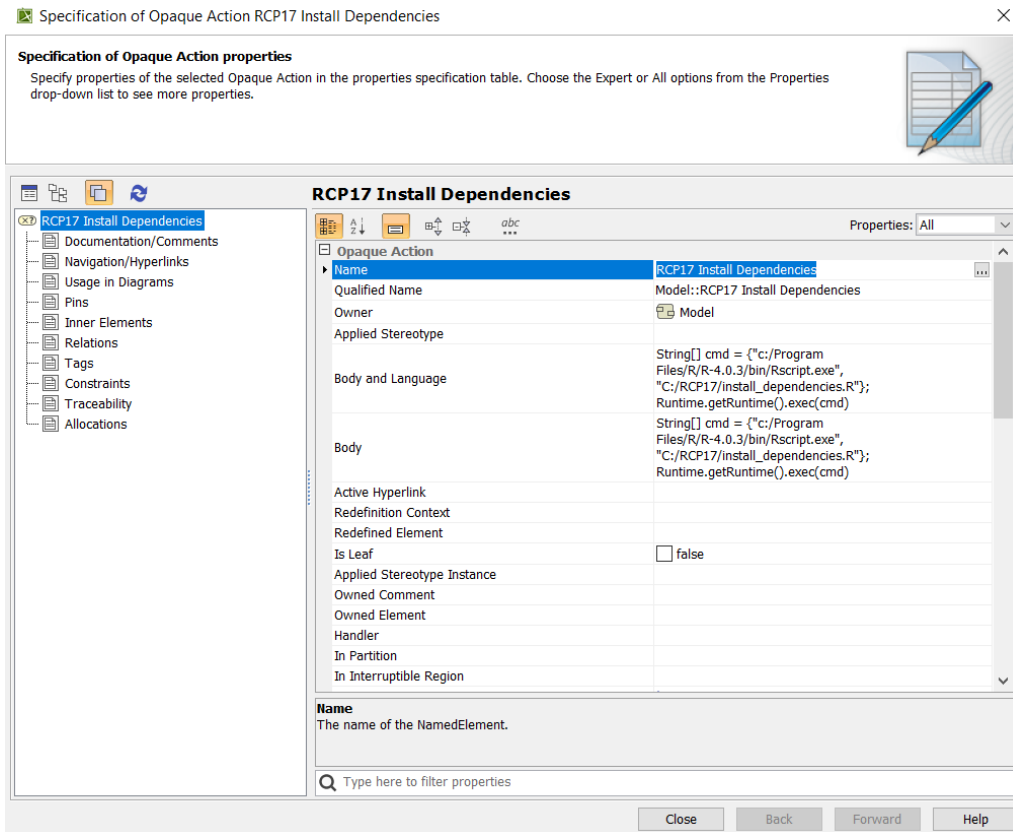
1. Use formal process modeling to document the solution workflows



Notation name	Concrete syntax	note
Initial Node	●	Starting point of an activity.
Activity Final Node	⦿	Marked the end or completion of an activity.
Fork/Join Node		Fork node has a single incoming edge and many outgoing edges, and it is the reverse for the join node. These elements help modelers in expressing the occurrence of concurrent sequences in an activity.
Decision/Merge Node	◇	Used to show the starting/ending of an alternative sequence in an activity. Usually merge nodes are used in conjunction with decision nodes to model loops within an activity.
Wait Time Action	⌞	Implies wait until certain action occurs.
Opaque Action	ⓧ	Introduced for implementation-specific actions. Allow modeler to implement action by running user defined script in languages such as Java, C++ and others.
Accept Event Action	⏏	Action that waits for the occurrence of an Event that meets specified conditions.
Action	▭	An Action is a named element that is the fundamental unit of an executable functionality. Regular actions are used to represent manual tasks in this work.

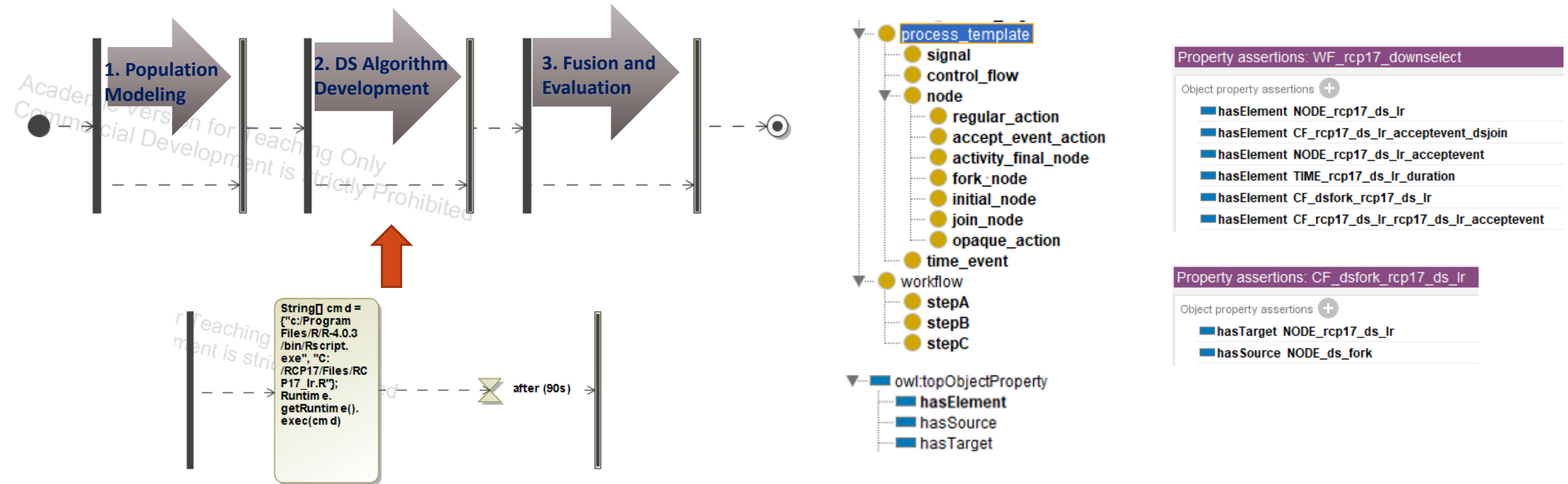
IEM Process Ontology Development

1. Use formal process modeling to document the solution workflows



IEM Process Ontology Development

- Generalizing solution methods by creating a process template and categorizing each part of the solution as a section of the template



IEM Process Ontology Development

3. Defining traceability links from problem requirements to each solution method – Problem Side

	Available Data								Data Characteristic								Down-Select Algorithm Types												Additional DS Requirements			Performance Evaluation			
	Correlation			IPE		ObsDist	Target Behavior		Correlation Amt		Num Obs		Num TB	Observable Data Types			Algorithms without Training Set						Algorithms with Training Set						HPTune	MukTrain	MukTest	CMR	SP	SPOQ	
	None	Obs	ObsTime	Yes	No	Yes	Yes	No	Full	Partial	<= 16	> 16	> 2	Both	ContOnly	DiscOnly	Alerts>=2	Cate>=2	Days>=8	PCADBS	NB	RF	DT	LR	GBM	SVM	HMM	NN							
Q1(RCP 1, 2, 3)		x			x	x	x		x		x					x	x																		x
RCP 4, 6		x			x	x		x			x					x	x																		x
RCP 5		x			x	x		x			x				x					x															x
RCP 7		x			x	x		x			x				x					x															x
RCP 8		x			x	x		x				x								x															x
RCP 9		x			x	x		x			x																								x
RCP 10		x			x	x		x				x		x										x											x
RCP 11		x			x	x		x			x			x										x											x
RCP 12	x				x	x		x			x		x											x											x
Q5(RCP 13, 14)			x		x	x		x				x		x										x											x
RCP 15			x		x	x		x				x																							x
RCP 16		x			x	x		x				x		x																					x
Q7(RCP 17, 18, 19)			x		x	x		x				x		x							x	x													x
IRCP 1		x			x	x		x				x			x																				x
IRCP 2		x			x	x		x				x		x										x											x
IRCP 3		x			x	x		x				x		x											x										x

- ◆ IRCP1
- ◆ IRCP2
- ◆ IRCP3
- ◆ LessThanEqualTo16
- ◆ LogisticRegression
- ◆ MoreThan16
- ◆ MultTest
- ◆ MultTrain
- ◆ NaiveBayes
- ◆ NeuralNetworks
- ◆ NewDownSelectNoTrain
- ◆ NewDownSelectTrain
- ◆ NoCorrelations
- ◆ NoTBInfo
- ◆ ObservableAndTemporalCorrelations
- ◆ ObservableCorrelations

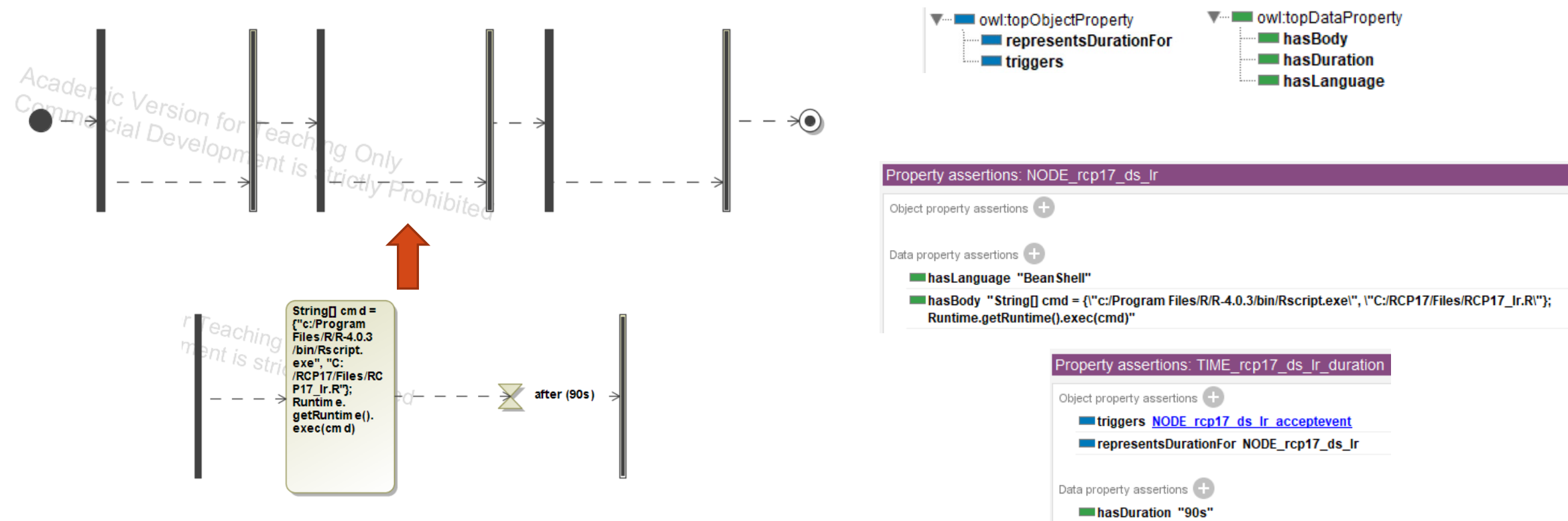
Property assertions: IRCP3

Object property assertions +

- hasTargetBehavior TargetLabelInfoProvided
- hasDSCharacteristics MultTrain
- hasDownSelectAlgorithm LogisticRegression
- hasCorrAmount FullCorrelations
- hasDSCharacteristics MultTest
- hasObsDistribution ObservableDistributions
- hasNumObs MoreThan16
- hasCorrelation ObservableCorrelations
- hasPerfEvalQuestions SysPerfObsQual
- hasObsDataVarType BothTypesObs
- hasIndepPerfEst IPENo

IEM Process Ontology Development

3. Defining traceability links from problem requirements to each solution method



IEM Process Ontology Development

3. Defining traceability links from problem requirements to each solution method – Solution Side

	Discrete Event Activity Counts Complete	Stochastic Optimization Complete	Discrete Event Activity Counts with IPE	Stochastic Optimization with IPE	Inverse CDF Copula Discrete with SME Correlation	Factorized Stochastic Optimization with DEAC	Inverse CDF Copula	Inverse CDF Copula and Logistic Regression with IPE	Inverse CDF Copula Continuous and Discrete	Rank Correlat ion Copula	2D Rank Correlat ion Copula	2D Rank Correlation Copula with SME Partial Correlation	Hierarch ical Bayes Net	Tree Augmented Naïve Bayes
MIEM Problem Specification														
no correlations					x									
observable correlations	x	x	x	x			x	x	x	x				
observable and temporal correlations											x	x		
IPE			x	x				x						
no IPE	x	x			x	x	x		x	x	x		x	x
observable distributions	x	x	x	x	x	x	x	x	x	x	x	x	x	x
target labels														x
no target label info														
full correlations	x	x					x	x	x	x	x			
partial correlations			x	x								x		
num obs restriction <= 16	x	x	x	x	x									
any observables									x	x	x		x	x
continuous observables only							x	x						
discrete observables only	x	x	x	x	x	x								

owl:topObjectProperty
 implementedBy
 hasReq

Class hierarchy: linreg

owl:Thing

trace_rules

linreg

problem_requirement

down_select

linear_regression

process_template

workflow

stepA

stepB

stepC

Description: linreg

Equivalent To

hasReq some linear_regression

SubClass Of

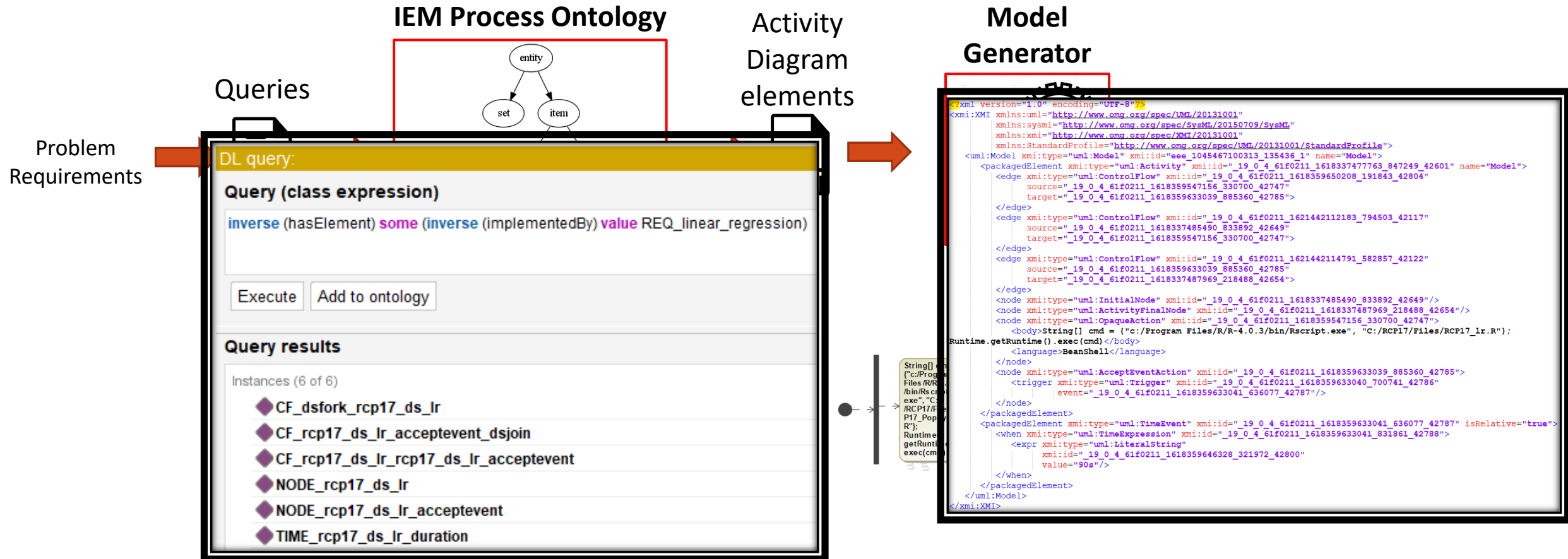
implementedBy value WF_rcp17_downselect

trace_rules

General class axioms

SubClass Of (Anonymous Ancestor)

Small Example



Simulation Tool

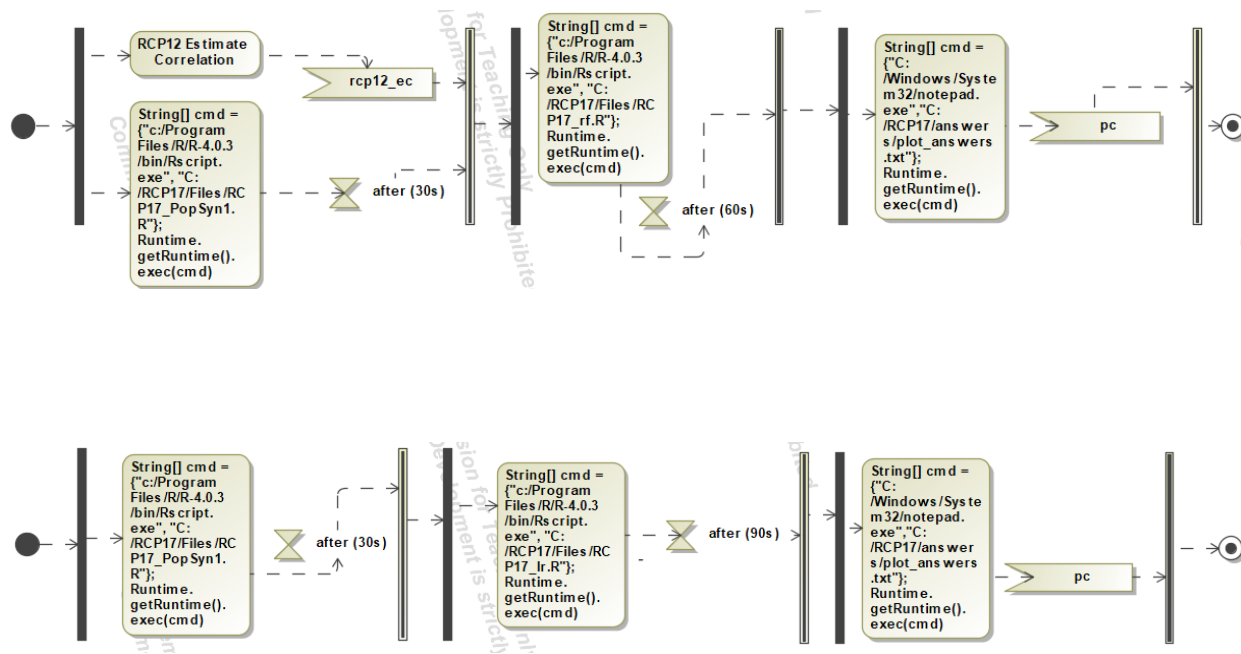


New Challenge Problems

Both problems use the GMU phishing experiment data set

- A: There are two models that use different sets of data to predict susceptibility to phishing. Since the two datasets are disjoint, correlation data between the sets do not exist. Forecast the performance of a model that uses all the data from both models.
 - Has continuous and discrete variables
 - Has binned data
 - Does not have full correlations
 - Has random forest
- B: There is a model that predicts staff members' susceptibility to phishing using staff data. Faculty data is not yet available. Forecast the performance of a model trained on staff data and tested on simulated faculty data.
 - Has full correlations
 - Has continuous variables
 - Has binned data
 - Has logistic regression

Suggested Workflows



1. Understand the overall flow of the suggested solution
2. Identify the software and documentation location by inspecting the activity elements of the activity diagram
3. Read the documentation, run the software modules as-is, inspect the input and output files, and understand its behavior.
4. Create a copy of the software modules that point to the new dataset of the challenge problems
5. Compare the input data file formats of the new problem to the input and output of the suggested module and identify places where appropriate changes are necessary
6. Make the appropriate code changes to the modules and execute

Research Contributions



Introduced IEM as digital twin technique for insider threat detection enterprises

Developed knowledge base of IEM expertise and model-based solution that can be used to rapidly prototype digital twins of new IEM scenarios

Future Works

Accurate problem requirement generation

Flexible template generation

Full simulation capabilities of systems modeling software

IEM – Ransomware protection



Thank you

Submitted full manuscript at SysCon 2022

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