

Data and Analysis Code for EFSM Inference

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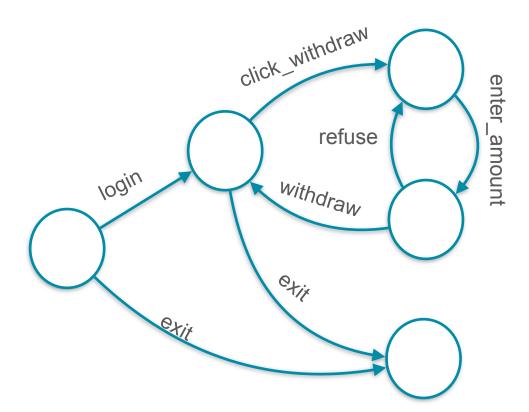
Computational Extended Finite State Machine Inference

login
[pwd=123,intpwd=123]
click_withdraw
enter_amount
withdraw
[x=5,bal=10]
click_withdraw
enter_amount
refuse
[x=6,bal=5]
enter_amount
withdraw
[x=5,bal=5]
exit

login
[pwd=123,intpwd=123]
click_withdraw
enter_amount
withdraw
[x=20,bal=100]
exit

login [pwd=xyz,intpwd=zz] exit

exit





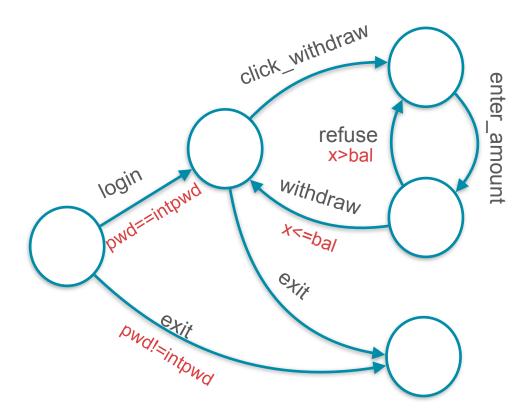
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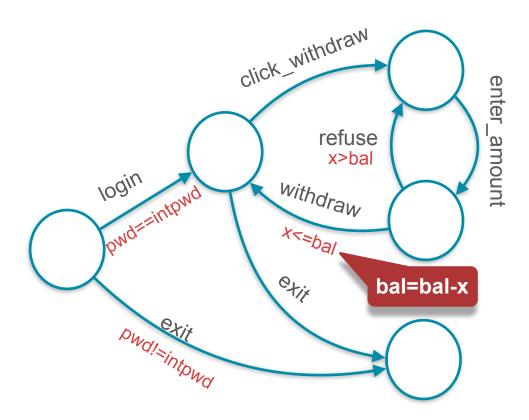
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Implementation

MINT EFSM Inference Tool
Java implementation - 19 KLOC, ~4 KLOC GP.
https://bitbucket.org/nwalkinshaw/efsminferencetool

GP

State Transition System Inference

Operators & Operands

Crossover & mutation

Configuration

Min. merge criteria

Merging heuristics

Classifier inference for constraints



Implementation

revisions, default settings, ...

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+ - * Pwr Cos AND NOT SQRT,TAN, /, ... Operators, random seeds, coefficients

Chromosom es, generations, term. criteria

k=0,1,2,3,...

BlueFringe, k-Tails, exhaustive? Daikon, C4.5, NaiveBayes , any WEKA



Experimental Variables

+ - * Pwr Cos AND NOT SQRT,TAN, / , ... COTS, network protocols, ...

Random
Test Suites
Judicious
execution, ...

Daikon, C4.5, NaiveBayes , any WEKA

Chromosom es, generations, term. criteria

BlueFringe, k-Tails, exhaustive? k=0,1,2,3,...

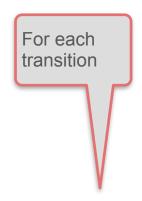
revisions, default settings, ...

Chickory, AspectJ, Log files, ...

Operators, random seeds, coefficients







Model inference

Large CSV files of inferred versus real values.





For each transition

Model inference

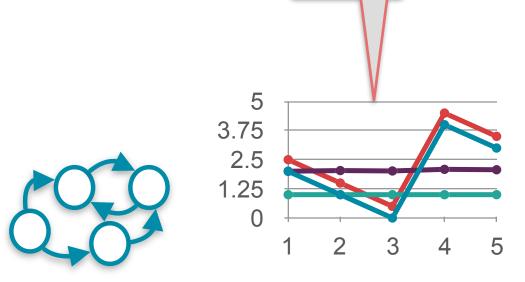


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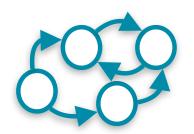


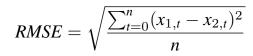
For each transition



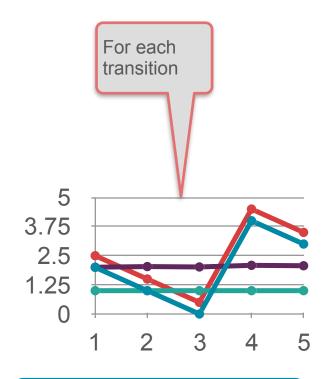
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Large CSV files of inferred versus real values.





$$NRMSE = \frac{RMSE}{max(x_1) - min(x_1)}$$



Root mean square error (punishes large errors)

Normalised RMSE

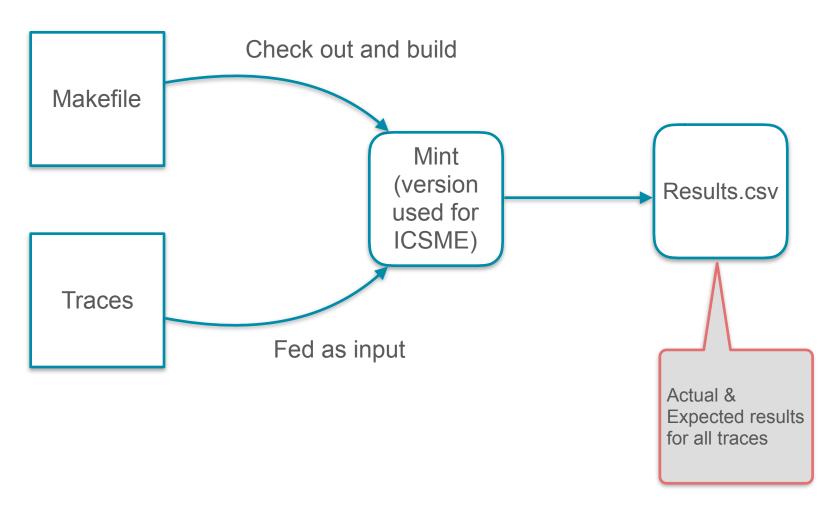


Challenges for Reproducibility

- Inference setup
 - Very large number of parameters (can be sensitive).
 - Default values can change from revision to revision.
- k-Folds Cross Validation
 - Cumbersome to process traces.
 - Error calculation is not part of MINT.



Artefact - Experiment Execution





Artefact - Analysis

- Results file contains raw data
- Provide R code to process
 - Includes code to generate all plots and stats in paper
- Explained with Jupyter
 - Literate programming for R







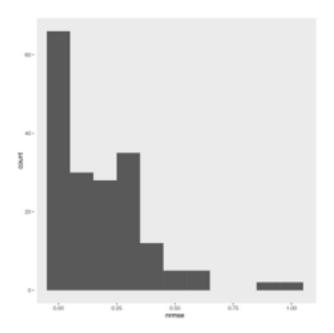
LiftDoors Example

The following figures replicate the plots in our paper. They show how to visualise the results from the update function inference step.

In this example there is one variable analysed. In the outputs from MINT a variable has two columns, the original value (t in this case, and its inferred value from the model t(inferred)). Note that R replaces brackets in column headers with dots, hence the variable names we analyse here are t and t.inferred..

Summarising the results

The histogram shows the distribution of NRMSE values across the traces used in each of the 10 folds of cross validation. NRMSE is the normalised root-mean square error and is best when its value is zero.





Any questions?

