

Tutorial on Neural Network Control Systems

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Requirements

MATLAB: requires license (academic, student, professional,...)

Dependencies: Breach (free and open source)

MATLAB Toolboxes:

- a) Simulink
- b) Deep Learning
- c) Symbolic Matlab Toolbox (Breach)
- d) Global Optimization Toolbox (Breach)
- e) Optimization Toolbox (Breach)
- f) Control System Toolbox (Tutorial)
- g) Stateflow (Tutorial)

Our Code for Neural Network Control Systems: https://github.com/nikos-kekatos/NNCS_matlab



Installation

Readme :

https://github.com/nikos-kekatos/NNCS_matlab/tree/nikos_comb

No special installation, as it is MATLAB-based and it is platform independent.

Provided the existence of toolboxes, it should work without (many) manipulations.

What is your OS?

github.com/nikos-kekatos/NNCS_matlab/tree/nikos_comb

- Optimization Toolbox
- Parallel Computing Toolbox (optional)

Models

- Control System Toolbox (models with PID control)
- Simulink Control Design Toolbox (mode with gain-scheduling)
- Image processing Toolbox (optional, previous dependency was avoided)
- Model Predictive Control toolbox (models with MPC control)
- Fuzzy Logic Toolbox (model with fuzzy inference rules & look-up table)
- Symbolic Matlab Toolbox (QuadMPC)

Dependencies

Apart from the MATLAB toolboxes, we use the [Breach](#) toolbox. Breach is a Matlab tool for simulation-based design of dynamical/CPS/Hybrid systems. According to the guidelines,

- Download Breach files
- Setup a C/C++ compiler using mex -setup
- add path to Breach folder
- Run InstallBreach

Installation

No special installation is required. The user has to add `Breach` and this repo (default name: `NNCS_matlab`) to the path. This can be done by right-clicking on the root directory and selecting `Add to Path >> Select Folder and SubFolder`. For Breach, the user has to run in the command window of MATLAB `>> InitBreach`.

Branches

There are several branches but they are not all properly maintained. The `nikos_comb` contains the latest version of the code and this is the branch that should be used. The `master` contains the stable version of the code. The `Akshay` branch was used to check the code in a Windows machine. The `nikosk` branch initially contained the ode without Breach falsification but only with coverage. The `falsif_breach` contained the initial integration of Breach without the iterative process. The `nikos_dev` contained the full procedure for a single nominal controllers.



Problem Definition

Standard (Simpler) Case

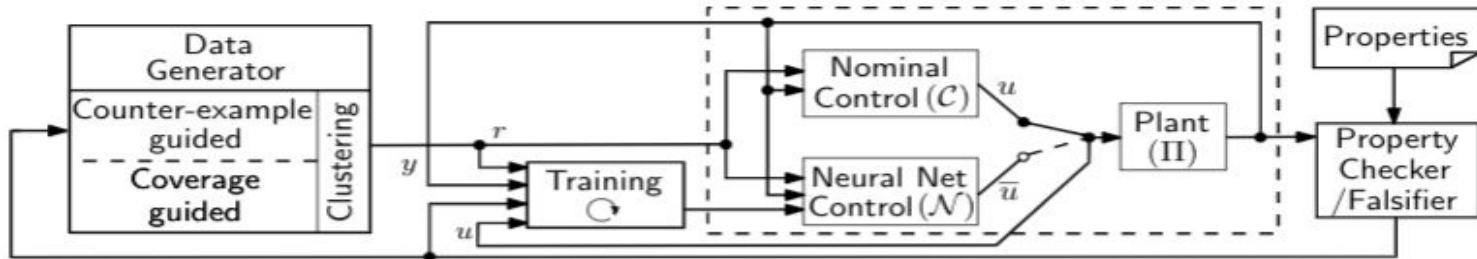
Given i) a closed-loop (controlled) system S , which is composed from a nominal controller C and a plant P ($S:=C||P$), and ii) a control objective/specification φ written in Signal Temporal Logic, design a neural network controller C_{nn} that such that the new closed-loop system $S_{nn}:=C_{nn}||P$ satisfies the property φ .

Regarding the satisfaction of the property, we do not aim at “full” formal verification but we instead target the lighter version of semi-formal verification (using falsification).



Methodology

1. Closed-loop modeling (design or given)
2. Trace generation (finding good examples), e.g. via coverage
3. Data Selection (trimming, preprocessing, etc.)
4. Neural Network Training
 - a. Architecture
 - b. Training hyperparameters
5. New closed loop system (with a neural network)
6. Falsification and Retraining in a loop





What do you need to run a model?

Inputs to the tool:

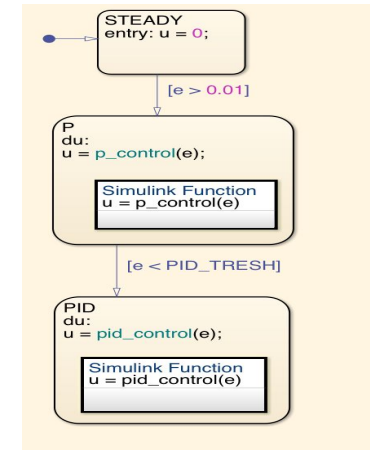
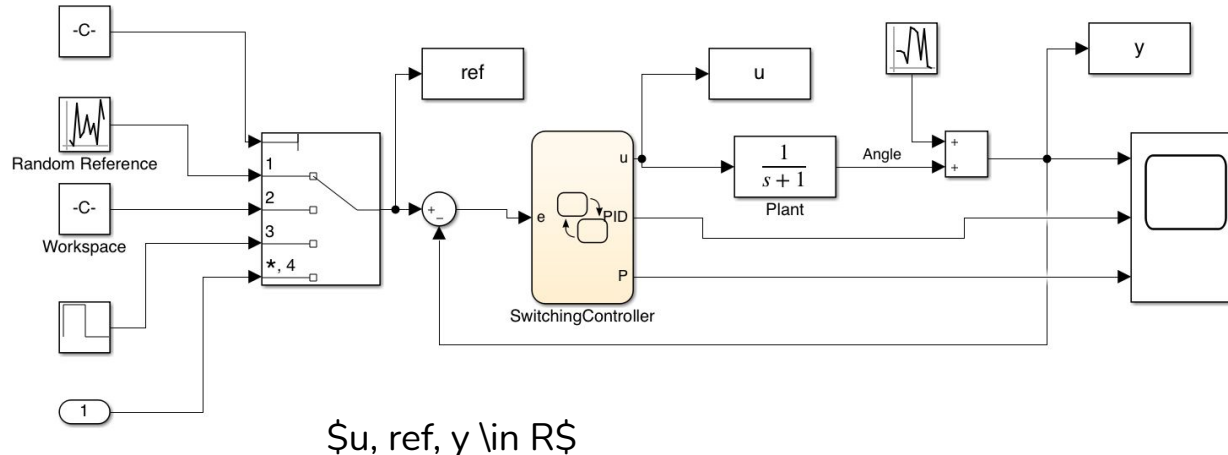
1. Nominal Closed-loop System in Simulink (not automated for MATLAB models)
2. Specification defined in Signal Temporal Logic (syntax of Breach)
3. Configuration

Then we need to run the “main” file which contains all the steps of our algorithm.
There are several “main” files for different models you can use them as references.

Example: Switched Controller

This example is a simple Simulink model which describes a closed-loop system.

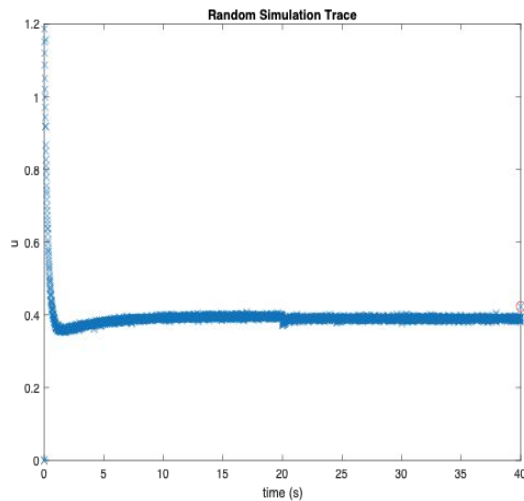
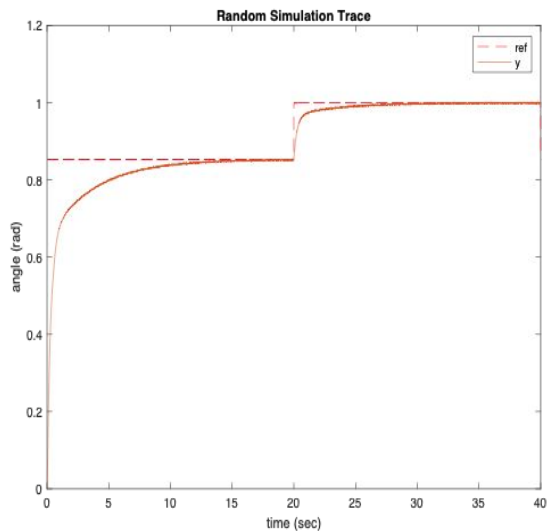
- Plant: linear
- Controller: switched (from P -> PID when a condition is satisfied)
- Property: if there is a step reference change, then plant output y should reach and stay close to the given reference.



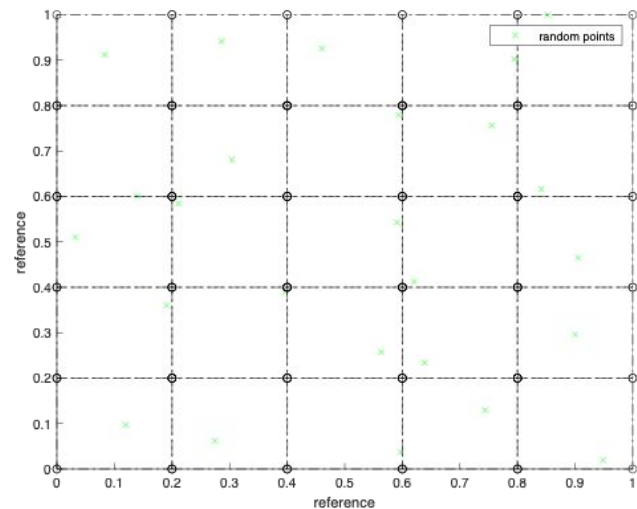


Trace Generation

Added small perturbation as noise.



Coverage: Piecewise constant reference with two pieces





Data Selection (1/2)

Data Selection can be important: Avoid duplicates, reduce size, choose area where you want to keep or delete data points.

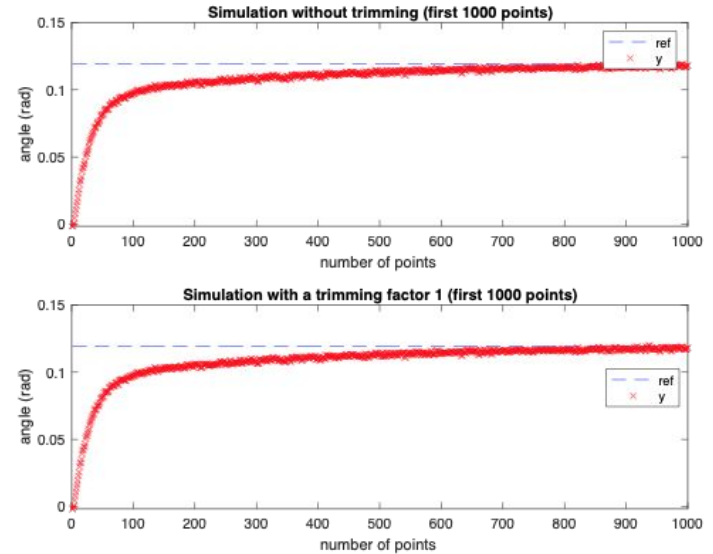
Three main options (named in the code as follows):

- Preprocessing
- Trimming
- Steady-State trimming

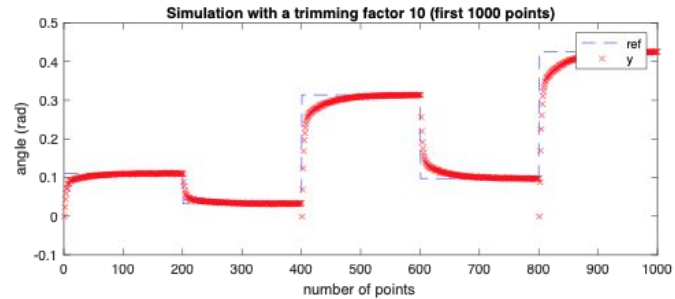
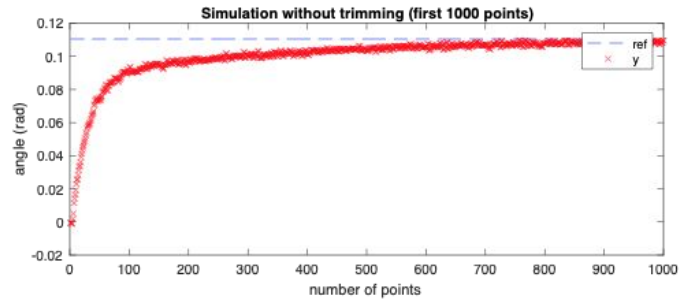
#different reference signals/simulation traces: 25

Each simulation $T=40s$, $dt=0.01$

In total: 100,000 points for y_{ref} , y_u , y_y With trimming: 62475



Data Selection (2/2)

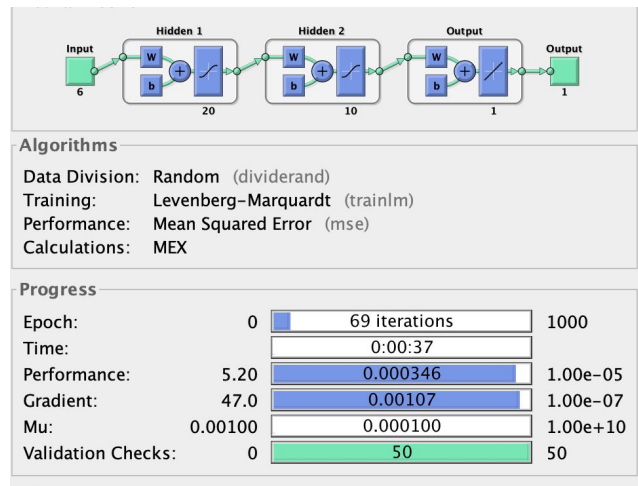
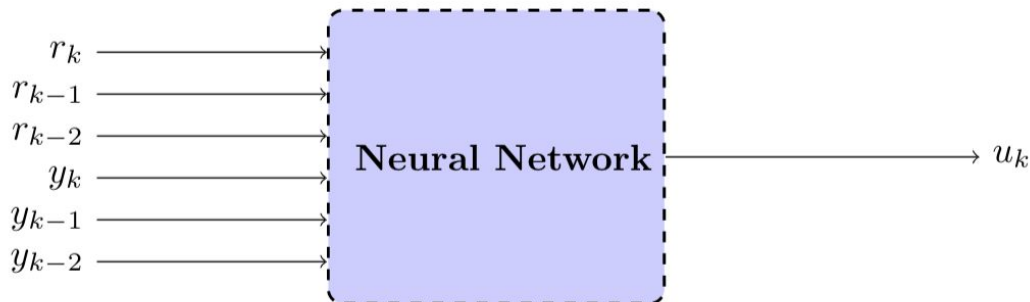


Training

== Need to choose (1) the architecture and (2) the training parameters.

- How to capture the time-dependence/evolution ? Use past values for r_{ref} , u , y .

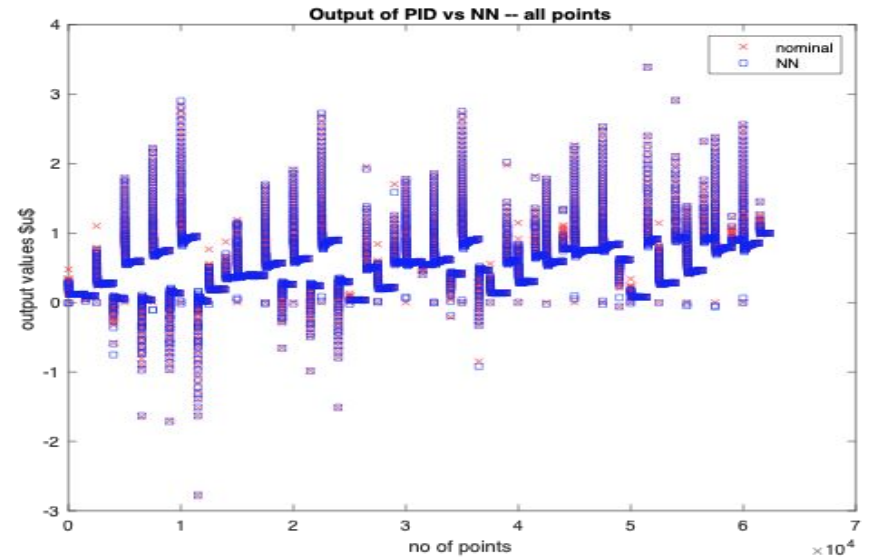
```
training_options.neurons=[20 10]; % 2 layers with 20 and 10 neurons each
training_options.use_error_dyn=0; % answer: 0 or 1, use error dynamics or not
training_options.use_previous_u=0; % answer: integer, number of previous u values
training_options.use_previous_ref=2; % answer: integer, number of previous references
training_options.use_previous_y=2;
```



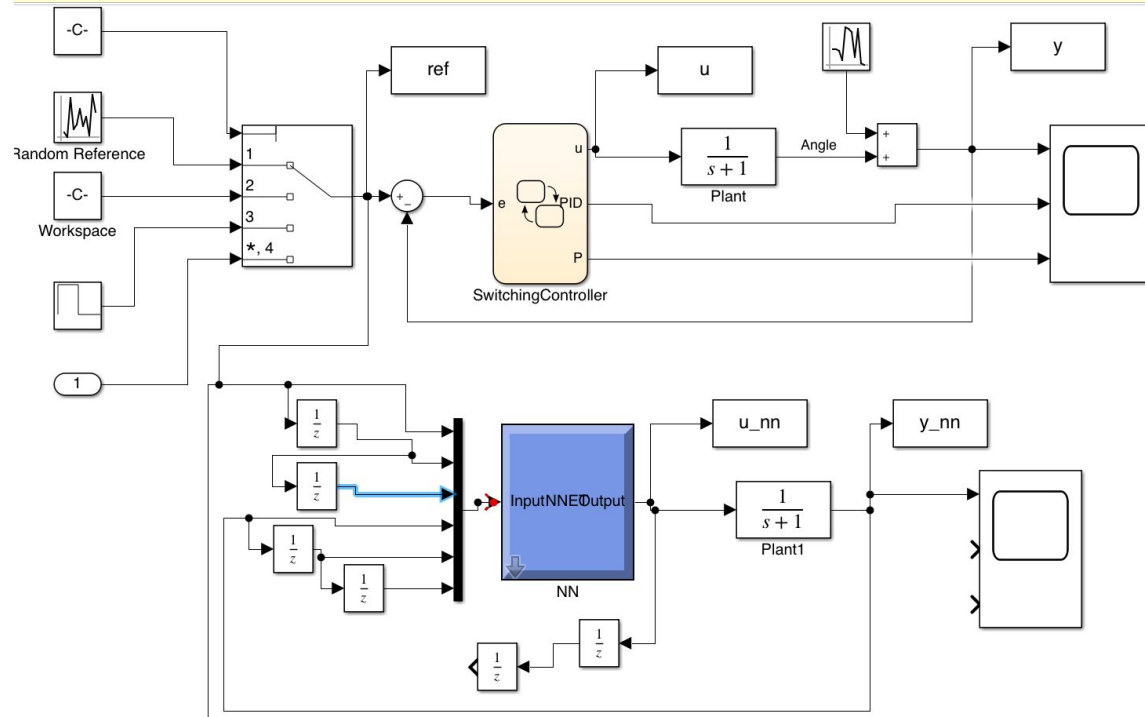


Training/Evaluation

- In Open Loop (with the same inputs)
- Neural Network Inputs: $r(k)$, $r(k-1)$, $r(k-2)$, $y(k)$, $y(k-1)$, $y(k-2)$
- Neural Network Output: $u(k)$

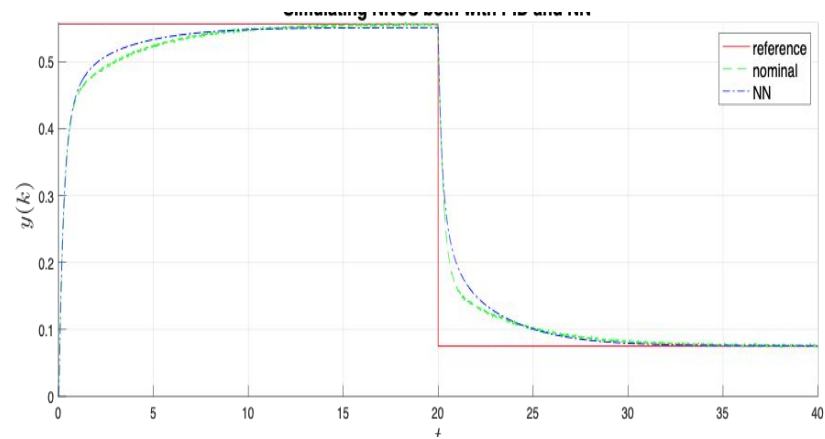
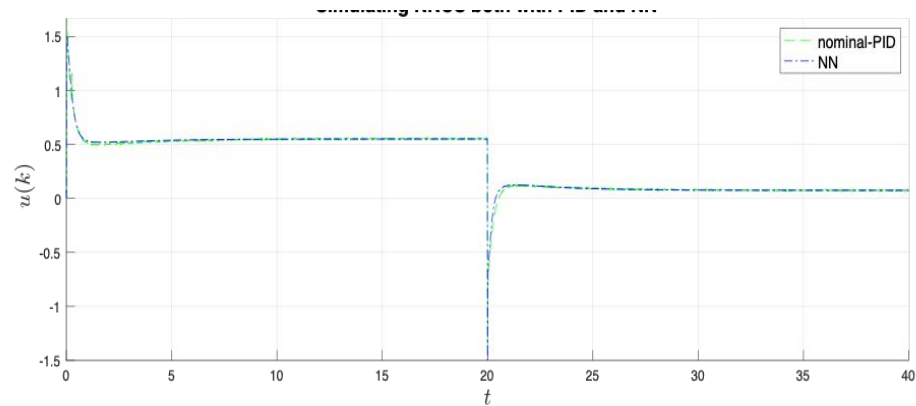


Create NNCS





Simulating the new NNCS





Testing against the STL property

Out of 25 traces, all NN generated traces satisfy given STL property (tested with Breach).

param sim_time=39.9, r=0.1, dt=0.01, T=10, d=0.01, T_1=6, T_2=4, e=0.05, ov=0.05

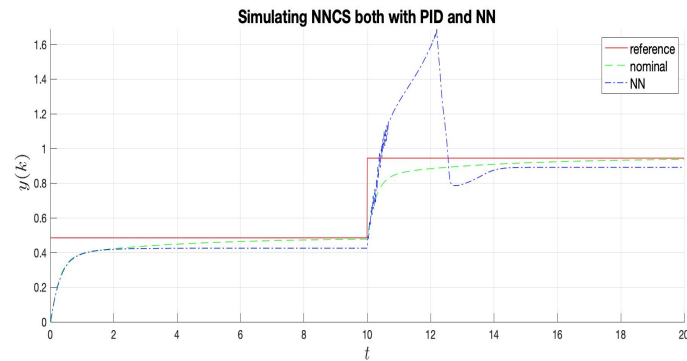
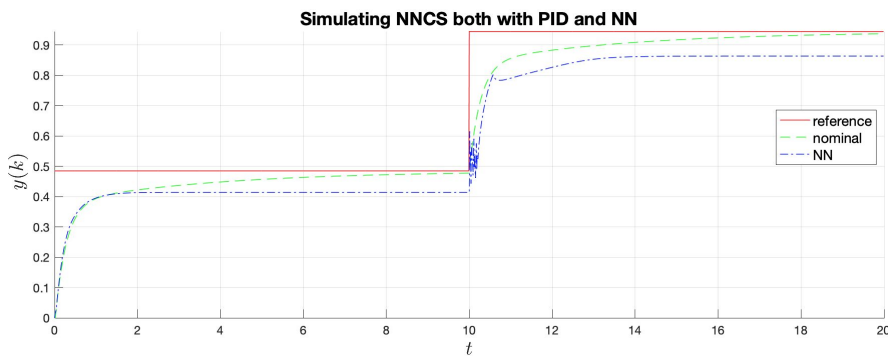
phi_1:= alw_[0,sim_time]((abs(ln1[t+dt]-ln1[t])>d or [t]==0) => (ev_[0,17] alw_[0,3] (abs(y[t]-ln1[t])<e)))

Is it enough?

Falsification with Breach

We use Breach to falsify the NNCS via sampling and optimization methods.

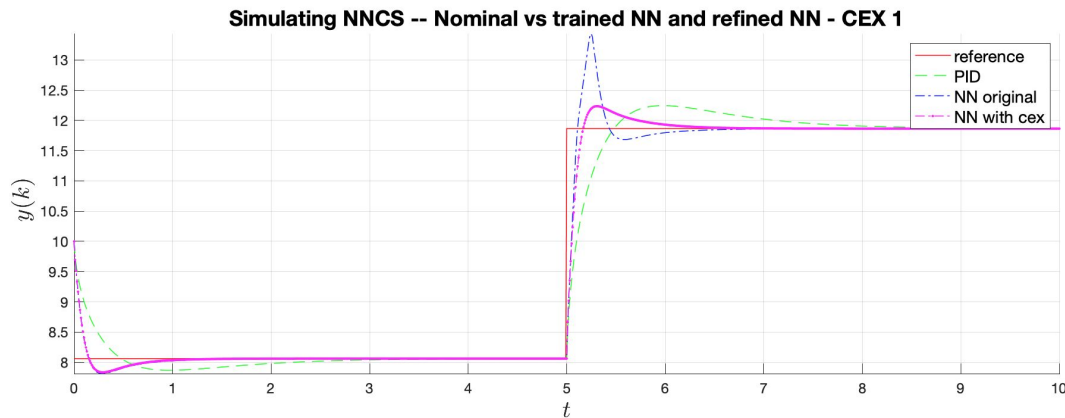
Fast restart (to speed up computations).



Falsification/Retraining

The nominal controller uses error dynamics, but we had to separate the references from the plant outputs. We chose $\mathcal{I} = \{R_k, R_{k-1}, R_{k-2}, Y_k, Y_{k-1}, Y_{k-2}\}$. We started with 25 examples and had 100,000 samples per state. We trained with [20, 10] neurons and falsification yielded 2/100 CEX. Retraining eliminated these cex. The second falsification/retraining loop eliminated 5 CEX.

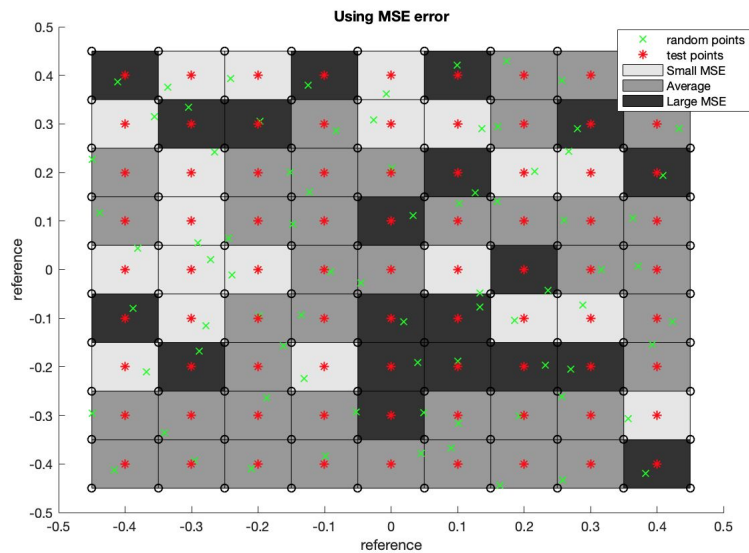
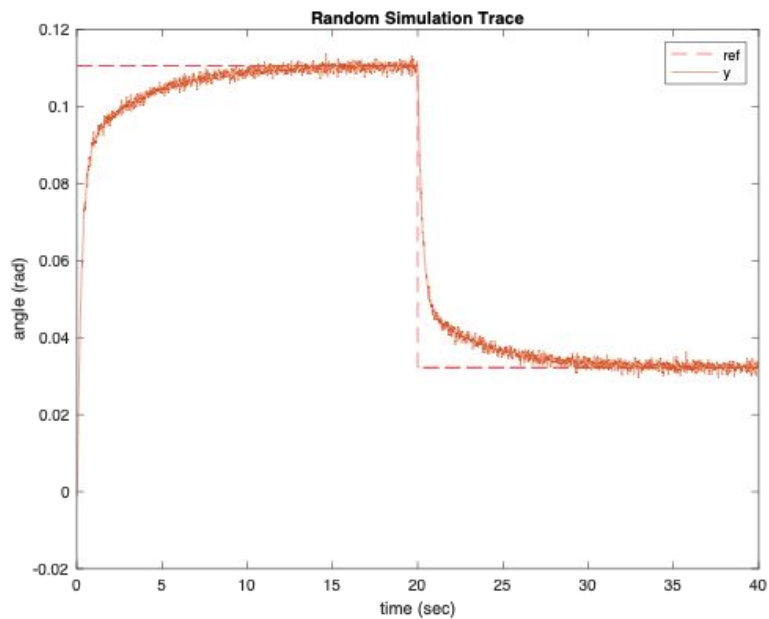
Able to design a controller that eliminates all CEX.



END



Backup 1





Some Discussion Points (Friday, 19 March)

NNCS:

- 1) Control Systems & Modeling (Simulink/MATlab)
- 2) Simulation & Falsification
- 3) Neural Network Structure & Architecture
- 4) Temporal Logic

CPSdebug? “MATLAB”

A: recurrent NN

E: Minimal part of the trace (capture memory but not too much)

AD: temporal implicants in Breach (missing part of Until)

CPS debug -> critical time (neighborhood)

Question: Which is the last point of the trace?

Retraining algorithms

