
```
classdef ARIC < handle
```

ARIC - Neuro-fuzzy controller class

ARIC properties

```
properties
    A, b, c, D, e, f, v, s, z, rhat;
end

properties
    n, h, rho, rhoh, beta, betah, gamma;
end

methods
```

Class construction Input argument 'param' contains the settings the learning process.

```
function obj = ARIC(param)
% ARIC Construct an instance of this class
% n = number of states + 1 (4 in the cart pole example)
% h = number of rules (13 in the cart pole example)
obj.n = param.aric.n;
obj.h = param.aric.h;

obj.rho = param.aric.rho;
obj.rhoh = param.aric.rhoh;
obj.beta = param.aric.beta;
obj.betah = param.aric.betah;
obj.gamma = param.aric.gamma;

% Action-state Evaluation Method weights
obj.A = rand(obj.n, obj.n); % SQUARE MATRIX
obj.b = rand(1, obj.n); % ROW VECTOR WITH LENGTH N
obj.c = rand(1, obj.n); % ROW VECTOR WITH LENGTH N

% Action Selection Network
obj.D = rand(obj.h, obj.n); % HxN MATRIX
obj.e = rand(1, obj.n); % ROW VECTOR WITH LENGTH N
obj.f = rand(1, obj.h); % ROW VECTOR WITH LENGTH H
obj.v = rand(); % SCALAR
obj.s = rand(); % SCALAR
obj.z = rand(obj.h, 1); % COLUMN VECTOR WITH LENGTH H

end
```

Not enough input arguments.

```
Error in ARIC (line 23)
    obj.n = param.aric.n;
```

Master loop This function is called by Simulink in order to compute the control force based on the states

```
function F = getControllerOutput(obj, x)
```

```

    % Action selection network
    u = obj.fuzzyInference(x); % Compute u based on fuzzy
rules
    p = obj.confidenceComputation(x); % Compute confidence p
    obj.updateWeights(x); % Apply learning algorithms
    F = obj.stochasticActionModifier(u, p); % Compute
Stochastic Object Modifier F
end

```

Action-State Evaluation

```

function [y, v] = stateEvaluation(obj, x)
    % Implementation of Maxime's AEN1 - neural network of AEN
    y = arrayfun(@sigmoid, obj.A*x); % Neural network
    v = obj.b*x + obj.c*y;
end

function rhat = internalReinforcement(obj, x, vNew)
    % Implementation of Maxime's AEN2 - Computation of the
internal reinforcement r_hat
    if abs(x(3)) > pi/15    %[rad]
        rhat = -1 - obj.v;
    elseif abs(x(1)) > 2.4  %[m]
        rhat = -1 - obj.v;
    else
        rhat = obj.gamma*vNew - obj.v;
    end

    % Update v
    obj.v = vNew;
end

function updateWeights(obj, x)
    % Compute necessary values from AEN
    [y, vNew] = obj.stateEvaluation(x);
    obj.rhat = obj.internalReinforcement(x, vNew);

    % Update AEN weights
    obj.b = obj.b + (obj.beta*obj.rhat*x)'; % b is stored as a
row vector
    obj.c = obj.c + (obj.beta*obj.rhat*y)'; % c is stored as a
row vector
    obj.A = obj.A + (obj.betah*obj.rhat).*sign(obj.c)'.*y.*(1-
- y)*x';

    % Update ASN weights
    obj.e = obj.e + (obj.rho*obj.rhat*obj.s*x)';
    obj.f = obj.f + (obj.rho*obj.rhat*obj.s*obj.z)';
    obj.D = obj.D + obj.rhoh*obj.rhat*obj.z.*(1-
obj.z).*sign(obj.f)'*obj.s*x';
end

```

Action Selection Network methods

```

function u = fuzzyInference(obj, x)

```

```

% Based on Bart's FuzzyInference

w = fuzzifier(x, obj.D);
m = defuzzifier(w);

utop=0;
ubot=0;

for i = 1:length(w)
    utop = utop + obj.f(i)*m(i)*w(i);
    ubot = ubot + w(i)*obj.f(i);
end
u = utop/ubot;
end

function p = confidenceComputation(obj, x)
    obj.z = arrayfun(@sigmoid, obj.D*x);
    p = obj.e*x + obj.f*obj.z;
end

function u_mod = stochasticActionModifier(obj, u, p)
    % Implementation of functions o and s in the paper

    q = (p + 1)/2;
    if u > 0
        u_mod = q;
    else
        u_mod = 1 - q;
    end

    if sign(u_mod) ~= sign(u)
        obj.s = 1 - p;
    else
        obj.s = -p;
    end
end
end

```

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