REQUIRED VALUE CLASSIFICATION USING KOHONEN NEURAL NETWORK

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Abstract. In this paper we describe situations classification using Kohonen neural network. Results are demonstrated on line following robot, where different curves are classificated. Each classified situation have corresponding required value output, which go into low level controll process. Network weights estimating is processed in real time.

Key words and phrases. kohonen neural network, learning controll, iterative learning, PID controll, mobile robot.

Klasifikácia žiadanej hodnoty Kohonenovou neurónovou sieťou

Abstrakt. V článku je popísaná klasifikácia situacií Kohonenovou neurónovou sieťou. Výsledky sú demonštrované na robotovi, sledujúceho čiaru, kde sú klasifikované rôzne typi zákrut. Každá situácia má zodpovedajúci výstup, ktorý je vstupom do nižšej úrovne riadenia.

Keywords. kohonenová neurónová sieť, inteligentné učenie, iteraivne učenie, PID riadenie, mobilný robot.

Introduction

Is well know, iterative learning controll can handle any repetitive controll process weel [1], [2]. Most of aplications we can found in manipulators controll in industry.

Some problems, have iterative (or semiterative) character, but more then one required output sequence is required. Consider problem of line following robot : to controll robot speed, is necessary to know curve shape. If we can compute asociative memory, where input is line shape and output is required speed, we can handle this problem.

1. Problem formalisation

Consider two wheel (two motors) robot with diferential drive [3]. For motors controll signals we can write

$$
\begin{aligned}
r(n) &= v(n) - d(n) \\
l(n) &= v(n) + d(n)
\end{aligned}
\tag{1}
$$

Where
$r(n)$ is right motor controll output
$l(n)$ is left motor controll output
$v(n)$ is common controll input
$d(n)$ is difference controll input

$d(n)$ can be computed well using PD controller [5]. Our goal is to properly estimate $d(n)$ which is speed of robot. Robot motion equation can be written as

$$
\begin{aligned}
\theta(n) &= (1 + \alpha)\theta(n - 1) - \alpha\theta(n - 2) + b_0(r(n) - l(n)) \\
\nu(n) &= \alpha\nu(n - 1) + (1 - \alpha)b_1(r(n) + l(n))
\end{aligned}
\tag{2}
$$

Where
θ is robot orientation (Yaw angle)
ν is robot speed
α is robot inertia constant, within $(0, 1)$
b_0 and b_1 are constants depending on wheel distances, wheel diameters, gear ratio, and maximum speed

If required θ and ν are know, we can very well use pure PID controll defining error as $e_\theta(n) = \theta_r(n) - \theta(n)$, $e_\nu(n) = \nu_r(n) - \nu(n)$, In our line following problem, is $\nu_r(n)$ uknow, value $\theta_r(n)$ $\theta_r(n)$ readed from line position sensor. In fact, inertia of motors is much smaller than inertia of whole robot, for controll $\theta(n)$ we can still use PD controller, and required value let be $\theta_r(n) = 0$, which mean robot is straight on line.

Our goal is to properly estimate $v(n)$ and controll robot speed, to give enough time for PD controll to setup $\theta(n)$. We are minimizing $e_\theta(n)$ and maximizing $\nu(n)$ (corespoding to $v(n)$).

2. Controller design

On figure 1 is block diagram of proposed controller.

Input is line poistion from sensor $s(n)$. All values are normalised into $\langle -1, 1 \rangle$. Required value of $\theta(n)$ is marked as $r(n)$. In simpplificated case, can be set to 0. Computed error is input into two PD controllers (PID in general, I-term is equal to zero, because 2 (θ part) have pole on [1, 0]. If I-term is nonzero, there will be double pole on [1, 0] and system will be unstable).
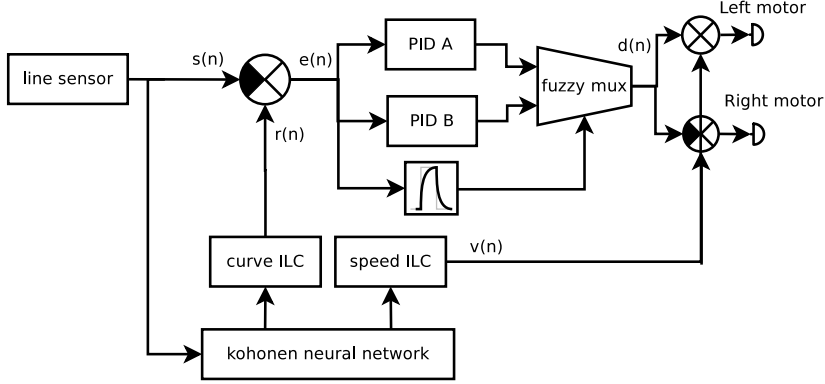
PD ontroller outputs are mixed using fuzzy mux, which works as

$$
d(n) = s(n)a(n) + (1 - s(n))b(n)
\tag{3}
$$

Where
$a(n)$ is PID A output
$b(n)$ is PID B output

Figure 1. Robot controller.

$s(n)$ is select input, from interval $\langle 0, 1 \rangle$

Select output is produced by nonlinear low pass filtered signal of $e(n)$ as

$$s(n) = \begin{cases} s(n-1)k + (1-k)|e(n)| & \text{if} |e(n)| < s(n-1) \\ |e(n)| & else \end{cases} \tag{4}$$

Where k is filter constant, from $(0, 1)$. This filter smooth error values, and rise to huge output value if error immediately rise.

This system properly switching between two controllers, one for straight line one for curved line.

To estimate $v(n)$ is Kohonen neural network used [4] and iterative learning controll. Input into neural networ is vector of $s(n)$ values (implemented as fifo, in experiments with size M = 16). Mark this vector as $S(n) = [s(n), ..., s(n-M)]$.

Our goal, is to train network to clasification vectors $S(n)$. Depending on classes count, we choose coresponding neurons count N. In experiments chooses as 16. Which means, 16 different curves shapes can be recognized.

J-th neuron transfer function can be written as

$$y_j(n) = \sum_{i=0}^{M-1} |s(n-i) - w_j(i)| \tag{5}$$

Where w_j are neuron weights, in initialization chosen randomly and modifieng during learning process.

Neuron with less $y_q(n)$ will be marked as winning neuron, and weights modification will be processed as

$$w_q(n) = \eta w_q(n-1) + (1-\eta)S(n) \tag{6}$$

Where η is learning rate, from $(0,1)$ close to 1. Weights of neuron are slowly adapted to corresponding pattern. This can be illustrated on figure 2. Input was 2 dimensional, and goal is to find centers (green) of some data set (red). input data set was produced by some Markov process. On figure 3 are show $y_q(n)$ values.
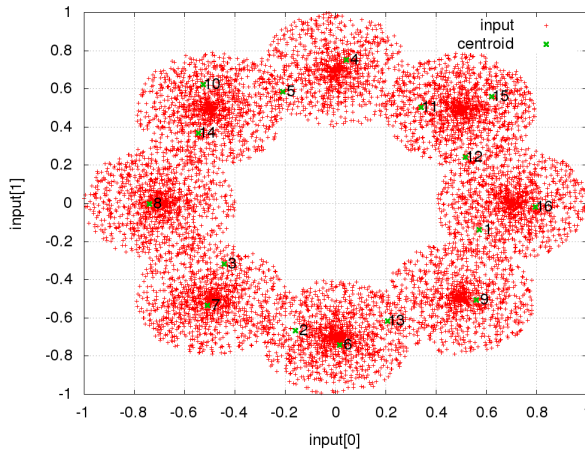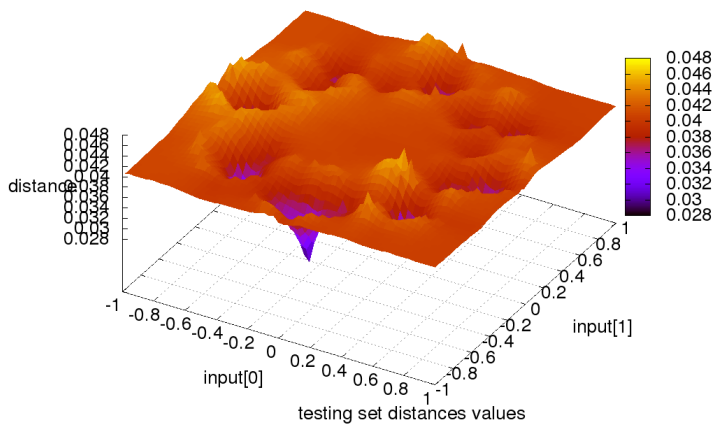

Figure 2. Kohonen test.

After found winning neuron, we need to find corresponding $v(n)$, mark this as $v_q(n)$.

$$v_q(n) = v_q(n-1) + 1 - ks_f(n) \tag{7}$$

This means robot speed $v(n)$ on curve type q is rising if $s_f(n)$ is low. Where $s_f(n)$ is low pass filtered $s(n)$ value.

3. Experimental results

Robot was learning few loops on short line race with different curves types. Resulting curves types are on figure 4. Each line represents one curve shape. To reduce space, absolute value of $|s(n)|$ as input into Kohonen neural network has been used. We can seen many straight line positions, and few curves.

Figure 3. Kohonen test distances.
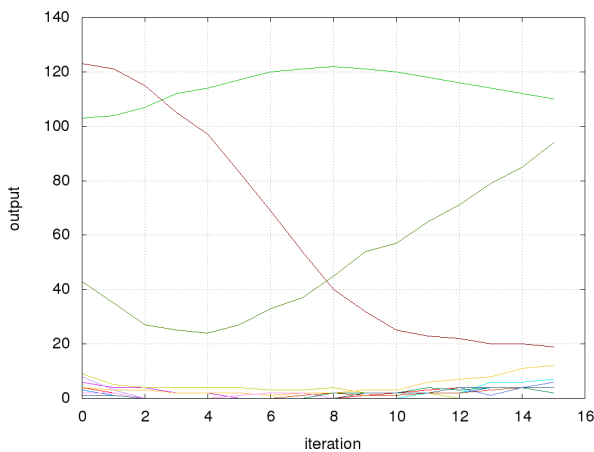
Figure 4. Resulting curves types.

4. Conclusions

In paper was describen learning system for curve shape recognition using Kohonen neural network. Neural network result have coresponding speed output which controll robot optional speed. Learning process is working in real time, on 75MHz ARM Cortex M4F microcontroller. Robot is on figure 5. Robot working video

can be shown on [6]. More pictures are availible on authors blog and sources on github [7].
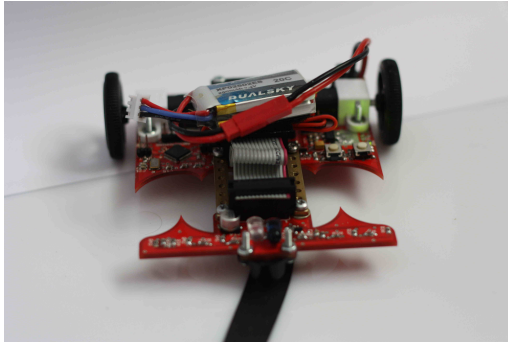
Figure 5. Testing robot.

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