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BOOK 5

"MATHEMATICS. INFORMATICS AND PHYSICS"

VOLUME 10

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This is the jubilee 10-th volume of book 5 Mathematics, Informatics and Physics. The beginning was in Spring, 2001, when the colleagues of the former section Mathematics and Physics decided to start publishing our own book of the Proceedings of the Union of Scientists – Ruse. The first volume included 24 papers. Through the years there have been authors not only from the Angel Kanchev University of Ruse but as well as from universities of Gabrovo, Varna, Veliko Tarnovo and abroad – Russia, Greece and USA.

Since the 6-th volume the preparation and publishing of the papers began to be done in English.

The new 10-th volume of book 5 Mathematics, Informatics and Physics includes papers in Mathematics, Informatics and Information Technologies, Physics and materials from the Scientific Conference 'Information Technologies in Education' (ITE), held at the University of Ruse in November 2012 in the frame of Project 2012-FNSE-02.

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APPLICATION OF MATLAB SOFTWARE FOR DIGITAL IMAGE PROCESSING

Milena Kostova, Ivan Georgiev

Angel Kanchev University of Ruse

Abstract: In this article an application of software product MATLAB in digital image processing using recursive discrete Kalman filter is presented. The potentiality of the product is analyzed examining the influence of the number of iterations in different disperse of the adaptive Gaussian noise upon the quality of the filtered image.

Key words: digital image, discrete recursive Kalman filter, Gaussian noise, iterations, MATLAB.

INTRODUCTION

The system MATLAB ("MATrix LABoratory") is a software environment for numerical analysis and independent fourth generation programming language. Created by the multinational software corporation for technical applications "The MathWorks", MATLAB allows operations using matrices, functions' drawing, data presenting, algorithm program implementation, development of human-machine interfaces and interfaces with different software products written in different programming languages [6]. One of the most important advantages of the system MATLAB is its ability to expand in purpose to be used as a tool for resolving new scientific tasks. This can be realized by creating a sequence of packages which can serve for expanding the common mathematical abilities of the system (symbolic calculations, optimization, statistics) as well as realization of the directions in computer mathematics like mathematical modeling, building an architecture of neural networks and fuzzy conclusions. MATLAB is used by more than 1 million academic and business users [6]. One possibility for applying the program environment MATLAB in digital image processing using discrete Kalman filter is presented.

DIGITAL IMAGE PROCESSING USING DISCRETE KALMAN FILTER

No matter the ways of obtaining, a given image is always accompanied with existence of noise and embarrassment of different kind. A main problem in its processing is removing the noise when keeping the important data for further details [7].

In most of the cases the most important goal of the image processing is to obtain a contour. This requires a sequence of procedures (filtration of Gaussian and impulse noise, segmentation aiming to receive black and white image, extraction of the contour line) which should be picked correctly of certain type and in determined sequence. An application of discrete Kalman filter in filtration of Gaussian at radiolocational images of dynamic objects (aircrafts) will be the case under review.

The Kalman filter is consistent recursive algorithm, using a model of dynamic system for receiving marks, which can be edited after every new measurement in temporal sequence [2,9]. This algorithm finds appliance in the process of managing complex dynamic processes, in which it is important to know the stage in each moment of time. The Kalman filter has a possibility to adjust in case of not precisely chosen parameters of the mathematical models. The advantage of the approach is that it does not require huge computational resources and can evaluate the condition vector in real time [5].

• Mathematical model of Discrete Kalman filter

The equations of the condition and observation are given:

$$x_k = A x_{k-1} + \omega_{k-1} \tag{1}$$

$$z_k = H x_k + \nu_k, \tag{2}$$

where $x_k \in \mathbb{R}^n$ is the state variable in *k*-th moment, and $z_k \in \mathbb{R}^m$ is the observation variable. Here *A* is $n \times n$ matrix, connecting the variables of the conditions of the current step *k* and the previous step *k*-1, and *H* is $m \times n$ -matrix connecting the observation and condition in the *k*-th step. The vectors w_k and v_k are normally distributed random variables with zero mathematical expectation and covariance matrixes:

$$Q_k = E \left[w_k w_k^T \right], \qquad R_k = E \left[v_k v_k^T \right], \tag{3}$$

where E is the mathematical expectation [8].

With $\hat{x}_k^- \in \mathbb{R}^n$ is marked the priori mark of the condition variable in the *k*-th step, and to $\hat{x}_k \in \mathbb{R}^n$ corresponds a posteriori evaluation in given observation z_k . Priori and posteriori errors are defined in the following way:

$$e_k^- \equiv x_k - \hat{x}_k^-, e_k \equiv x_k - \hat{x}_k \tag{4}$$

Then the priori covariance matrix of the error will look like:

$$P_k^- = E \left[e_k^- e_k^{-T} \right],$$
(5)
and the posterior covariance matrix of the error will be respectively:

$$P_k = E\left[e_k e_k^T\right]. \tag{6}$$

Posterior condition mark is a nonlinear combination of the priori assessment \hat{x}_k^- and the difference between provided measurement $H\hat{x}_k^-$ and the real measurement z_k , multiplied with $n \times m$ weight matrix K_k [4].

$$\hat{x}_{k} = \hat{x}_{k}^{-} + K_{k} \left(z_{k} - H \hat{x}_{k}^{-} \right).$$
(7)

The Kalman matrix K_k in (7) has the look:

$$K_{k} = P_{k}^{-} H^{T} \left(H P_{k}^{-} H^{T} + R_{k} \right)^{-1}.$$
(8)

Software realization of the model in MATLAB environment

The function which realizes the Kalman algorithm is the function s=kalmanf(s) [9]. Typical for this function is that it is executed for every dot of the image. It uses the primary data from function start_kal (Nt). In the beginning a validation check is executed to ensure the main parameters and corresponding assignments. Based upon the choice made concerning the number of filtering, Nt times filtering are executed. This is set as a parameter in function start_kal and is chosen from a menu. Appliance of Kalman filter for every row of the image starts, as it starts from the second column. The typical special feature here is that the input data initialized in the beginning are edited for each pixel from the image based on the previous filtered pixel and if the previous pixel is the first it is based on it. The filtration algorithm includes procedure for brightness segmentation (Otsu method) [3] and contour receiving (Roberts operator).

```
function varargout = kfig(varargin)
     % Visualization Function
     gui Singleton = 1;
     gui State = struct('gui Name', mfilename,
     if nargin & isstr(varargin{1})
      gui State.gui Callback = str2func(varargin{1});
     end
     if nargout
      [varargout{1:nargout}] = gui mainfcn(gui State, varargin{:});
     else
      gui mainfcn(gui State, varargin{:});
     end
     function kfig OpeningFcn(hObject, eventdata, handles, varargin)
     % Choose default command line output for kfig
     handles.output = hObject;
     % Update handles structure
     guidata(hObject, handles):
     % Executing actions when visualizing the object
     if strcmp(get(hObject,'Visible'),'off')
      I = imread('l1gaus.bmp','bmp'); % Load file (with matrix size 75x100) in I
      mshow(I);
     end
     % --- Outputs from this function are returned to the command line
     function varargout = kfig OutputFcn(hObject, eventdata, handles)
     varargout{1} = handles.output;
     % Start (pushbutton1) button Kalman filtering
     function pushbutton1 Callback(hObject, eventdata, handles)
                           % We create I global matrix
      global I;
     bw = get(handles.slider1,'Value') % Load from Slider in bw (value for black/white)
     popup sel index = get(handles.popupmenu1, 'Value'); % From the drop down menu
(n) times repetitions in k filter.
     switch popup sel index
      case 1 % 1 път
             start kal(1); % Function Start Kalman
             fig bw(bw); % Function towards black/white
      case 2 % 3 times
             start kal(3);
             fig bw(bw);
      case 3 % 5 times
             start kal(5);
             fig bw(bw);
      case 4
             start kal(10);
             fig bw(bw);
      case 5
             start kal(20);
             fig_bw(bw);
```

```
case 6
        start kal(30);
        fig bw(bw);
end
function start kal(Nt)
figure(1);
global I
lm=l
clear s
s.x = 1;
s.A = 1;
                     % Measurement
s.Q = 2^2;
s.H = 1;
s.R = 2^2;
s.B = 0;
s.u = 0;
s.x = nan;
s.P = nan;
for t2=1:Nt
for t1=1:346
for t=2:572
 tr(end+1)=I(t1,t);
 s(end).z=tr(end);
 s(end+1)=kalmanf(s(end));
                              % Filtration
 Im(t1,t)=s(end).x;
end
 clf;
 J = imresize(Im, 4);
 imshow(J);
end
I=Im;
end
global Im;
function fig bw(bw)
figure(2);
global Im;
bw_210 = Im > bw;
bw 210 = imresize(bw 210,4);
imshow(bw_210);title(['Prag B/W # ',int2str(bw)]);
global bw 210;
end
return
```

AN APPROACH FOR DETERMINING THE OPTIMAL NUMBER OF ITERATIONS FOR DIFFERENT DISPERSE OF GAUSSIAN NOISE IN FILTRATION WITH DISCRETE KALMAN FILTER

The optimal number of iterations in different disperse of Gaussian noise is determined by basing on quantity characteristics of two geometrical characteristics of the object (aircraft) - fuselage axis length and wingspan of the aircraft. The fuselage axis length and the wingspan are measured in number of pixels, respectively between the "nose" and "tail" of the aircraft and the end points of the two wings. The optimal number of iterations is determined by using the following criteria: the values of the chosen geometrical characteristics, received after image filtering to be most similar to the corresponding etalon aircraft before filtration. The change of the geometrical characteristics is examined in different number of iterations with discrete Kalman filter. Digital images from 10 types of aircrafts, excepted as etalon, have been used. They have been chosen by random pick of data bases [1]. They are: F16, An 124, Mc Donnell, B 52, Buccaneer, F117, Jaguar, Mig 29, Miraj 2000, Su34. A random image is chosen out of the choice. It is noised with adaptive Gaussian noise with normal distribution at zero mathematical expectation and disperse relatively 0,03; 0,05; 0,07. The results of the filtration are examined after 5, 10, 20 and 30 iterations. The results after filtration of etalon object are given respectively in Table 1, Table 2 and Table 3.

Geometrical	Number of iterations			
characteristics	5	10	20	30
Fuselage length	60	62	63	58
Wingspan	39	40	41	37

Table 1. Geometrical characteristics of an ob	ject, received after
filtration with Kalman filter-disprse (Ĵ,03

Table. 2.	Geometrical characteristics of an object, received after
	filtration with Kalman filter-disprse 0,05

Geometrical	Number of iterations			
characteristics.	5	10	20	30
Fuselage length	55	57	60	56
Wingspan	35	36	38	33

Table.3. Geometrical characteristics of an object, received after filtration with Kalman filter-disprse 0,07

Geometrical	Number of iterations			
characteristics	5	10	20	30
Fuselage length	40	42	42	38
Wingspan	28	29	30	27

The results are given graphically on *Fig. 1.* They show that the filter gives best results at 20 iterations and Gaussian noise disperse 0,03

For etalon object, presented in black color, the wingspan is 69 pixels and the fuselage axis length - 45 pixels.





Fig. 1. Dependence of the values of the geometrical characteristics upon the number of iterations in Discrete Kalman filter

Stages of the image processing wish Discrete Kalman filter at 0,03 and 20 iterations are visualized on *Fig.* 2. In Table 4 are given the geometrical characteristics of all the aircrafts, noised with adaptive Gaussian noise (disperse 0, 03) after filtrating with Kalman filter.

Table. 4 Aircrafts geometrical characteristics from the sample noised with
adaptive Gaussian noise with adaptive Gaussian noise (disperse 0, 03)
after filtration with Kalman filter (20 iterations)

Type of aircraft		Etalon	Etalon object		After filtration of Kalman Filter	
		Fuselage length	Wingspan	Fuselage length	Wingspan	
1	F16	50	33	47	29	
2	AN 124	91	100	89	90	
3	Mc Donnell	69	59	66	57	
4	Buccaneer	50	36	47	33	
5	F117	51	38	48	35	
6	Jaguar	80	45	75	41	
7	Mig29	43	38	39	32	
8	Miraj 2000	50	31	47	26	
9	Su34	87	61	84	55	
10	B52	69	45	68	44	

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Fig. 2. Stages of image processing with Discrete Kalman filter in disperse 0,03 and 20 iterations

As a criteria for accuracy is used the relative error [8]:

$$\hat{\delta}_{x} = \frac{\delta_{x}}{x_{u_{3M}}}.100, \qquad \% , \qquad (9)$$

where $\delta_x = |x_{em} - x_{u_{3M}}|$ - absolute error;

 x_{em} is the value of the parameter of the etalon object:

 $x_{\scriptscriptstyle \mathcal{U}\!\mathcal{S}\!\mathcal{M}}$ - value of the parameter's object after Kalman filter applying

Values of the relative error for the examined types of aircrafts, calculated on (9), are showen in Table 5.

Table.5 Values of the relative error for the studied types of aircraft

Type of	$\hat{\delta}_{x}$, %			
aircraft	Length of the fuselage axis	Wingspan		
F16	6.38	13.79		
AN 124	1.12	11.11		
Mc Donnell	4.55	3.5		
Buccaneer	6.38	9.09		
F117	6.25	8.57		
Jaguar	6.67	9.75		
Mig29	10.25	18.75		
Miraj 2000	6.66	19.23		
Su34	3.57	10.9		
B52	3.7	6.25		

For image filter using the Kalman method is created graphical user interface (GUI). The overall appearance is presented on *Fig. 3.*



Fig. 3. Overall appearance of GUI after filtration for digital image using the Kalman method

The interface gives the following possibility for choice:

- Number of repetitions for filtration;
- Grey threshold for receiving the binary image;
- > Object detecting in the image.

The input and the filtered image are visualized.

CONCLUSION

The presented development shows one application of program product MATLAB in software realization for created mathematical model, which gives possibility for visualizing the stages of image processing as well as examining the influence of iteration numbers of Discrete Kalman filter upon the quality of the filtered image.

For reasonable error usually is accepted the value $\delta_x = 5$ %. From the results we can see that the relative error when using Kalman filter reaches up to 10% for length of the fuselage axis and up to 19% for wingspan. The result gives rise for looking to another filter (eventually based on neural network), which would give better results (minimization the error), connected with the inputted geometrical characteristics.

This work is appropriate for classes, related to Digital image processing. It could be an interesting subject for graduate students studying Computer systems, Mathematics, Informatics, Radiolocation and navigations and so on.

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CONTACT ADDRESSES

Assoc. Prof. Milena Kostova, PhD Department of Mathematics Faculty of Natural Sciences and Education Angel Kanchev University of Ruse 8 Studentska Str., 7017 Ruse, Bulgaria Phone: (++359 82) 888 453 E-mail: mpk@ami.uni-ruse.bg

Pr. Assist. Ivan Georgiev Department of Applied Mathematics and Statistics Faculty of Public Health and Healthcare Angel Kanchev University of Ruse 8 Studentska Str., 7017 Ruse, Bulgaria Phone: (++359 82) 888 725 E-mail: <u>igeorgiev@uni-ruse.bg</u>

ПРИЛОЖЕНИЕ НА ПРОГРАМЕН ПРОДУКТ МАТLAВ ПРИ ОБРАБОТКА НА ЦИФРОВА ИНФОРМАЦИЯ

Милена Костова, Иван Георгиев

Русенски университет "Ангел Кънчев"

Абстракт: В статията е представено приложение на програмен продукт МАТLAВ при обработка на цифрово изображение чрез дискретен рекурсивен Калманов филтър. Анализирани са възможностите на продукта при изследване влиянието на броя итерации, при различна дисперсията на адитивния Гаусов шум, върху качеството на отфилтрираното изображение.

Ключови думи: цифрово изображение, дискретен рекурсивен Калманов филтър, Гаусов шум, итерации, MATLAB.

