

# Comparison of Principal Component Analysis (PCA), Residual Vector Quantization (RVQ), Exhaustive Search Vector Quantization (ESVQ), and Tree Structured Vector Quantization (TSVQ)

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## Contents

<b>1</b>	<b>Introduction</b>	<b>2</b>
<b>2</b>	<b>Experiments</b>	<b>2</b>
<b>3</b>	<b>Results</b>	<b>2</b>
<b>4</b>	<b>Conclusions</b>	<b>2</b>
<b>A</b>	<b>Notation</b>	<b>2</b>
<b>B</b>	<b>Source code</b>	<b>3</b>
B.1	Stand-alone applications for sanity checking. . . . .	4
B.2	Training data manipulation . . . . .	6
B.3	Training . . . . .	8
B.4	Testing . . . . .	10
B.5	Tracking algorithms . . . . .	11
B.6	Utility functions . . . . .	13
B.6.1	RVQ . . . . .	13
B.6.2	Signal metrics . . . . .	13
B.6.3	Indexing . . . . .	15
B.6.4	Plotting . . . . .	15
B.6.5	Files . . . . .	16

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## Abstract

# 1 Introduction

# 2 Experiments

# 3 Results

# 4 Conclusions

## A Notation

- $N$ . Number of data points.
- $D$ . Dimensionality of data points.
- **Design matrix**. A matrix containing input data. The terminology is used by Andrew Ng from Stanford in his Machine Learning course. In the design matrix (DM), each row contains one data observation, i.e., there is one data vector per row. This is a  $N \times D$  matrix. For one data vector per column, I use DM2 which is a  $D \times N$  matrix and is simply the transpose of DM.
- **cfn**. This stands for "complete filename."
- **posneg image**. A planar image with the following planes: R, G, B, 255-R, 255-G, 255-B. A plane consists of all pixels belonging to a particular channel. In other words, a posneg image contains all red pixels, followed by all green pixels, followed by all blue pixels, followed by the inverted channels. To view such an image, open it in the Irfanview free software. Make sure the image has a .raw extension.

## B Source code

## B.1 Stand-alone applications for sanity checking.

```
1 %% This file compares PCA, RVQ and TSVQ by generating its own simple data.
2 %
3 % It can be used without any outside data, except if useDataSet=2 is used.
4 %
5 % usage example
6 %
7 % A good way to check RVQ is to use the following settings:
8 % useDataSet      = 4    %this is 256 numbers going from 1:1:256 as mentioned in the IDDM paper.
9 % rvq_struct.maxP  = 8
10 % rvq_struct.M     = 2
11 % Then rvq_struct.PHI_r (the red channel of the codebooks) is
12 % 192.4961  64.4961 -32.0000  32.0000 -16.0000 16.0000 -8.0000  8.0000 -4.0000  4.0000 -2.0000  2.0000 -1.0000 \
→ 1.0000 -0.4961  0.5039
13 % The green and the blue channels are also the same.
14 % This behavior is correct, as confirmed by Dr Barnes.
15 % The first M=2 numbers are the scalar codevectors for stage 1,
16 % the second M=2 numbers are the scalar codevectors for stage 2, and so on.
17
18 % Copyright (C) Salman Aslam. All rights reserved.
19 % Data created      : April 20, 2011
20 % Date last modified : July 7, 2011.
21 %%
22
23 %
24 %INITIALIZATIONS
25 %
26
27 %matlab
28 clear;
29 clc;
30 close all;
31 format compact;
32
33 %data input (5 different datasets, pick one of them)
34 useDataSet      = 4; %change this to 1, 2, 3, 4 or 5
35 if (useDataSet==1) ...
36     [DM2, sw, sh] = DATAMATRIX_create_simple_DM2;
37 elseif (useDataSet==2)
38     load testS-DM2-small; %don't use if you don't have this file
39 elseif (useDataSet==3) DM2 = [4 6 8 10 20 22 24 26];sw=1;sh=1; %sw=1, sh=1 means scalar example
40 elseif (useDataSet==4) DM2 = 1:256; sw=1; sh=1; %scalar example mentioned in IDDM
41 elseif (useDataSet==5) a = rand(1089,2); DM2 = [a a a a];sw=33;sh=33; %this is bizarre, %i.e. has repeated data points
42
43 end
44 [D, N] = size(DM2); %dimensions of DM2
45
46 %algorithm parameters
47 %bpca
48 bpca_struct.tstprm-Neig-1x1 = 16;
49 bpca_struct.trgout_descriptors = [];
50 bpca_struct.tstout_descriptor = [];
51
52 %tsvq
53 tsvq_struct.P = 5; %number of stages
54 tsvq_struct.M = 2; %2 is for binary TSVQ
55 tsvq_struct.trgout_descriptors = [];
56 tsvq_struct.tstout_descriptor = [];
57
58 %rvq
59 rvq_struct.maxP = 8; %number of stages
60 rvq_struct.M = 2; %number of codevectors/stage
61 rvq_struct.targetSNR = 1000;
62 rvq_struct.sw = sw; %snippet width
63 rvq_struct.sh = sh; %snippet height
64 rvq_struct.dir_out = '';
65 rvq_struct.trgout_descriptors = [];
66 rvq_struct.tstout_descriptor = [];
67
68
69 %
70 % PROCESSING
71 %
72 %
73 %training
74 bpca_struct = bPCA_1_train(DM2, bpca_struct);
75 rvq_struct = RVQ_1_training(DM2, rvq_struct);
76 tsvq_struct = TSVQ_1_train(DM2, tsvq_struct);
77
78 %testing
79 tst_idx = 8; %any number between 1 and N, index
80 %of training data that you want to test
81 tst_Dx1 = DM2(:, tst_idx); %test vector
82 bpca_struct = bPCA_3_test(tst_Dx1, bpca_struct);
83 rvq_struct = RVQ_3_testing(tst_Dx1, rvq_struct);
84 tsvq_struct = TSVQ_3_test(tst_Dx1, tsvq_struct);
85
86 %
87 %RESULTS
88 %
89 %view
90 numDisplayRows = 5;
91 numDisplayCols = 10;
92 DATAMATRIX_display_DM2_as_image(DM2, sh, sw, numDisplayRows, \
→ numDisplayCols);
93 DATAMATRIX_display_DM2_as_image(bpca_struct.trgout_U-DxD, sh, sw, numDisplayRows, \
→ numDisplayCols);
94
95 %print
96 [DM2(:, tst_idx) bpca_struct.tstout_recon_Dx1 rvq_struct.tstout_recon_Dx1 tsvq_struct.tstout_recon_Dx1]
97
98 [bpca_struct.tstout_snr-1x1 rvq_struct.tstout_snr-1x1 tsvq_struct.tstout_snr-1x1]
99 [bpca_struct.tstout_rmse-1x1 rvq_struct.tstout_rmse-1x1 tsvq_struct.tstout_rmse-1x1]
```

Listing 1: main\_compare\_bPCA\_ESVQ\_RVQ\_TSVQ.m.

## B.2 Training data manipulation

```

1  %% This file generates a DxN matrix of test data.
2
3  % I call this a design matrix DM2 using Andrew Ng's words.
4  % A DM is NxN, DM2 is DxN.
5  % In DM, every D-dimensional vector is in one column.
6  % In DM2, every D-dimensional vector is in one row.
7  %
8  % Copyright (C) Salman Aslam. All rights reserved.
9  % Data created      : April 18, 2011
10 % Date last modified : July 7, 2011
11 %%
12
13 function [DM2, sw, sh] = DATAMATRIX_create_simple_DM2()
14
15     D                = 11;
16     X1                = [1 5 3 8 15 16 17 28 29 40 45]; %D dimensional data
17     X2                = X1+0.5*randn(1,D);
18     X3                = X1+0.7*randn(1,D);
19     X4                = X1+0.3*randn(1,D);
20     DM               = [X1; X2; X3; X4];               %create a total of 4 D-dimensional
21                                                         %data points clustered around X1
22
23
24     for i=1:5
25         X1                = X1+50*i;
26         X2                = X1+0.1*randn(1,D);          %row vector
27         X3                = X1+0.2*randn(1,D);
28         X4                = X1+0.3*randn(1,D);
29         DM               = [DM; X1; X2; X3; X4];        %4 data points centered around X1
30                                                         %4 data points centered around X1+50,
31                                                         %4 data points centered around X1+150,
32                                                         %4 data points centered around X1+300,
33                                                         %4 data points centered around X1+500,
34                                                         %4 data points centered around X1+750
35                                                         %24 total points
36     end
37
38     DM2                = DM';
39     sw                 = 1;          %we're saying the snippet width is 1
40     sh                 = D;          %snippet height is D, so this is a 1D snippet
41                                     %rather than say an image

```

Listing 2: DATAMATRIX\_create\_simple\_DM2.m.

```

1  %% This function writes input data (training vectors) to a file to be used for training an RVQ codebook.
2  %
3  % For all my algorithms, PCA, ESVQ, RVQ, TSVQ, the input data is formatted as a matrix called DM2.
4  % Refer to readme.pdf for a description of DM2.
5  %
6  % The training routines for PCA, ESVQ and TSVQ take the DM2 as input.
7  % However, RVQ requires its input to be saved as a file before it can be used for training by RVQ_1a_train_gen8.exe or \
8  % →RVQ_1a_train_gen16.exe.
9  % This function takes a DM2 and creates a file in the following manner: (a) 512 byte header, followed by (b) each posneg \
10 % →training vector.
11 %
12 % The file created by this function is what used to be called F1.sml in the original setup.
13 %
14 % Copyright (C) Salman Aslam. All rights reserved.
15 % Date created      : March 23, 2011
16 % Date last modified : July 9, 2011.
17 %%
18
19 function DATAMATRIX_saveInFormat_rvq(DM2, cfn_trainingFile, sw, sh) %cfn_trainingFile is the complete filename for the ouput \
20 →file                                                              %this file will contain a header and data in the DM2 \
21                                                                     →matrix
22                                                                     %sw is snippet width, sh is snippet height
23
24 %-----
25 %INITIALIZATIONS
26 %-----
27 Nc                = 6;          %number of channels in a training snippet, i.e., training vector
28 [D, N]            = size(DM2);  %N is the number of training vectors,
29 shc               = sh*Nc;      %number of rows in a single training snippet (vector)
30 headerSize        = 512;       %number of bytes in header
31 numBytes          = headerSize + sw*shc*N; %total bytes in entire file
32
33 %-----
34 %PRE-PROCESSING
35 %-----
36 % open the file
37 fid              = fopen(cfn_trainingFile, 'w'); %if this file exists, it is overwritten, fid is for file ID
38                 UTIL_FILE_checkFileOpen(fid, cfn_trainingFile);
39
40 %initialize header with all zeros
41 for i=1:headerSize
42     fwrite(fid, 0, 'char');
43 end
44
45 %-----
46 %PROCESSING
47 %-----
48 %write header
49 TSH              = 1;
50 UCH              = 1;
51 REAL             = 0;
52
53 fseek(fid, 0, 'bof');      fwrite(fid, TSH, 'int32');
54 fseek(fid, 4, 'bof');      fwrite(fid, 'abcdefg', 'char');
55 fseek(fid, 36, 'bof');     fwrite(fid, num2str(N), 'char');
56 fseek(fid, 44, 'bof');     fwrite(fid, sw, 'int32');

```

```

54     fseek(fid, 48, 'bof');      fwrite(fid, shc,      'int32');
55     fseek(fid, 52, 'bof');      fwrite(fid, UCH,      'int32');
56     fseek(fid, 56, 'bof');      fwrite(fid, REAL,    'int32');
57     fseek(fid, 60, 'bof');

59     for i=1:240+212
60         fwrite(fid, 204,      'uint8');
61     end

63 %write data
64 fseek(fid, 512, 'bof');
65 for i=1:N
66     I = DATAMATRIX_extract_ith_image_from_DM2(DM2, i, sw, sh);%this returns an image
67     Iposneg = RVQ__create_posnegImage(I, '', false, false); %this converts the image to posneg format
68     fwrite(fid, Iposneg); %write each posneg image one after the \
        →other
69 end

71 %-----
72 %POST-PROCESSING
73 %-----
74 %close file
75 fclose(fid);

```

Listing 3: "DATAMATRIX\_saveInFormat\_rvq.m."

```

1 %% This function assumes that each column of the input matrix (DM2) is an
2 %% image. It extracts a given column and returns it as an image.
3 %
4 % Copyright (C) Salman Aslam. All rights reserved.
5 % Date created : April 17, 2011
6 % Date last modified : July 9, 2011.
7 %%

9 function I_h_x_w = DATAMATRIX_extract_ith_image_from_DM2(DM2, col, w, h)

11     I_Dx1 = DM2(:, col); %extract D-dimensional column
12     I_h_x_w = reshape(I_Dx1, h, w); %reshape as an image and return

```

Listing 4: DATAMATRIX\_extract\_ith\_image\_from\_DM2.m.

```

1 %% Takes a DM2 matrix and displays it as an image.
2 %
3 % Copyright (C) Salman Aslam. All rights reserved.
4 % Date created : April 17, 2011
5 % Date last modified : July 9, 2011.
6 %%

8 function DATAMATRIX_display_DM2_as_image(DM2, h, w, numRows, numCols)

10     [D, N] = size(DM2); %N: number of training observations, D: dimensionality of data

12     for n = 1:N
13         col_vec = DM2(:, n);
14         img = reshape(col_vec, h, w);
15         UTIL_PLOT_tightsubplot(numRows, numCols, n, img);
16     end

```

Listing 5: DATAMATRIX\_display\_DM2\_as\_image.m.

```

1 %% Takes a DM2 matrix and displays it as an image.
2 %
3 % Assumption is that the data is grayscale.
4 %
5 % Here, the design matrix has observations in each row, that's standard
6 % however, each observation is an image which was vectorized by stacking
7 % columns onto each other, not row onto each other
8 %
9 % Copyright (C) Salman Aslam. All rights reserved.
10 % Date created : April 5, 2011
11 % Date last modified : July 9, 2011.
12 %%

14 function DATAMATRIX_display_DM_as_image_col(DM, h, w, numCols) %h and w are snippet height and width

16     [N, D] = size(DM); %N: number of training observations, D: dimensionality of data

18     for n = 1:N
19         row_vec = DM(n, :);
20         img = reshape(row_vec, h, w);
21         UTIL_PLOT_tightsubplot(numCols, n, img)
22     end

```

Listing 6: DATAMATRIX\_display\_DM\_as\_image\_col.m.

## B.3 Training

```
1 %% This function creates RVQ codevectors.
2 %
3 % This is the only function out of all my work that is closed source and does not belong to me.
4 % This is not to say that I have not incorporated other open source stuff in my work (always with appropriate permissions).
5 %
6 % Codebooks are denoted by PHI.
7 %
8 % Copyright (C) Salman Aslam. All rights reserved.
9 % Data created : April 17, 2011
10 % Date last modified : July 7, 2011
11 %%
12
13 function rvq_struct = RVQ_training(DM2, rvq_struct)
14
15 %-----
16 %INITIALIZATIONS
17 %-----
18 DM2_u8 = uint8(DM2); %design matrix, one observation per column
19 M = rvq_struct.M; %number of templates per stage
20 maxP = rvq_struct.maxP; %max number of stages
21 sw = rvq_struct.sw; %snippet width
22 sh = rvq_struct.sh; %snippet height
23 targetSNR = rvq_struct.targetSNR; %desired SNR
24 dir_out = rvq_struct.dir_out; %directory to store results in
25
26 %!! attention: these should be parameters but I'm fixing them !!
27 iFlag = 0.0005;
28 jFlag = 0.0005;
29
30 %filenames (original names in brackets in comments)
31 cfn_trainingFile = [dir_out 'positiveExamples.raw']; %file 1: vectorized positive examples, (F1.sml)
32 cfn_ecbk = [dir_out 'codebook.ecbk']; %file 2: encoder codebooks, (F1.ecbk)
33 cfn_dcbk = [dir_out 'codebook.dcbk']; %file 3: decoder codebooks, (F1.dcbk)
34 cfn_nodes = [dir_out 'codebook.nodes']; %file 4, linked list of training paths, (F1.nodes)
35 cfn_gentxt = [dir_out 'rvq_trg_verbose.txt']; %file 5, verbose output of gen.exe, (F1.stat.gen.txt)
36 cfn_bndintxt = [dir_out 'bnd.in.txt']; %file 6, verbose output of bnd.in.exe (F1.stat.bnd.in.txt)
37 cfn_trgsoc = [dir_out 'positiveExamples.idx']; %file 7, XDRs for training examples, (F1.idx)
38
39 %delete existing training files
40 UTIL_FILE_deleteFile(cfn_trainingFile);%file 1
41 UTIL_FILE_deleteFile(cfn_ecbk); %file 2
42 UTIL_FILE_deleteFile(cfn_dcbk); %file 3
43 UTIL_FILE_deleteFile(cfn_nodes); %file 4
44 UTIL_FILE_deleteFile(cfn_gentxt); %file 5
45 UTIL_FILE_deleteFile(cfn_bndintxt); %file 6
46 UTIL_FILE_deleteFile(cfn_trgsoc); %file 7
47
48 %-----
49 %PRE-PROCESSING
50 %-----
51 DATAMATRIX_saveInFormat_rvq (DM2_u8, cfn_trainingFile, sw, sh); %takes DM2 as input and writes it to a file
52
53 %-----
54 %PROCESSING
55 %-----
56
57 if (ispc)
58
59     if (maxP==8)
60         system(['RVQ_training_gen8.exe_' cfn_trainingFile '_' cfn_ecbk '_' cfn_dcbk '_' num2str(M+1) '_-S' num2str(\
61             →targetSNR) '_-i' num2str(iFlag) '_-j' num2str(jFlag) '_>-_' cfn_gentxt]);
62     elseif (maxP==16)
63         system(['RVQ_training_gen16.exe_' cfn_trainingFile '_' cfn_ecbk '_' cfn_dcbk '_' num2str(M+1) '_-S' num2str(\
64             →targetSNR) '_-i' num2str(iFlag) '_-j' num2str(jFlag) '_>-_' cfn_gentxt]);
65     end
66
67 elseif (isunix)
68
69     if (maxP==8)
70         system(['./RVQ_training_gen8.linux_' cfn_trainingFile '_' cfn_ecbk '_' cfn_dcbk '_' num2str(M+1) '_-S' num2str(\
71             →targetSNR) '_-i' num2str(iFlag) '_-j' num2str(jFlag) '_>-_' cfn_gentxt]);
72     elseif (maxP == 16)
73         system(['./RVQ_training_gen16.linux_' cfn_trainingFile '_' cfn_ecbk '_' cfn_dcbk '_' num2str(M+1) '_-S' num2str(\
74             →targetSNR) '_-i' num2str(iFlag) '_-j' num2str(jFlag) '_>-_' cfn_gentxt]);
75     end
76
77 end
78
79 %-----
80 %POST-PROCESSING
81 %-----
82 %read decoder codebook
83 [actualP, M_check, sw_check, sh_check, PHI_r, PHI_g, PHI_b, PHIn_r, PHIn_g, PHIn_b] = RVQ_read_codebook (\
84     →cfn_dcbk);
85
86 %error checking
87 if (M ~= M_check || sw ~= sw_check || sh ~= sh_check)
88     disp('ERROR: M, sw, or sh not correct')
89 end
90
91 %save to structure that will be passed back from function
92 rvq_struct.P = actualP;
93 rvq_struct.PHI_r = PHI_r;
94 rvq_struct.PHI_g = PHI_g;
95 rvq_struct.PHI_b = PHI_b;
96 rvq_struct.PHIn_r = PHIn_r;
97 rvq_struct.PHIn_g = PHIn_g;
98 rvq_struct.PHIn_b = PHIn_b;
99
100 %compute training SNR
101 %compute training SNR and training XDR descriptors using myExplorer
102 rvq_struct = UTIL_METRICS_compute_training_error(DM2, rvq_struct, 3);
```



```

100 %compute training XDR descriptors using gen -l
101 %rvq_struct.trgout_descriptors_genl= RVQ_3_check_soc_file(cfn_trgsoc, actualP, M, true);

103 %compare the 2 above
104 %diff = RVQ_compare_myExplorer_w_genl((rvq_struct.trgout_descriptors_genl)', rvq_struct.\x
    →trgout_descriptors);

106 %old code no longer used but here just in case
107 %system(['RVQ_0_create_posraw.exe', cfn_poscsv, num2str(sw), num2str(sh), cfn_trainingFile]);
108 %system(['./RVQ_0_create_posraw_linux', cfn_poscsv, num2str(sw), num2str(sh), cfn_trainingFile]);
109 %system(['RVQ_1b_train_bndin8.exe', cfn_trainingFile, cfn_ecbk, cfn_dcbk, num2str(M+1)
    →', '-S', num2str(targetSNR), -i, num2str(iFlag), -j, num2str(jFlag), >, cfn_bndintxt]);
110 %system(['RVQ_1a_train_gen8.exe', cfn_trainingFile, cfn_ecbk, cfn_dcbk, num2str(M+1)
    →', '-S', num2str(targetSNR), -l, >, 'temp.txt']);
111 %system(['./RVQ_1b_train_bndin8_linux', cfn_trainingFile, cfn_ecbk, cfn_dcbk, num2str(M+1)
    →', '-S', num2str(targetSNR), -i, num2str(iFlag), -j, num2str(jFlag), >, cfn_bndintxt]);
112 %system(['./RVQ_1a_train_gen8_linux', cfn_trainingFile, cfn_ecbk, cfn_dcbk, num2str(M+1)
    →', '-S', num2str(targetSNR), -l, >, 'temp.txt']);

```

Listing 7: "RVQ\_training.m"

## B.4 Testing

```

1 % Tests a test vector using an RVQ codebook.
2 %
3 % Copyright (C) Salman Aslam. All rights reserved.
4 % Date created : April 17, 2011.
5 % Date last modified : July 7, 2011.
6 %%
7 %Description:
8
9 %%
10 function rvq_struct = RVQ_testing(tst_Dx1, rvq_struct)
11
12     PHI_r = rvq_struct.PHI_r; %codebook, just the red channel
13     P = rvq_struct.P; %actual number of stages
14     M = rvq_struct.M; %number of codevectors/stage
15     sw = rvq_struct.sw; %snippet width
16     sh = rvq_struct.sh; %snippet height
17     D = sw*sh; %dimension of data
18
19     XDR = P + ones(P,1); %i initialize with P+1, the code for early termination
20     recon_Dx1 = zeros(D,1);
21
22     psnr_dB = 0;
23     PSNR_dB = [];
24     successiveRecon_DxP = zeros(1*D,P); %1st col contains 1st recon., 2nd column contains refined recon.,
25
26     err_Dx1 = tst_Dx1;
27     numStagesUsed = 0;
28
29     %go over all stages
30     for p=1:P
31         dmin = 1E15;
32
33         %for this stage: go over all codevectors
34         for m=1:M
35             idx = UTIL_xy_to_idx(m, p, M);
36             PHI_Dx1 = PHI_r(:,idx); %!!!caution!!! notice that only R \
37             %channel used
38             e = err_Dx1 - PHI_Dx1;
39             d = norm(e, 2); %if e is a matrix, this is largest eigenvalue, if it's a vector, \
40             %it's L2 norm
41             if (d < dmin)
42                 dmin = d;
43                 best_m = m;
44             end
45         end
46
47         %for this stage: temporarily store metrics
48         best_idx = UTIL_xy_to_idx(best_m, p, M);
49         best_PHI = PHI_r(:,best_idx);
50         temp_recon_Dx1 = recon_Dx1 + best_PHI;
51         temp_err_Dx1 = tst_Dx1 - temp_recon_Dx1;
52         temp_psnr_dB = UTIL_METRICS_compute_PSNRdB (255, temp_err_Dx1);
53         temp_snr_dB = UTIL_METRICS_compute_SNRdB (tst_Dx1, temp_err_Dx1);
54
55         %for this stage: decide if it will be "retained"
56         if (temp_psnr_dB < psnr_dB)
57             %discard
58             break
59         else
60             %keep
61             recon_Dx1 = temp_recon_Dx1;
62             successiveRecon_DxP(:,p) = temp_recon_Dx1;
63             err_Dx1 = temp_err_Dx1;
64             psnr_dB = temp_psnr_dB;
65             PSNR_dB(p) = temp_psnr_dB;
66             XDR(p) = best_m;
67             numStagesUsed = p;
68         end
69     end
70     %end: going over stages
71
72
73
74 %pass out
75 rvq_struct.tstout_recon_Dx1 = recon_Dx1;
76 rvq_struct.tstout_err_Dx1 = err_Dx1;
77 rvq_struct.tstout_descriptor = XDR;
78
79 rvq_struct.tstout_Nstages_used = numStagesUsed;
80
81 rvq_struct.tstout_snr_1x1 = UTIL_METRICS_compute_SNRdB (tst_Dx1, err_Dx1); %for PSNR, you only give \
82 %error signal
83 rvq_struct.tstout_rmse_1x1 = UTIL_METRICS_compute_rms_value (err_Dx1);
84 rvq_struct.tstout_psnr_1x1 = max(PSNR_dB);

```

Listing 8: "RVQ\_testing.m"

## B.5 Tracking algorithms

```

1 %% This function implements the particle filter (condensation algorithm).
2 %
3 % 0t1 means that the min value is 0, max value is 1.
4 %
5 % I_0t1 is the input image.
6 % f is the frame number.
7 %
8 % Copyright (C) Jongwoo Lim and David Ross (modified by Salman Aslam with permission)
9 % Date created      : April 25, 2011
10 % Date last modified: July 14, 2011
11 %%

13 function cond_struct = TRK_condensation(I_0t1, f, algo_struct, cond_struct, options_struct, RandomData_sample, RandomData_cdf, \
    → algo_code)
14 %
15 %INITIALIZATIONS
16 %
17     sw      = algo_struct.sw;           %snippet width
18     sh      = algo_struct.sh;           %snippet height
19     sz      = [sh sw];
20     D       = sw*sh;                    %dimensionality of input data

22     Np      = options_struct.Np;        %particle filter: # of particles (samples) from density)
23     rn1     = RandomData_cdf(f,:);      %pre-stored random numbers to ensure repeatability
24     rn2(:, :) = RandomData_sample(f, :, :); %same as above

26 %
27 %PRE-PROCESSING
28 %
29 %1. resample (in some cases it's done at the end, here it's done at the beginning, same thing really)

31     if ~isfield(cond_struct, 'affineCandidates_6xNp')
32         %one time: initialize 6 affine parameters, one for each of the Np candidate snippets
33         cond_struct.affineCandidates_6xNp = ...
34             → repmat(affparam2geom(cond_struct.tstout_bestAffineParams(:)), [1,Np]); %initialize \
35     else
36         %recurring: resample distribution in 3 lines (read details of this in my article on resampling)
37         cumconf = cumsum(cond_struct.weights);
38         idx     = floor(sum(repmat(rn1,[Np,1]) > repmat(cumconf,[1,Np])))+1;
39         cond_struct.affineCandidates_6xNp = ...
40             → cond_struct.affineCandidates_6xNp(:,idx); %keep only good candidates (resample)
41     end

43 %2. apply motion model (brownian, so just add randomness)
44     cond_struct.affineCandidates_6xNp = ...
45         → cond_struct.affineCandidates_6xNp + rn2.*repmat(options_struct.affsig(:),[1,Np]);

47 %extract the candidate snippets from the image based on motion model above
48     PFCandidateSnippets_0t1 = warping(I_0t1, affparam2mat(cond_struct.affineCandidates_6xNp), sz); %now create actual \
    → snippet candidates

50 %
51 %PROCESSING
52 %
53 %3a. weighting (find how well the algorithm model explains each snippet, find distances)
54     if (algo_code==1) %i.e. iPCA
55
56         diff_0t1 = repmat(algo_struct.mean(:),[1,Np]) - reshape(PFCandidateSnippets_0t1,[D,Np]); %diff_0t1: (sw)(\
    → sh) x Np
57         coeiffiff = 0;

59         if (size(algo_struct.basis,2) > 0)
60             coef = algo_struct.basis'*diff_0t1;
61             diff_0t1 = diff_0t1 - algo_struct.basis*coef;
62             if (isfield(cond_struct, 'coef'))
63                 coeiffiff = (abs(coef)-abs(cond_struct.coef))*algo_struct.reseig./repmat(algo_struct.eigval,[1,Np]);
64             else
65                 coeiffiff = coef ./ algo_struct.reseig ./ repmat(algo_struct.eigval,[1,Np]);
66             end
67             cond_struct.coef = coef;
68         end

73     elseif(algo_code==2) %i.e. bPCA
74         diff_0t1 = [];
75         for i = 1:Np
76             Itst = 255*PFCandidateSnippets_0t1(:, :, i);
77             algo_struct = bPCA_3_test(Itst(:), algo_struct);
78             diff_0t1(:, i) = algo_struct.tstout_err_Dx1/255;
79         end

83     elseif(algo_code==3) %i.e. RVQ
84         diff_0t1 = [];
85         for i = 1:Np
86             Itst = 255*PFCandidateSnippets_0t1(:, :, i);
87             algo_struct = RVQ_3_testing(Itst(:), algo_struct);
88             diff_0t1(:, i) = algo_struct.tstout_err_Dx1/255;
89         end

93     elseif(algo_code==4) %i.e. TSVQ
94         diff_0t1 = [];
95         for i = 1:Np
96             Itst = 255*PFCandidateSnippets_0t1(:, :, i);
97             algo_struct = TSVQ_3_test(Itst(:), algo_struct);
98             diff_0t1(:, i) = algo_struct.tstout_err_Dx1/255;
99         end

```

```

100     end
101
102 %3b. raise distances to exponentials
103 if (~isfield(options_struct,'errfunc'))
104     options_struct.errfunc ...
105         = [];
106     end
107
108     switch (options_struct.errfunc)
109     case 'robust';
110         cond_struct.weights ...
111             = exp(-sum(diff_0t1.^2./(diff_0t1.^2+options_struct.rsig.^2))./options_struct.condenssig)';
112     case 'ppca';
113         cond_struct.weights ...
114             = exp(-(sum(diff_0t1.^2) + sum(coefdiff.^2))./options_struct.condenssig)';
115     otherwise;
116         cond_struct.weights ...
117             = exp(-sum(diff_0t1.^2)./options_struct.condenssig)';
118     end
119
120 %4. pick MAP estimate
121 cond_struct.weights = cond_struct.weights ./ sum(cond_struct.weights); %normalize weights
122 [maxprob,maxidx] = max(cond_struct.weights); %MAP estimate: pick best index
123 cond_struct.tstout_bestAffineParams ...
124     = affparam2mat(cond_struct.affineCandidates_6xNp(:,maxidx)); %MAP estimate: pick best affine ↘
    %→parameters based on best index
125 cond_struct.tstout_bestPFcandidate_0t1 ...
126     = PFCandidateSnippets_0t1(:, :, maxidx); %MAP estimate: pick best candidate ↘
    %→ snippet based on best index
127
128 %-----
129 %POST-PROCESSING
130 %-----
131 %error and reconstruction
132 cond_struct.err_0t1 = reshape(diff_0t1(:,maxidx), sz); %get reconstruction error
133 if (algo_code==1)
134     cond_struct.err_0t1 = -cond_struct.err_0t1;
135 end
136 cond_struct.recon = cond_struct.tstout_bestPFcandidate_0t1 - cond_struct.err_0t1; %get reconstructed image
137
138 %metrics
139 cond_struct.tstout_snr = UTIL_METRICS_compute_SNR (cond_struct.tstout_bestPFcandidate_0t1, cond_struct.err_0t1);
140 cond_struct.tstout_rmse = UTIL_METRICS_compute_rms_value (cond_struct.err_0t1(:)*255);
141

```

Listing 9: TRK\_condensation.m.

## B.6 Utility functions

### B.6.1 RVQ

```
1 %% This function takes a grayscale or RGB image and creates a posneg image.
2 %
3 % Refer to readme.pdf for a description of posneg images.
4 %
5 % Copyright (C) Salman Aslam. All rights reserved.
6 % Date created : Feb 4, 2011
7 % Date last modified : July 9, 2011.
8 %%
9
10 function Iposneg = RVQ__create_posnegImage(I, cfn_Iraw, bView, bSave)
11
12 %-----
13 %INITIALIZATIONS
14 %-----
15     sz = size(I);
16
17     h = sz(1);
18     w = sz(2);
19     Nc = 6; %number of channels in raw image
20
21 %-----
22 %PRE-PROCESSING
23 %-----
24 %extract planar channels, if grayscale, make all channels the same
25 if (length(sz)==3) %RGB image
26     posR = I(:,:,1);
27     posG = I(:,:,2);
28     posB = I(:,:,3);
29 else if (length(sz)==2) %gray scale image
30     posR = I;
31     posG = posR;
32     posB = posR;
33 end
34
35 %vectorize (the reason is that we'll be stacking these in planar form)
36 posR = posR';
37 posG = posG';
38 posB = posB';
39
40 posR = posR(:);
41 posG = posG(:);
42 posB = posB(:);
43
44 %create negative channels
45 negR = 255-posR;
46 negG = 255-posG;
47 negB = 255-posB;
48
49 %-----
50 %PROCESSING
51 %-----
52 %create posneg image
53 Iposneg = [posR; posG; posB; negR; negG; negB];
54
55 %-----
56 %POST-PROCESSING
57 %-----
58 %save it
59 if (bSave)
60     fid = fopen(cfn_Iraw, 'w');
61     fwrite(fid, Iposneg);
62     fclose(fid);
63 end
64
65 %view it
66 if (bView)
67     M = reshape(Iposneg, w, h*Nc);
68     imshow(M);
69     title(cfn_Iraw);
70     pause
71 end
```

Listing 10: "RVQ\_\_create\_posnegImage.m."

### B.6.2 Signal metrics

```
1 %% This function computes the power signal of an input signal.
2 %
3 % Sig_Dx1: signal vector
4 % Pow_Dx1: power vector (normalized since we assume resistance R=1)
5
6 % Copyright (C) Salman Aslam. All rights reserved.
7 % Date created : April 15, 2011
8 % Date last modified : July 7, 2011
9 %%
10
11 function Pow_Dx1 = UTIL_METRICS_compute_powerSignal(Sig_Dx1)
12
13     Pow_Dx1 = Sig_Dx1.^2;
```

Listing 11: UTIL\_METRICS\_compute\_powerSignal.m.

```
1 %% This function computes the total energy of an input signal.
2 %
```

```

3 % Sig_Dx1: signal vector
4 %
5 % Copyright (C) Salman Aslam. All rights reserved.
6 % Data created : April 15, 2011
7 % Date last modified : July 7, 2011
8 %%
9 function energy = UTIL_METRICS_compute_energy(Sig_Dx1)
11
    energy = sum ( UTIL_METRICS_compute_powerSignal(Sig_Dx1) );

```

Listing 12: UTIL\_METRICS\_compute\_energy.m.

```

1 %% This function computes the power of a signal.
2 %
3 % Sig_Dx1: signal vector
4 %
5 % Copyright (C) Salman Aslam. All rights reserved.
6 % Data created : March 17, 2011
7 % Date last modified : July 7, 2011
8 %%
9 function power = UTIL_METRICS_compute_power(Sig_Dx1)
11
    power = UTIL_METRICS_compute_energy(Sig_Dx1) / length(Sig_Dx1); %norm(s,2) is the same as sqrt(sum(\
    →s.^2)), or square root of energy

```

Listing 13: UTIL\_METRICS\_compute\_power.m.

```

1 %% This function computes the rms energy of a signal.
2 %
3 % Sig_Dx1: signal vector
4 %
5 % Copyright (C) Salman Aslam. All rights reserved.
6 % Data created : March 17, 2011
7 % Date last modified : July 7, 2011
8 %%
10 function rms_value = UTIL_METRICS_compute_rms_value(Sig_Dx1)
12
    rms_value = sqrt( UTIL_METRICS_compute_power(Sig_Dx1) );

```

Listing 14: UTIL\_METRICS\_compute\_rms\_value.m.

```

1 %% This function computes SNR given a signal and an error signal.
2 %
3 % Sig_Dx1: signal vector
4 % Err_Dx1: error (noise) vector
5 %
6 % Copyright (C) Salman Aslam. All rights reserved.
7 % Data created : March 17, 2011
8 % Date last modified : July 7, 2011
9 %%
11 function SNR = UTIL_METRICS_compute_SNR(Sig_Dx1, Err_Dx1)
13
    %signal
14     power_signal = UTIL_METRICS_compute_power(Sig_Dx1); %Sig_Dx1 can also be a row vector
16
    %error
17     power_noise = UTIL_METRICS_compute_power(Err_Dx1); %Err_Dx1 can also be a row vector
19
    %SNR
20     SNR = power_signal / power_noise;

```

Listing 15: UTIL\_METRICS\_compute\_SNR.m.

```

1 %% This function computes SNR in dB given a signal and an error signal.
2 %
3 % Copyright (C) Salman Aslam. All rights reserved.
4 % Data created : March 20, 2011
5 % Date last modified : July 7, 2011
6 %%
8 function SNRdB = UTIL_METRICS_compute_SNRdB(Sig_Dx1, Err_Dx1) %although shown as column vectors, they can be row vectors as \
    →well
10
    SNR = UTIL_METRICS_compute_SNR(Sig_Dx1, Err_Dx1);
11     SNRdB = 10*log10(SNR);

```

Listing 16: UTIL\_METRICS\_compute\_SNRdB.m.

```

1 %% This function computes the PSNR of an input signal.
2 %
3 % Err_Dx1: error (noise) vector
4 %
5 % Copyright (C) Salman Aslam. All rights reserved.
6 % Data created : March 20, 2011
7 % Date last modified : July 7, 2011
8 %%
9 function psnr_dB = UTIL_METRICS_compute_PSNRdB(max_signal_value, Err_Dx1)
11
    power_noise = UTIL_METRICS_compute_power(Err_Dx1);
12     psnr_dB = 10*log10(max_signal_value^2 / power_noise);

```

Listing 17: UTIL\_METRICS\_compute\_PSNRdB.m.

```

1 %% This function computes training error, i.e., loss
2 %
3 % The training error computed is for DM2, i.e., all the training data.
4 %
5 % For computing SNRdB, the following approach is taken:
6 % (a) for each training vector, compute the error vector algo_struct.tstout_err_Dx1
7 % (b) concatenate all training vectors into S_NDx1 to make one giant signal
8 % (c) concatenate all error vectors into E_NDx1 to make one giant error signal
9 % (d) compute SNRdB using these giant signals
10 %
11 % The reason for this approach is that it's similar to the approach taken
12 % in Explorer, Dr Barnes' software.
13 %
14 % Copyright (C) Salman Aslam. All rights reserved.
15 % Data created : April 20, 2011
16 % Date last modified : July 7, 2011
17 %%

19 function algo_struct = UTIL_METRICS_compute_training_error(DM2, algo_struct, algo_code)

21 %=====
22 %INITIALIZATIONS
23 %=====
24 [D, N] = size(DM2);
25 S_NDx1 = []; %all N D-dimensional training vectors are concatenated to form a large signal
26 E_NDx1 = []; %all N D-dimensional error vectors are concatenated to form a large signal
27 algo_struct.trgout_descriptors = []; %the above step is carried out to be compatible with the way Explorer computes \
    → training error

29 %=====
30 %PRE-PROCESSING
31 %=====
32 %!!caution, this needs checking
33 if (algo_code==1) str='algo_struct=====bPCA_3_test===== (tst_vec_Dx1, _algo_struct);'; %probabilistic \
    →PCA
34 %!!end caution
35 elseif (algo_code==2) str='algo_struct=====bPCA_3_test===== (tst_vec_Dx1, _algo_struct);'; %batch PCA
36 elseif (algo_code==3) str='algo_struct=====RVQ_testing===== (tst_vec_Dx1, _algo_struct);'; %RVQ
37 elseif (algo_code==4) str='algo_struct=====TSVQ_3_test===== (tst_vec_Dx1, _algo_struct);'; %TSVQ
38 end

40 %=====
41 %PROCESSING
42 %=====
43 for i=1:N
44
45     %pick out a single vector from DM2, this is one training example
46     tst_vec_Dx1 = DM2(:,i);
47
48     %evaluate error against model
49     eval(str); %so for RVQ, test against the RVQ codebook
50
51     %concatenate signals (training vectors) and errors
52     S_NDx1 = [S_NDx1; tst_vec_Dx1]; %save the SNR
53     E_NDx1 = [E_NDx1; algo_struct.tstout_err_Dx1]; %save the error
54 end
55
56 %compute SNR for the large signal
57 snr_dB = UTIL_METRICS_compute_SNRdB (S_NDx1, E_NDx1); %S_NDx1 is now one big signal
58 rmse = UTIL_METRICS_compute_rms_value (E_NDx1);

60 %=====
61 %POST-PROCESSING
62 %=====
63 %save results
64 algo_struct.trgout_snr = snr_dB;
65 algo_struct.trgout_rmse = rmse;

```

Listing 18: UTIL\_METRICS\_compute\_training\_error.m.

## B.6.3 Indexing

```

1 %Copyright (C) Salman Aslam. All rights reserved.
2 %Date: July 7, 2011.
3 %%
4 %Description:
5 %- This file takes an index and image width (Iw) and returns (x,y) coordinates.
6 %%

8 function [y,x] = UTIL_idx_to_xy(idx, Iw)

10 %go to C style indexing (0-based)
11 idx = idx-1;

13 %compute x, y in C-style
14 y = idivide( int32(idx), int32(Iw), 'floor' );
15 x = mod(int32(idx), int32(Iw));

17 %back to Matlab style (1-based)
18 y = y+1;
19 x = x+1;

```

Listing 19: UTIL\_idx\_to\_xy.m.

## B.6.4 Plotting

```

1 %Downloaded from web (don't know from where) and modified. All rights reserved.
2 %Date: July 7, 2011.
3 %%
4 %Description:
5 %This file displays several images in one figure.
6 %%
7
8 function UTIL_PLOT_tightsubplot(numRows, numCols, idx, I_rgb)
9
10     colormap('gray');
11
12     [row, col] = UTIL_idx_to_xy(idx, numCols);
13     row = double(row-1);
14     col = double(col-1);
15     x = col*(1/numCols);
16     y = (numRows-row-1)*(1/numRows);
17
18
19     w = 1/numCols-.001;
20     h = 1/numRows-.001;
21     subplot('position', [x, y, w, h]);
22
23     imagesc(I_rgb);
24     axis equal
25     axis off;

```

Listing 20: UTIL\_IMG\_tightsubplot.m.

## B.6.5 Files

```

1 %% This function throws an exception if the fileID is invalid, i.e. -1
2
3 % Copyright (C) Salman Aslam. All rights reserved.
4 % Date created : Feb 19, 2011
5 % Date last modified : July 9, 2011.
6 %%
7
8 function UTIL_FILE_checkFileOpen(fid, cfn)
9
10     try
11         if (fid == -1)
12             err = MException('ResultChk:FileNotOpen', ['salman:_ ' cfn '_could_not_be_opened']);
13             throw(err);
14         end
15
16     catch err
17         disp(err.message);
18         rethrow(err);
19     end

```

Listing 21: UTIL\_FILE\_checkFileOpen.m

```

1 function UTIL_FILE_deleteFile(cfn) %cfn is complete file name
2
3     if (ispc) [status, result] = system(['del_' cfn]);
4     elseif (isunix) [status, result] = unix(['rm_' cfn]);
5     end

```

Listing 22: UTIL\_FILE\_deleteFile.m

## References