Sehbau: Descriptor Extraction

The use of the program for descriptor extraction is explained, called dscx. It extracts the features and describes them geometrically by a few parameters, resulting in a very versatile representation that can be used for identification and categorization of any image size and any structure. The program outputs a number of files containing the following information: the bounding boxes for the regions; the keypoints of contour segments; the descriptor vectors; the histograms of attributes (for fast classification); and files containing boundary information.

Read https://www.researchgate.net/publication/360033329

Repository https://github.com/Sehbau/DescExtraction

NEXT matching vectors, https://github.com/Sehbau/MtchVec

DOCU **overview:** https://github.com/Sehbau/Docu/blob/main/overview.pdf

The directories in the repository serve for immediate probing with commands as shown in the upcoming sections:

- /Desc example of what the output should look like
- -/Imgs sample images for immediate testing
- /UtilMb Matlab scripts to read the output data files

Three types of binaries are available, all 64 bit:

- dscx.exe, Windows 10 [x64]
- dscx_ubu, Ubuntu 20.04.3 LTS (Focal Fossa) [x64]
- dscx_deb, Debian GNU/Linux 11 (bullseye) [ARM]

If the linux binaries fail on your release, then try using 'wine' (windows emulator) and running the windows executable. That would be the easiest for me at the moment. Or let me know with what options I should compile the code (info@sehbau.com).

The use of the program is explained first, and the options that are available for it (Section 1). Then we explain the type of data files that are generated and their format (Section 2).

1 Program Use

Two arguments are required, the image path and the output path for the data files:

> dscx pathImg pathOutFile

The input image can be jpg or png (or anything offered by https://github.com/nothings/stb). The output path must include a slash (as the program checks for that). Here is an example,

```
> dscx Imgs/img1.jpg Desc/img1
```

in which the output file name img1 is chosen to be the same as the image name, for convenience. This will then write the following files into directory Desc:

- img1.Bbox bounding boxes of regions (connected components)
- imgl.CntEpt endpoints of ridge, river and edge segments
- imgl.vec descriptor vectors (attributes)
- img1.hst descriptor histograms
- img1.BonBbox bounding boxes concatenated across depth
- img1.BonAsp aspects of boundaries
- img1.BonPix boundary pixels

The directory /UtilMb contains example scripts for Matlab on how to read the data files. It also contains an example script for contour pixel detection to demonstrate the technique, called e_CntMap.m. An example for segmentation is given in the repository called 'DemoTree'.

There are no particular image preprocessing algorithms carried out in this program as I have never observed any consistent advantage of using for instance smoothing or low-pass filtering. If you think that your image collection would profit from preprocessing, then carry it out separately (beforehand) and then feed the image to dscx.

This is a huge and novel framework. It might fail with unusual images. If you find one, inform me on <code>info@sehbau.com</code>. It should take sizes up to [3000 x 4000], such as a cell phone photo, but have not tried larger images yet. The program output should end with the term <code>EndOfProgram</code>. It might also fail if you choose unreasonable parameter values, as I have not included reasonability checks everywhere.

1.1 Options

The following long options are available, to be specified with a double dash '--'; single letter options are not in use. The first set allows to adjust architecture parameters, the second set contour parameters, the third set region parameters, the fourth set partition parameters and eventually some utility parameters.

Many of the parameters will mainly regulate the number of descriptor vectors saved to .vec. For histograms, typically a fuller set of descriptors is taken, as the exact choice of those parameters has lesser influence on recognition accuracy. For manipulation with vectors however, some of these parameters can make a huge difference. In particular for place recognition I had observed substantial variations, but a full systematic search is still to be carried out.

The four descriptors are abbreviated as:

- cnt: contours
- rsg: radial signature
- arc: arc segment (from boundary partitioning)
- cnt: straighter segment (from bound. partitioning)

See also the overview document for more introduction on variables:

https://github.com/Sehbau/Docu/blob/main/overview.pdf

1.1.1 Architecture:

--depth: depth of the segmentation process. Default depth=3. For depth=1 no tree is grown: this corresponds to global thresholding only (with a single threshold). depth=4 can be useful for large images, e.g. larger than 1000 pixel for one image side.

--nLev: number of levels of the pyramid (with downsampling factor equal 2). Default is automatically calculated with top level not smaller than 16 pixels for one map side. For example for a 256x256 image, a five-level pyramid is generated: 256 (original resolution), 128, 64, 32 and 16.

1.1.2 Contours:

--cntMinCtr: contrast threshold for contours. Default =0.05. This is the proportion, of the largest difference found in the range map of the gray-scale intensity image (for a 3x3 neighborhood).

The following two parameters - starting with <code>skl</code> - modify the output of the contour selection, the skeleton. This concerns only the vectors - the histograms are generated with all contours without subselection. To understand those changes see plot <code>ImgPyrSkel.png</code> [run with <code>plot</code>] or used the example <code>script</code> exampleLoadVect.m.

--sklMinSpc: minimum spacing. Default = 0.05. This is as proportion of the image side length $\in [0.0.1.0]$. Changes here can have a huge effect on performance, recognition accuracy in particular.

--sklMinLen: minimum length. Default =0.05. This as well is as proportion of the image side length $\in [0.0.1.0]$. Changes here are less significant, in particular for large spacing values (set with **sklMinSpc**), as then only few short segments remain.

1.1.3 Regions:

--regMinPixNode: minimum number of pixels for a region to be segregated by the thresholding mechanism. This will affect the region count from the 2nd segmentation map on. It will not affect the 1st segmentation map, as that is the input to the 2nd for which segregation starts.

Default equal 6. With larger values, processing occurs more rapidly, but may not capture texture properly anymore.

The following two parameters - starting with ${\tt rsg}$ - concern the creation of the radial descriptor.

--rsgMinPix: minimum number of boundary pixels for a radial region descriptor. The number is set for the original image resolution. For higher levels of the pyramid, a correspondingly lower number is used, namely rsgMinPix-level. E.g. for a value of 10, the higher pyramidal levels utilize values 9, 8, 7,....

Default equal 4 for all levels. For larger values, you risk loosing texture, thus if you interested in the global structure only, it can be useful to set higher values.

--rsgMinCtr: mininum boundary contrast for a radial region descriptor. The value is relative to the average contrast value for all extracted boundaries. Boundaries below that contrast value will not be described as radial descriptor.

1.1.4 Partitions (Arcs/Straighters):

Parameter names starting with **cvp** regulate the partitioning process and therefore affect the outcome of both arc and straighter partitions. Parameter names starting with **arc** and **str** are specific to the respective descriptors.

--cvpMinSiz: minimum boundary size entering the boundary partitioning process. This is relative to the larger image side. For example a value of 0.02 for a [1024 x 2048] image will set the minimum size to 41 pixels. Default equal 0.05.

For small values this will seemingly regulate the number of straighter segments only, as those are rarer. For higher values, it will also start omitting arc segments.

--cvpInclBord [flag]: includes partitions at borders. The default is OFF, meaning that partitions at image borders are ignored. Turning them ON (by listing this option) can improve place recognition significantly.

--cvpMinCtr: minimum boundary contrast entering the boundary partitioning process. This is a proportion of the largest boundary contrast determined. Default equal 0.05.

--arcMinLen: minimum arc length to be described. This is relative to image side length. This is a better way of directly regulating the number of arcs than **cvpMinSiz**. Default equal 0.08.

--strMinGer: minimum straightness value for a straighter segment to be accepted. This is a fixed threshold $\in [0..1]$ based on the measure chord length divided by segment arc length. Default equal 0.8.

The following options set parameters for selecting the skeleton of partition segments, called here 'Gerust', abbreviated <code>gst/Gst</code>. Reducing the number of segments for finer scales often improves recognition - and also reduces matching duration. But I suspect that for smaller images (or higher levels of the pyramid; ie. smaller 100 pixels per side), that the reduction might have less effect.

——arcGstMinSmo: minimum smoothness for arcs set for entire pyramid. Default approximately 0.20. This parameter is intended to eliminate segments (resulting from segregation of luminance gradients) that typically show high irregularity in their spatial course. Images of scenes taken at night tend to have those in particular.

—arcGstMinSpc: minimum spacing for arcs set for entire pyramid. Default approximately 0.05.

--arcGstOff: turns off any selection by setting all parameters values to zero (for smoothness, spacing and length). This then takes the full set of extracted arcs.

--strGstMinSpc: minimum spacing for straighters set for entire pyramid. Default approximately 0.05.

--strGstOff: turns off any selection by setting all values to zero. This then takes the full set of extracted straighters.

1.1.5 Utility:

--prms [flag]: displays the parameter values used. Default OFF.

--noBbox [flag]: turns off saving of bounding boxes (.Bbox file) and contour endpoints (.CntEpt). Default ON.

--noBon [flag]: turns off saving of boundary information (files .BonBbox, .BonAsp and .BonPix). Default ON.

--plot [flag]: plots contours and region boundaries for the entire pyramid. Default OFF. The following images will be written:

- -Icnt.png: contours plotted onto the color image for the original resolution. This serves to observe the matching between contour type (ridge, river, edge) and the image.
- -ImgPyrBonOnly.png: boundaries of the entire pyramid and for different depths.
- -ImgPyrCntOnly.png: contours of the entire pyramid and for different levels.
- -ImgPyrSkel.png: the selected contours (skeleton).

--verbose [flag]: for illustration or for tracking errors. Default OFF.

2 Output Data Files

The following five Matlab scripts load the data files using various function scripts:

- -/UtilMb/exampleLoadBbox.m, function scripts in /Bbox
- -/UtilMb/exampleLoadBon.m, function scripts in /Bound

```
-/UtilMb/exampleLoadBonPix.m, function scripts in /Bound
-/UtilMb/exampleLoadHist.m, function scripts in /Hist
-/UtilMb/exampleLoadVect.m, function scripts in /Vect
```

The script exampleLoadBbox.m also loads the contour endpoints. The following sections give some more explanations.

2.1 Bounding Boxes (.Bbox)

The bounding boxes for regions are saved in text format to a file with extension . Bbox. The list contains all bounding boxes. This allows you to subselect according to your tasks.

The information in this file is minimal. More information is provided in the .BonXXX files (upcoming in Section 2.5). But here we introduce the organization of the bounding boxes as well as the border values.

The first two integer values of the file hold the number of levels <code>nLev</code> and segmentation depth depth applied in the run. The following numbers hold the region count for each segmentation map, <code>nBbox</code>, saved looping levels as the outer loop and looping depth as the inner loop. The example below shows that for two levels and depth equal 3 (using zero-indexing).

```
nLev
depth
nBbox_Lev0_Depth0
nBbox_Lev0_Depth1
nBbox_Lev0_Depth2
nBbox_Lev1_Depth0
nBbox_Lev1_Depth1
nBbox_Lev1_Depth1
```

Then the bounding boxes follow. They are organized analogously to the above inner/outer loop: first all bounding boxes of [lev=0,depth=0], then those of [lev=0,depth=1], etc. A bounding box contains 6 parameters.

```
top, bottom, left, right, area, border
top, bottom, left, right, area, border
...
top, bottom, left, right, area, border
(for lev=0, depth=0)
(for lev=0, depth=0)
(for lev=1, depth=2)
```

The parameters describe:

-top, bottom, left, right: absolute coordinates that correspond to the map size of the pyramidal level. Thus you need to upsample them by multiplying with the corresponding factor (2, 4, 8, ...).

-area: size of bounding box, calculated with the first four parameters.

-border: number of touches with the four image sides. The values are:

```
0 no touches: off border

1-4 at one border, directions NESW (top, rite, bot, left)

11-14 at two borders, directions NE,ES,SW,NW (topright, ...)

15,16 " " , NS axis, WE axis

101-3 at three borders

200 touching all borders
```

The bounding box sizes are typically slightly too small in comparison to annotations in datasets, partly due to the segmentation procedure and partly due to downsampling. Adding margins achieves better annotation correspondence, ie. margin values that correspond to the pyramid level.

Of course you can perform better selections with more information such as boundary contrast and perhaps region attributes, which is included in the .BonXXX files, upcoming in Section 2.5.

Selecting the inside regions (0 touches / off border) returns the objects in a scene - if there are any. In landscape scenes (without any large objects in focus), such inside regions can be rare, as the scene parts often touch one or more image sides. Small objects rather appear as ridge or river contours.

2.2 Contour Endpoints (.CntEpt)

The points for the contour segments consist of the two endpoints as well as their midpoint. The points are written per level, per contour type and per point type. They are saved in binary format with extension .CntEpt.

The first value holds the number of levels. Then each level of the pyramid is written separately with firstly the points of the ridge contours, then those of the river contours and eventually those of the edge contours. The points are written blockwise (and not rowwise as in case of the bounding boxes). The first value holds the number of descriptors. Then follow first all coordinates of the first endpoint (for that level); then all coordinates of the second endpoint; followed by all coordinates for the midpoint. The coordinates coords are saved as row/column pairs, per point.

```
nLev
nRdg (# of ridge contours for lev=0)
[ridge coords of 1st endpoint for lev=0]
[ridge coords of 2nd endpoint for lev=0]
[ridge coords of midpoint for lev=0]
nRiv (# of river contours for lev=0)
[river coords of 1st endpoint for lev=0]
[river coords of 2nd endpoint for lev=0]
[river coords of midpoint
nEdg (# of edge contours for lev=0)
[edge coords of 1st endpoint for lev=0]
[edge coords of 2nd endpoint for lev=0]
[edge coords of midpoint
                            for lev=0]
nRdg (# of ridge contours for lev=1)
[ridge coords of 1st endpoint for lev=1]
[ridge coords of 2nd endpoint for lev=1]
[ridge coords of midpoint
                             for lev=11
```

As with bounding boxes, the segment coordinates are absolute values corresponding to the map size of the pyramidal level. They need to be upsampled to match the original image resolution if they are used as object/part proposals, and given some spatial width by adding some corresponding value.

2.3 Histograms (.hst)

The histograms are saved in binary format with extension .hst. They are organized according to increasing dimensionality and spatial order:

- flat, univariate: one-dimensional for the entire image
- flat, bivariate: two-dimensional for the entire image
- spatial, univariate: taken from a 3x3 grid, one-dimensional (univariate)
- spatial, bivariate: taken from a 3x3 grid, two-dimensional (bivariate)

For each type, the order of the descriptor types is: contour, radial signature, arc segment and straighter segment. The reading routine for each of those is in directory /UtilMb/Hist.

The radial signature was saved with a number of parameters not described in the (latest) paper (cnev to bis4). They only slightly improve classification results however.

We recommend starting classification with the first two types, the flat univariate and bivariate histograms. The third type improves accuracy in particular for whole images. The last (fourth) type of histogram, the spatial bivariate, did not consistently improve classification accuracy and can be omitted as it also increases the total dimensionality quite a bit.

2.4 Vectors (.vec)

The vector file with extension . **vec** contains the descriptor attributes for all four types (contour, radial signature, arc and straighter segment). It can be loaded as shown in script <code>exampleLoadVect.m</code>. The values are organized per attribute (dimension), not per (descriptor) vector. Thus, for the purpose of clustering or classification in Matlab (or Python), one has to concatenate them horizontally to a <code>[nDsc x nAtt]</code> matrix.

Most attribute values are normalized to unit range, some only to approximate unit range in case of complex attribute definitions. Angle values

come in radians. Depending on the exact type of use of the vectors, one should certainly consider scaling. The script ReadCntAtt contains the most comments. Some more explanations:

- PosV, PosH: vertical and horizontal position. Typically the center of the feature, ie. the midpoints of contours and boundary segments; the pole for radial descriptors.
- Tif: holds range information ("Tiefe"), as measured by a depth camera. Not (fully) operational yet.
- OrgCrv: shape/region label. The shape from which the boundary segment was taken (arc or straighter). Try correlating the respective segments. It works decently for letter recognition. I've never had a chance to exploit that in gray-scale images, but it bears enormous potential.
- OrgDth: the depth map at which the boundary segment was obtained (0-indexing). Perhaps useful in a high-contrast first analysis.

This vectorial output could also be tried with Transformer networks.

The vector file will be loaded by the program matching vectors, mvec, to be explained in https://github.com/Sehbau/MtchVec.

2.5 Boundary Information (.BonBbox, .BonAsp, .BonPix)

The two scripts exampleLoadBon.m and exampleLoadBonPix.m load those data files.

The file with extension .BonBbox contains essentially the same information as the .Bbox file as described under Section 2.1, but in slightly different format and with additional information. The differences are:

- the bounding boxes are concatenated across depth, but the depth information is still available as the 6th parameter.
- the box coordinates are scaled to original size already.
- the contrast value for the boundaries is given as 5th parameter.

The file with extension .BonAsp lists some additional boundary aspects:

- chromatic values red, green, blue for the pixels along the boundary, not the region inside. Thus for small regions it may not be an optimal chromatic representation.
- coverage of the boundary area, as proportion of the image/map
- border values as introduced under Section 2.1 already.
- perimeter, which is given as absolute value.
- area in pixels for the connected component (not the boundary itself), thus excluding holes
- area of boundary and thus including holes.

The file with extension .BonPix contains the boundary pixels.