# Layered BDM as a Texture and Weighted Network Descriptor to Estimate Kolmogorov Complexity

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A layered version of the Block Decomposition Method[1], serves as a descriptor of both weighted networks and grayscale or color images. This descriptor provides an estimate of Kolmogorov Complexity that's sensitive to morphological perturbative [2]. To estimate the complexity of a grayscale texture, we quantize it and aggregate the estimated Kolmogorov complexity values of binary 4 x 4 squares, estimated through the Coding Theorem Method [3, 4].

```
Clear["Global`*"]
In[4]:=
      SetDirectory[NotebookDirectory[]]
Out[5]= C:\Users\Antonio Rueda Toicen\Documents\Python\EstimationKolmogorovComplexityImages\
       ImageAnalysisWithAlgorithmicInformation
      data = Import["fourByFourCTMs.csv", "CSV", "Numeric" → False]
In[6]:=
       {000000000000001, 23.347935957593144}, {000000000000010, 24.325701071360243},
           Out[6]=
        {111111111111110, 23.347935957593144}, {11111111111111, 22.006706292292176}}
                                          show all
                                                    set size limit...
      large output
                   show less
                              show more
 In[7]:= fourByFourCTMs = Transpose@{data[[All, 1]], ToExpression /@ data[[All, 2]] };
 In[8]:= Table
       (CTM[fourByFourCTMs[[i, 1]]] = fourByFourCTMs[[i, 2]]), {i, 1, 65536, 1}];
 In[9]:= CTM [ "0000000000000000000000"]
Out[9]= 22.0067
In[10]:= Clear[data, fourByFourCTMs]
```

"Layered BDM" works through the binary quantization of a texture's digital levels. The examples below quantize textures using 256 digital levels (byte resolution);  $2^{16}$ ,  $2^{32}$ , etc. quantizations are obviously also possible.

```
In[11]:= layerDecomposition[image_] :=
      Module[{getLayers, getBlocks, blockCount, stringifiedBlocks},
       getLayers[imag_] := ParallelTable[Unitize[
          ImageData[ColorConvert[imag, "Grayscale"], "Byte"], i], {i, 1, 255, 1}];
       getBlocks[layers_] := Nest[Flatten[#, 1] &,
         Partition[#, {4, 4}, 1] & /@ layers, 2];
       blockCount = Tally[getBlocks[getLayers[image]]];
       stringifiedBlocks =
        StringJoin /@ Map[ToString, (Flatten /@ blockCount[[All, 1]]), {2}];
      Total[CTM /@ stringifiedBlocks] + Total[Log2[blockCount[[All, 2]]]]
```

### Layered BDM as a Texture Descriptor

```
In[12]:= woodTextures =
        (Image[ColorConvert[#, "Grayscale"], "Byte"] & /@ ImageResize[#, {128, 128}] & /@
           (ExampleData[#] & /@
             {{"Texture", "Wood"}, {"Texture", "Wood2"}, {"Texture", "Wood3"}}));
In[13]:= smallWT = ImageResize[woodTextures[[1]], {15, 15}]
Out[13]=
In[14]:= layerDecomposition[smallWT]
Out[14]= 28 364.8
In[15]:= woodTextures
Out[15]=
```

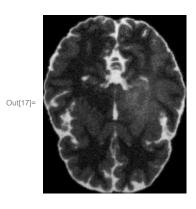
BDM seems to align with the intuitive notion of a texture's complexity.

```
In[16]:= layerDecomposition /@ woodTextures
Out[16]= {457899., 569941., 398956.}
```

## Test on MR Image with/without an Artifact

ln[17]:= t2BrainSlice = Image[ColorConvert[ImageCrop@

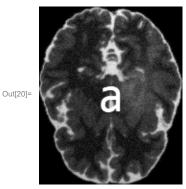




```
In[18]:= t2BrainSlice // ImageData // Dimensions
Out[18]= \{185, 149\}
In[19]:= aImage = ColorNegate@
        ColorConvert[Image[Rasterize[Style["a", FontSize → 50]], "Byte"], "Grayscale"]
```



In[20]:= brainWLetter = Image[ColorConvert[ImageAdd[t2BrainSlice, aImage], "Grayscale"], "Byte"]



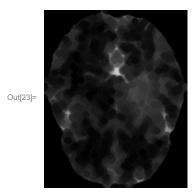
```
In[21]:= brainWLetter // ImageData // Dimensions
Out[21]= \{185, 149\}
```

The image with a simple texture artifact has lower BDM and lower texture complexity.

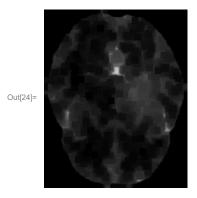
```
In[22]:= layerDecomposition /@ {t2BrainSlice, brainWLetter}
Out[22]= \left\{1.39659 \times 10^6, 1.38208 \times 10^6\right\}
```

# Sensitivity to Morphological Perturbation

In[23]:= t2BrainSliceWDiskErosion = Erosion[t2BrainSlice, DiskMatrix[3]]



In[24]:= t2BrainSliceWBoxedErosion = Erosion[t2BrainSlice, BoxMatrix[3]]



LayeredBDM is highly sensitive to morphological perturbations of the data.

ln[25]:= layerDecomposition /@ {t2BrainSliceWBoxedErosion, t2BrainSliceWDiskErosion} Out[25]= { **59 706.4**, **66 779.8**}

## Layered BDM as a Weighted Network Descriptor

BDM can be used to evaluate networks trained through backpropagation.

ln[26]:= SeedRandom[1]; rg = RandomGraph[{10, 30}];

```
In[27]:= SeedRandom[1];
       {wrg = Graph[EdgeList[rg], EdgeWeight → RandomReal[1, Length[EdgeList[rg]]],
           EdgeLabels → "EdgeWeight", ImageSize → Medium],
        ArrayPlot@WeightedAdjacencyMatrix[wrg] }
                                     0.306427
                              0.819967
                                           0.925275
                                  0.542247
                      0.70047
               0.211826
                              0.580474
                                          0.247495
                                  0.977172
 Out[27]=
                   0.817389
                           0.241361
                                        0 11142
                                                 0.29287
                                                      0.422851
        0.748657
                                      0.712012
                 0.59326
                                            0.9780526
                                0.325351
                     0.396006 0.390582
                                    0.231 P5825163
                                                  0.208051
                               0.0657388
                         0.169013
                                    0.518774
 In[28]:= floatToByte[float_] := Floor[If[float === 1.0, 255, float * 256.0]]
 In[29]:= floatToByte /@ {0.001, 0.999}
 Out[29]= \{0, 255\}
 In[30]:= byteWeightedMatrix = Map[floatToByte, Normal[WeightedAdjacencyMatrix[wrg]], {2}];
       MatrixForm[byteWeightedMatrix]
Out[30]//MatrixForm=
          0
              209
                   28
                       202
                            48
                                  61
                                      16
                                           138
                                                59
                                                     101
         209
              0
                   0
                        0
                            179
                                 54
                                       0
                                            0
                                                 0
                                                     191
                       108
                                 250
                                     211
                                           236
         28
              0
                   0
                            63
                                               147
                                                      0
         202
              0
                  108
                        0
                             74
                                  0
                                       0
                                            0
                                                53
                                                      0
             179
                        74
         48
                   63
                             a
                                 148
                                      32
                                           78
                                                182
                                                      a
                                           209
              54
                  250
                        a
                            148
                                  a
                                       99
                                                83
                                                     151
         61
              0
                  211
                        0
                             32
                                 99
                                       0
                                            0
                                                132
         16
                                                      0
         138
               a
                  236
                        0
                             78
                                 209
                                       a
                                            0
                                                 0
                                                      0
         59
               0
                  147
                        53
                            182
                                 83
                                      132
                                            0
                                                 0
                                                     43
        101
             191
                   0
                        0
                             0
                                 151
                                            0
                                                43
                                                      0
 ln[31]:= layerDecompositionForWeightedGraphs [weightedGraph_] :=
        Module[{floatToByte, getLayers, getBlocks,
          blockCount, stringifiedBlocks, weightedAdjMatrix},
         floatToByte[float_] := Floor[If[float === 1.0, 255, float * 256.0]];
         getLayers[w_] := ParallelTable[
            Unitize[Map[floatToByte, weightedAdjMatrix, {2}], i], {i, 1, 255, 1}];
         getBlocks[layers ] := Nest[Flatten[#, 1] &,
            Partition[#, {4, 4}, 1] & /@ layers, 2];
         weightedAdjMatrix = Normal[WeightedAdjacencyMatrix[weightedGraph]];
         blockCount = Tally[getBlocks[getLayers[weightedAdjMatrix]]];
         stringifiedBlocks =
          StringJoin /@ Map[ToString, (Flatten /@ blockCount[[All, 1]]), {2}];
         Total[CTM /@ stringifiedBlocks] + Total[Log2[blockCount[[All, 2]]]]
```

In[32]:= layerDecompositionForWeightedGraphs[wrg]

Out[32]= 12 323.2

#### References

- [1] Antonio Rueda-Toicen, "Morphological Image Analysis through Estimations of Kolmogorov Complexity" (in preparation)
- [2] Hector Zenil, Santiago Hernández Orozco, Narsis A. Kiani, Fernando Soler-Toscano, Antonio Rueda-Toicen, and Jesper Tegner "A Decomposition Method for Global Evaluation of Shannon Entropy and Local Estimations of Algorithmic Complexity", https://arxiv.org/abs/1609.00110
- [3] Fernando Soler Toscano, Hector Zenil, Jean-Paul Delahaye, and Nicolas Gauvrit (2014) "Calculating Kolmogorov Complexity from the Output Frequency Distributions of Small Turing Machines." PLoS ONE 9 (5): e96223.