

# Texture flattening with Poisson image editing

This project focuses on the gradient domain blending based on Poisson equation. The goal of this part is to flatten the texture of an image (or of a selected region of an image) based on the Poisson equation. We will follow the technique described in the following paper:

"Poisson Image Editing", Perez, P., Gangnet, M., Blake, A. SIGGRAPH 2003

It discusses the *texture flattening* method in section 4, which you are strongly encouraged to read prior to starting this part of the project.

Recall from par 1, that the key idea of the gradient domain blending is to apply a so-called guidance vector field  $v$ , which might be or not the gradient field of a source function  $g$ , to the target image's replacement pixels, but keep other pixels. For continuous image function, it can be summarized as the following equation:

$$\min_f \int_{\Omega} |\nabla f - v|^2, \quad \text{with } f|_{\partial\Omega} = f^*|_{\partial\Omega},$$

where  $f$  is the blending image function,  $f^*$  is the target image function,  $v$  is the guidance vector field,  $\Omega$  is the blending region, and  $\partial\Omega$  is the boundary of blending region, that is  $\partial\Omega = \{p \in S \setminus \Omega : N_p \cap \Omega \neq \emptyset\}$ , where  $S$  is the source image domain and  $N_p$  is the set of 4-connected neighbors of pixel  $p$  in  $S$ . In the discrete pixel grid, the previous equation admits the form:

$$\min_{f|_{\Omega}} \sum_{p \in \Omega} \sum_{q \in N_p \cap \Omega} ((f_p - f_q) - v_{pq})^2, \quad \text{with } f|_{\partial\Omega} = f^*|_{\partial\Omega}$$

The optimal solution  $\{f_p\}_{p \in \Omega}$  of the above minimization problem satisfies the following set of linear equations:

$$|N_p|f_p - \sum_{q \in N_p \cap \Omega} f_q = \sum_{q \in N_p \cap \partial\Omega} f_q^* + \sum_{q \in N_p} v_{pq} \quad \forall p \in \Omega$$

In the previous part of this project, the guidance field depended, partly or wholly, on the gradient field of a source image  $g$ . Alternatively, in-place image transformations can be defined by using a guidance field depending entirely on the original image (target image). In the texture flattening approach, the image gradient  $\nabla f^*$  is passed through a sparse sieve that retains only the most salient features:

$$v(x) = M(x) \nabla f^*(x)$$

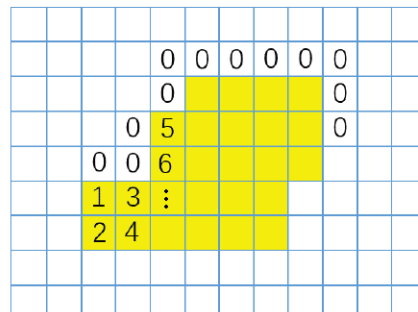
where  $M$  is a binary mask turned on at a few locations of interest. A good choice for  $M$  is an edge detector, in which case the discrete counterpart is given by

$$v_{pq} = \begin{cases} f_p^* - f_q^*, & \text{if } M_p = 1 \text{ or } M_q = 1 \\ 0, & \text{otherwise} \end{cases}$$

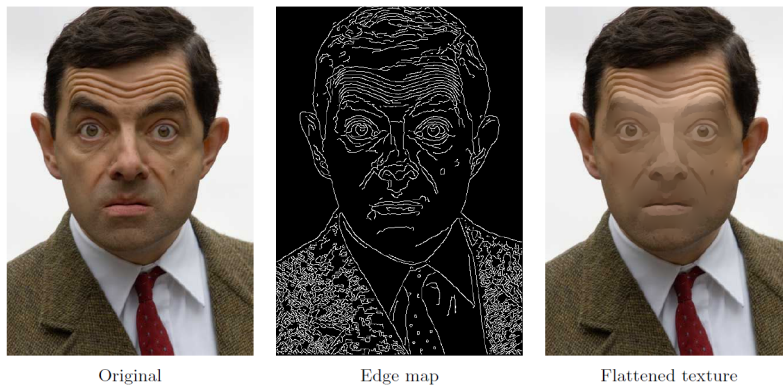
for all  $p \in \Omega$  and  $q \in N_p$ .

As before, our task is to solve all  $\{f_p\}_{p \in \Omega}$  from the set of linear equations. If we form all  $\{f_p\}_{p \in \Omega}$  as a vector  $x$ , then the set of linear equations can be converted to the form  $Ax = b$ , whose solution can be efficiently computed with in `Matlab` using the command:

`x = A\b`. The replacement pixels' intensity are solved by the linear system  $Ax = b$ . But not all the pixels in target image need to be computed. Only the pixels mask as 1 in the mask image will be used to blend. In order to reduce the number of calculations, you need to index the replacement pixels so that each element in  $x$  represents one replacement pixel. As shown in the figure below, the yellow locations are the replacement pixels (indexed from top to bottom).



Your results should look like this:



The images for the assignment are attached below:



## Your Function

Save Reset MATLAB Documentation (<https://www.mathworks.com/help/>)

```

1 function I = TextureFlattening
2
3 % read image and mask
4 target = im2double(imread('bean.jpg'));
5 mask = imread('mask_bean.bmp');
6
7 % edge detection
8 Edges = edge(rgb2gray(target),'canny',0.1);
9
10
11
12 N=sum(mask(:)); % N: Number of unknown pixels == variables
13
14
15
16 % YOUR CODE STARTS HERE
17
18 % enumerating pixels in the mask
19 mask_id = zeros(size(mask));
20 mask_id(mask) = 1:N;
21
22 % neighborhood size for each pixel in the mask
23 % find gets row,column of nonzero elements
24 [ir,ic] = find(mask);
25
26 Np = zeros(N,1);
27

```

```

28 for ib=1:N
29
30     i = ir(ib);
31     j = ic(ib);
32
33     Np(ib)= double((i> 1))+ double((j> 1))+ ...
34             double((i< size(target,1))) + double((j< size(target,2)));
35 end
36
37
38 % start with Np along diag
39 % add at most 4 -1s
40 % in case less than four, boundary values are already added (to b!)
41
42 i = 1:N;
43 j = 1:N;
44 A = sparse(i,j,Np,N,N,4*N);
45
46 for p = 1:N
47     % get row and column of current pixel
48     row = ir(p);
49     col = ic(p);
50
51     % setup row vector to enter in (sparse) matrix
52     v = A(p,:);
53
54     % now check four cases
55     % check right (same row, col+1)
56     % for id:
57     numeral_in_mask = sub2ind(size(mask),row,col+1)
58     % id = mask_id(numeral_in_mask)
59     if mask(row,col+1) ~= 0
60         numeral_in_mask = sub2ind(size(mask),row,col+1);
61         right_id = mask_id(numeral_in_mask);
62         v(right_id) = -1;
63     end
64     % check up (row+1, same col):
65     if mask(row+1,col) ~= 0
66         numeral_in_mask = sub2ind(size(mask),row+1,col);
67         up_id = mask_id(numeral_in_mask);
68         v(up_id) = -1;
69     end
70     % check left (same row, col-1):
71     if mask(row,col-1) ~= 0
72         numeral_in_mask = sub2ind(size(mask),row,col-1);
73         left_id = mask_id(numeral_in_mask);
74         v(left_id) = -1;
75     end
76     % check down (row-1, same col):
77     if mask(row-1,col) ~= 0
78         numeral_in_mask = sub2ind(size(mask),row-1,col);
79         down_id = mask_id(numeral_in_mask);
80         v(down_id) = -1;
81     end
82
83     % update A:
84     A(p,:) = v;
85
86 end
87
88 % output intialization
89 I = target;
90
91
92 for color=1:3 % solve for each colorchannel
93
94     % compute b for each color
95     b=zeros(N,1);

```

```

96
97     for ib=1:N
98
99         i = ir(ib);
100        j = ic(ib);
101
102
103        if (i>1)
104            fpq = target(i,j,color)-target(i-1,j,color);
105            Mp = Edges(i,j);
106            Mq = Edges(i-1,j);
107            M = Mp*Mq;
108            vpq = M*fpq;
109            b(ib)=b(ib)+ target(i-1,j,color)*(1-mask(i-1,j))+ vpq;
110        end
111
112        if (i<size(mask,1))
113            fpq = target(i,j,color)-target(i+1,j,color);
114            Mp = Edges(i,j);
115            Mq = Edges(i+1,j);
116            M = Mp*Mq;
117            vpq = M*fpq;
118            b(ib)=b(ib)+ target(i+1,j,color)*(1-mask(i+1,j))+vpq;
119        end
120
121        if (j>1)
122            fpq = target(i,j,color)-target(i,j-1,color);
123            Mp = Edges(i,j);
124            Mq = Edges(i,j-1);
125            M = Mp*Mq;
126            vpq = M*fpq;
127            b(ib)= b(ib) + target(i,j-1,color)*(1-mask(i,j-1))+vpq;
128        end
129
130
131        if (j<size(mask,2))
132            fpq = target(i,j,color)-target(i,j+1,color);
133            Mp = Edges(i,j);
134            Mq = Edges(i,j+1);
135            M = Mp*Mq;
136            vpq = M*fpq;
137            b(ib)= b(ib)+ target(i,j+1,color)*(1-mask(i,j+1))+vpq;
138        end
139
140
141
142
143    end
144
145
146    % solve linear system A*x = b;
147    % your CODE begins here
148
149    x = A\b;
150    % your CODE ends here
151
152
153
154
155
156
157    % impaint target image
158
159    for ib=1:N
160        I(ir(ib), ic(ib),color) = x(ib);
161    end
162
163 end

```

```
164
165
166 % YOUR CODE ENDS HERE
167
168 figure(1), imshow(target);
169 figure(2), imshow(I);
170
171 end
```

## Code to call your function

 Reset

```
1 I = TextureFlattening;
```

 Run Function



Your request timed out. If the problem persists, check your code for an infinite loop or improve your code's efficiency.

## Previous Assessment: Incorrect

Submit



 Test 1