

# Total Variation Denoising

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## Total Variation Denoising: Color Images

The Rudin-Osher-Fatemi Total Variation (TV) Energy is the most famous and widely used image restoration model and for an input image  $f$  is defined as:

$$\min E_{TV}[u|f] = \int_{\Omega} \|\nabla u\| d\vec{x} + \lambda \int_{\Omega} (u - f)^2 d\vec{x}$$

Where  $f$  is the original noisy image and  $\lambda$  is parameter (fidelity weight) that controls the relative importance of the terms. According to *Lagrange-Euler* equation, the first variation of energy of this function is:

$$\nabla E = -\nabla \cdot \left( \frac{\nabla u}{\|\nabla u\|} \right) + 2\lambda(u - f)$$

Where

$$\|\nabla u\| = \sqrt{u_x^2 + u_y^2}$$

The energy is convex, so we can use steepest descent to evolve the PDE:

$$\frac{\partial u}{\partial t} = \frac{u_{xx}u_y^2 - 2u_xu_yu_{xy} + u_{yy}u_x^2}{(u_x^2 + u_y^2)^{3/2}} - 2\lambda(u - f)$$

To apply TV denoising on a color image, we calculate the equation above on each of the 3 RGB channels and storing it in the corresponding channel of a new image. As the first step inside the for loop, we calculate the central differences to approximate the first derivatives  $u_x$  and  $u_y$ :

$$\begin{aligned} u_x &\approx D_x^0 u = (u(x+1, y) - u(x-1, y))/2 \\ u_y &\approx D_y^0 u = (u(x, y+1) - u(x, y-1))/2 \end{aligned}$$

And to calculate  $u_{xx}$  and  $u_{yy}$  we perform central differences twice on the input image; to calculate  $u_{xy}$  we use diagonal derivative:

$$\begin{aligned} u_{xx} &\approx D_x^0(D_x^0 u) = u(x+1, y) - 2u(x, y) + u(x-1, y) \\ u_{yy} &\approx D_y^0(D_y^0 u) = u(x, y+1) - 2u(x, y) + u(x, y-1) \\ u_{xy} &\approx D_x^0(D_y^0 u) = u(x+1, y+1) + u(x-1, y-1) - u(x-1, y+1) - u(x+1, y-1) \end{aligned}$$

Finally the PDE becomes:

$$u^{n+1} = u^n + \Delta t \left[ \frac{u_{xx}u_y^2 - 2u_xu_yu_{xy} + u_{yy}u_x^2}{(u_x^2 + u_y^2)^{3/2}} - 2\lambda(u - f) \right]$$

We use Forward Euler Method with Neumann boundary conditions to solve the above equation.

## Coding Total Variation on Color Images

*Listing 1* shows the Total Variation denoising applied on a grayscale image in Matlab; *Listing 2* uses the same function to apply TV on color images, and calculates SNR/RMSE of the output:

```

1 function [u] = tv(f, lambda)
2     % Computes the total variation of an input grayscale image
3
4     dt = 0.1;                % time step
5     T = 20;                  % stopping time
6     a = 0.1;                 % fudge factor
7     [m,n] = size(f);         % image size
8     f = double(f);           % convert to double
9     u = f;                   % initialization
10
11    for t = 0:dt:T
12        % u_x = (u(x+1,y) - u(x-1,y)) / 2
13        u_x = (u(:, [2:n,n]) - u(:, [1,1:n-1])) / 2;
14
15        % u_y = (u(x,y+1) - u(x,y-1)) / 2
16        u_y = (u([2:m,m], :) - u([1,1:m-1], :)) / 2;
17
18        % u_xx = u(x+1,y) - 2u(x,y) + u(x-1,y)
19        u_xx = u(:, [2:n,n]) - 2 * u + u(:, [1,1:n-1]);
20
21        % u_yy = u(x,y+1) - 2u(x,y) + u(x,y-1)
22        u_yy = u([2:m,m], :) - 2 * u + u([1,1:m-1], :);
23
24        % u_xy = (u(x+1,y+1) + u(x-1,y-1) - u(x-1,y+1) - u(x+1,y
25                - 1)) / 4
26        u_xy = (u([2:m,m], [2:n,n]) + u([1,1:m-1], [1,1:n-1]) - u
27                ([2:m,m], [1,1:n-1]) - u([1,1:m-1], [2:n,n])) / 4;
28
29        k_num = (u_xx.*u_y.^2) - 2*(u_x.*u_y.*u_xy) + (u_yy.*u_x.^2);
30        k_denom = (u_x.^2 + u_y.^2).^(3/2) + a;
31        pde = k_num ./ k_denom - 2 * lambda * (u - f);
32
33        u = u + dt * pde;
34    end
35
36    u = uint8(u);
37 end

```

Listing 1: Total Variation function on grayscale images in Matlab

```

1 function [u, snr, rmse] = colortv(f, lambda)
2     u = f;
3     for i=1:3; u(:,:,i) = tv(f(:,:,i), lambda); end;
4     snr = SNR(rgb2gray(f), rgb2gray(u));
5     rmse = RMSE(rgb2gray(f), rgb2gray(u));
6 end

```

Listing 2: Total Variation + RMSE/SNR on color images in Matlab

Figure 1 shows a test image and three TV denoised images for different values of fidelity weight parameter. As shown below, when fidelity weight is too small the output image becomes blurry; on contrast when fidelity weight is too large, the output image resembles the original noisy input image:

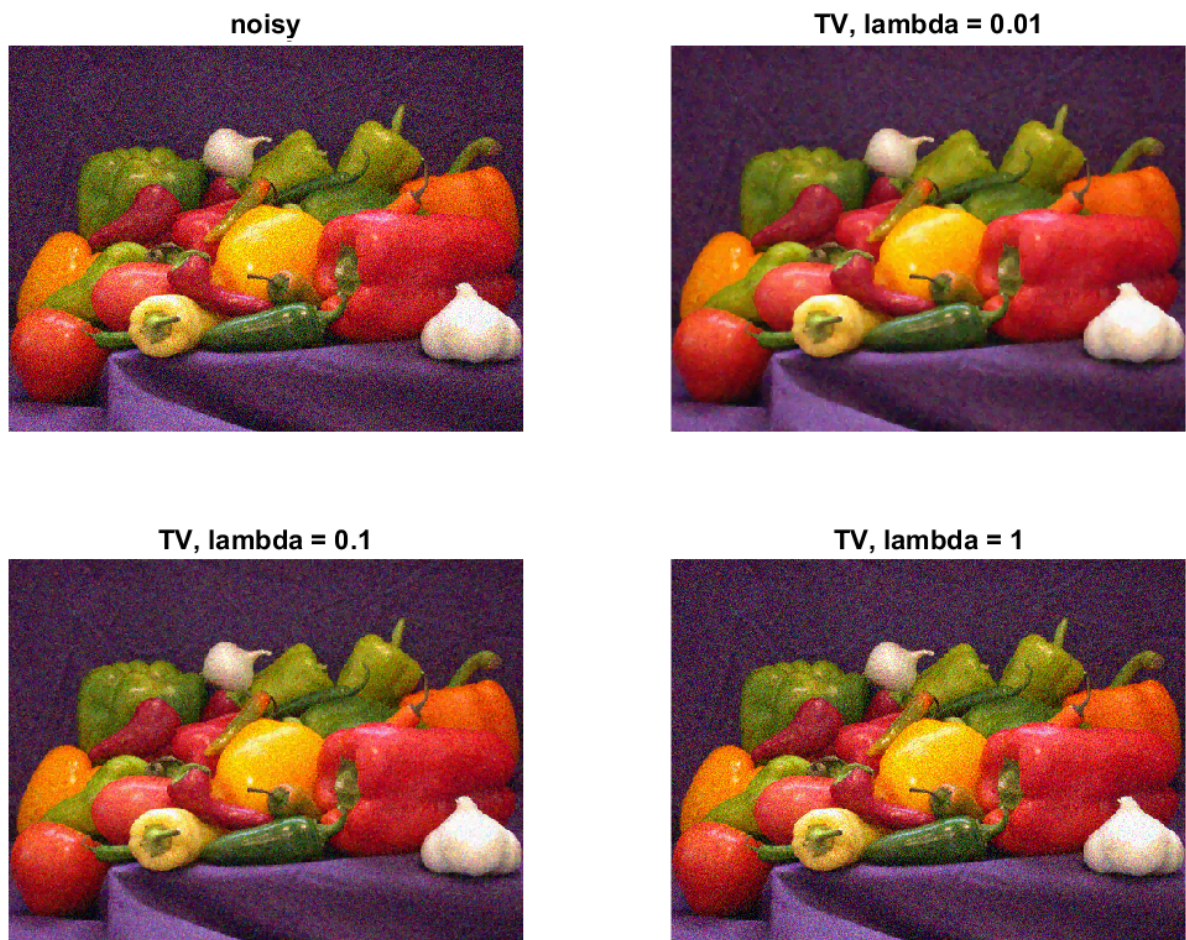


Figure 1: Total Variation denoising applied on a color image with different fidelity weight values.