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function [u,q] = AdaptiveSmoothingUpwind(f, lambda, sigma)
%This function implements an alteration of the upwind TV scheme that
%smooths less near edges as described in "Variable Exponent, Linear Growth
%Functionals in Image Restoration" by Yunmei Chen, Stacey Levine, Murali Rao
%This matlab code was written by Katie Heaps

%Input is noisy image, lambda (optional), and standard deviation (optional)
%Output is improved image

if(nargin<1)
    error('requires input image');
end
if(nargin<2)
    lambda = 35;
end
if(nargin<3)
    sigma = 20;
end

%initializations
f = double(f);
u = f;
tau = 1/32;           %step size
scale = .05;          %scale for Gaussian
k = .0075;            %parameter edge detector
[M,N] = size(f);      %M,N = size
gauss = size(M,N);

%boundary condition = zero padding
g=zeros(M+2,N+2);
g(2:M+1,2:N+1) = f;

%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%
%Calculate q
%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%

%Calculate Gaussian
for i=1:M
    for j=1:N
        gauss(i,j) = sqrt(2*pi)*scale*exp(-2*pi^2*scale^2*(i^2+j^2));
    end
end

%%perform convolution

%calculate and shift FFT
F = fftshift(fft2(f));
GAUSS = fftshift(fft2(gauss));

%performs multiplication
R = F.*GAUSS;

%calculate and shift inverse of result
conv = ifft2(R);
conv = real(conv);

for i=1:M
    for j=1:N
        conv(i,j) = conv(i,j)*(-1)^(i+j);
    end
end

%calculate q and q*
q1 = zeros(M+2,N+2);
q1(2:M+1,2:N+1) = conv;
gradq1 = zeros(M+2,N+2,2);
q = zeros(M+2,N+2);
qstar = zeros(M+2,N+2);

gradq1(2:M+1,2:N+1,1) = q1(3:M+2,2:N+1)-q1(1:M,2:N+1);

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gradq1(2:M+1,2:N+1,2) = q1(2:M+1,3:N+2)-q1(2:M+1,1:N);
magq1(:, :) = sqrt(gradq1(:, :, 1).^2+gradq1(:, :, 2).^2);

q = 1+1./(1+k*magq1.^2);
qstar = q./(q-1);

%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%
%Perform Gradient Descent
%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%

w = zeros(M+2,N+2);
mag = zeros(M+2,N+2);
divp = zeros(M+2,N+2);
grad = zeros(M+2,N+2,4);
p = zeros(M+2,N+2,4);
pn = zeros(M+2,N+2,4);

for iter = 1:200

    %calculate gradient
    w = divp - g/lambda;

    grad(2:M+1,2:N+1,1) = w(2:M+1,2:N+1) - w(3:M+2,2:N+1);
    grad(2:M+1,2:N+1,2) = w(2:M+1,2:N+1) - w(1:M,2:N+1);
    grad(2:M+1,2:N+1,3) = w(2:M+1,2:N+1) - w(2:M+1,3:N+2);
    grad(2:M+1,2:N+1,4) = w(2:M+1,2:N+1) - w(2:M+1,1:N);

    mag = (p(:, :, 1).^2+p(:, :, 2).^2+p(:, :, 3).^2+p(:, :, 4).^2).^5;

    %calculate p^(n+1/2)
    pn(:, :, 1) = p(:, :, 1) - tau*(grad(:, :, 1)+1/lambda.*mag.^(qstar-2).*p(:, :, 1));
    pn(:, :, 2) = p(:, :, 2) - tau*(grad(:, :, 2)+1/lambda.*mag.^(qstar-2).*p(:, :, 2));
    pn(:, :, 3) = p(:, :, 3) - tau*(grad(:, :, 3)+1/lambda.*mag.^(qstar-2).*p(:, :, 3));
    pn(:, :, 4) = p(:, :, 4) - tau*(grad(:, :, 4)+1/lambda.*mag.^(qstar-2).*p(:, :, 4));

    %take only positive differences
    pn = max(pn, 0);

    %calculate magnitue of pn1
    mag = (pn(:, :, 1).^2+pn(:, :, 2).^2+pn(:, :, 3).^2+pn(:, :, 4).^2).^5;

    %if magnitude of p is greater than 1, divide by magnitude
    p(:, :, 1) = pn(:, :, 1)./max(1, mag);
    p(:, :, 2) = pn(:, :, 2)./max(1, mag);
    p(:, :, 3) = pn(:, :, 3)./max(1, mag);
    p(:, :, 4) = pn(:, :, 4)./max(1, mag);

    %calculate divp
    divp(2:M+1,2:N+1) = (p(2:M+1,2:N+1,1) - p(1:M,2:N+1,1)) + (p(2:M+1,2:N+1,2)-p(
3:M+2,2:N+1,2)) + (p(2:M+1,2:N+1,3) - p(2:M+1,1:N,3)) + (p(2:M+1,2:N+1,4)- p(2:M+1,3
:N+2,4));

    %calculate new lambda (as described on page 93 of An Algorithm for
    %Total Minimization and Applications
    %lambda = ((M*N)^.5*sigma / (sum(sum(divp(2:M+1,2:N+1).^2)))^5 );

end

%calculate denoised image
u = u - lambda*divp(2:M+1,2:N+1);

end

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