Computer Vision: Homework 5: Panorama

From GitHub Repository: https://github.com/dcyoung/ImageAlign
Functions:

```
• detect features
   • describe features
   • estimate homography
   • RANSAC H
   • calc residuals
function [ r1, c1, r2, c2 ] = detect features( grayImg1, grayImg2 )
%use harris corner detector
    sigma = 2;
    thresh = 0.05; From GitHub Repository
    radius = 2;
    disp = 5;
    [~, r1, c1] = harris(grayImg1, sigma, thresh, radius, disp);
    [~, r2, c2] = harris(grayImg2, sigma, thresh, radius, disp);
end
Alternatively:
% corner1 = corner(I1 gray, 'harris');
% r1 = corner1(:,1);
% c1 = corner1(:,2);
% corner2 = corner(I2 gray, 'harris');
% r2 = corner2(:,1);
% c2 = corner2(:,2);
function [ featDescriptions ] = describe features( img, radius, r, c )
    numFeat = length(r); %number of features
    featDescriptions = zeros(numFeat, (2 * radius + 1)^2);
    % matrix with a single 1 in the center and zeros all around it
    padHelper = zeros(2 * radius + 1);
    padHelper(radius + 1, radius + 1) = 1;
    % use the pad Helper matrix to pad the img such that the border values
    % extend out by the radius
    paddedImg = imfilter(img, padHelper, 'replicate', 'full');
    %Extract the neighborhoods around the found features
    for i = 1 : numFeat
        rowRange = r(i) : r(i) + 2 * radius;
        colRange = c(i) : c(i) + 2 * radius;
        neighborhood = paddedImg(rowRange, colRange);
        featDescriptions test(i,:) = [rowRange colRange];
        flattenedFeatureVec = neighborhood(:);
        featDescriptions(i,:) = flattenedFeatureVec;
    end
    %Normalize all descriptors to have zero mean and unit standard deviation
```

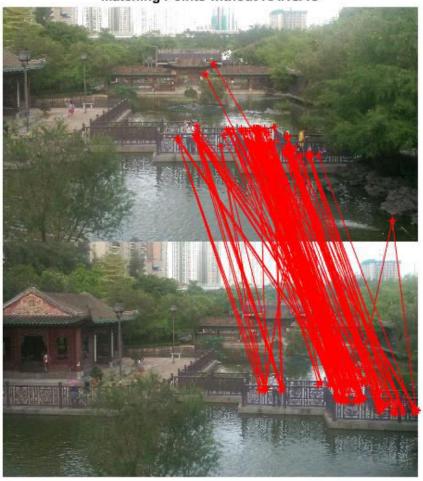
featDescriptions test = zscore(featDescriptions');

```
function [ H, inlierIndices ] = estimate homography( img1Feat, img2Feat )
%ESTIMATE HOMOGRAPHY Summary of this function goes here
  Detailed explanation goes here
   parameters.numIterations = 150; %the number of iterations to run
   parameters.subsetSize = 4; %number of matches to use each iteration
   parameters.inlierDistThreshold = 10; %the minimum distance for an inlier
   parameters.minInlierRatio = .3; %minimum inlier ratio required to store a
fitted model
    [H, inlierIndices] = ransac H(parameters, img1Feat, img2Feat,
@fit homography, @calc residuals);
    display('Number of inliers:');
   display(length(inlierIndices));
   display('Average residual for the inliers:')
    display(mean(calc residuals(H, img1Feat(inlierIndices,:),
img2Feat(inlierIndices,:))));
end
function [ bestFitModel, inlierIndices ] = ransac H( parameters, x, y, fitModelFxn,
    [numMatches, \sim] = size(x);
   numInliersEachIteration = zeros(parameters.numIterations,1);
    storedModels = {};%zeros(parameters.numIterations,3,3);
    for i = 1 : parameters.numIterations
        %display(['Running ransac Iteration: ', num2str(i)]);
        %select a random subset of points
        subsetIndices = randsample(numMatches, parameters.subsetSize);
        x subset = x(subsetIndices, :);
        y subset = y(subsetIndices, :);
        %fit a model to that subset
        model = fitModelFxn(x subset, y subset);
        %compute inliers, ie: find all remaining points that are
        %"close" to the model and reject the rest as outliers
        residualErrors = errorFxn(model, x, y);
        %display(['Mean Residual Error: ', num2str(mean(residualErrors))]);
        inlierIndices = find(residualErrors < parameters.inlierDistThreshold);</pre>
        %record the number of inliers
        numInliersEachIteration(i) = length(inlierIndices);
        %keep track of any models that generated an acceptable numbers of
        %inliers. This collection can be parsed later to find the best fit
        currentInlierRatio = numInliersEachIteration(i)/numMatches;
        if currentInlierRatio >= parameters.minInlierRatio
        %if numInliersEachIteration(i) >= max(numInliersEachIteration)
            %re-fit the model using all of the inliers and store it
```

```
x inliers = x(inlierIndices, :);
            y inliers = y(inlierIndices, :);
            storedModels{i} = fitModelFxn(x inliers, y inliers);
        end
    end
    %display(storedModels);
    %display(numInliersEachIteration);
    %retrieve the model with the best fit (highest number of inliers)
    bestIteration = find(numInliersEachIteration ==
max(numInliersEachIteration));
    bestIteration = bestIteration(1); %incase there was more than 1 with same value
    bestFitModel = storedModels{bestIteration};
    %recalculate the inlier indices for all points, this was done once before
    %when calculting this model, but it wasn't stored for space reasons.
    Recalculate it now so that it can be returned to the caller
    residualErrors = errorFxn(bestFitModel, x, y);
    inlierIndices = find(residualErrors < parameters.inlierDistThreshold);</pre>
end
function residuals = calc residuals(F, matches)
    numMatches = size(matches,1);
    L = (F * [matches(:,1:2) ones(numMatches,1)]')'; % transform points from
    % the first image to get epipolar lines in the second image
    % find points on epipolar lines L closest to matches(:,3:4)
    L = L . / repmat(sqrt(L(:,1).^2 + L(:,2).^2), 1, 3); % rescale the line
    distances = sum(L .* [matches(:,3:4) ones(numMatches,1)],2); %distances from
each pt to its line
    residuals = abs(distances);
end
% MAIN
addpath('C:\Users\Dave\Desktop\Computer Vision\Project 5');
imnames={'IMAG4688.jpg','IMAG4689.jpg'};
I1=imread(imnames{1});
I2=imread(imnames{2});
I1 gray = rgb2gray(I1);
I2 gray = rgb2gray(I2);
% Find row and column locations of corner points in images
[r1, c1, r2, c2] = detect features(I1 gray, I2 gray);
% corner1 = corner(I1 gray, 'harris');
% r1 = corner1(:,1);
% c1 = corner1(:,2);
% corner2 = corner(I2 gray, 'harris');
% r2 = corner2(:,1);
% c2 = corner2(:,2);
% Create feature descriptions for each point. Find corner points in the
% 'neighborhood' radius of each point.
numMatches = 150;
neighborhoodRadius = 20;
```

```
featDescriptions 1 = describe features (I1 gray, neighborhoodRadius, r1, c1);
featDescriptions 2 = describe features(I2 gray, neighborhoodRadius, r2, c2);
distances = pdist2(featDescriptions_1, featDescriptions_2);
[~, distance idx] = sort(distances(:), 'ascend');
BestMatches = distance idx(1:numMatches);
[~,distance idx] = sort(distances(:), 'ascend');
[img1_matchedFeature, img2_matchedFeature] =
ind2sub(size(distances), BestMatches);
match r1 = r1(img1 matchedFeature);
match c1 = c1(img1 matchedFeature);
match r2 = r2(img2_matchedFeature);
match c2 = c2(img2 matchedFeature);
% refined lists of matching points in both pictures
xy1 = [match c1, match r1, ones(numMatches, 1)]';
xy2 = [match c2, match r2, ones(numMatches,1)]';
xx = affine fit(xy1,xy2);
figure(1)
visualize match(xy1,xy2,I1,I2);
title('Matching Points without RANSAC')
xy1 = xy1';
xy2 = xy2';
% RANSAC
[H, inlierIndices] = estimate homography(xy1,xy2);
numMatches = length(inlierIndices);
match c1 H = match c1(inlierIndices);
match c2 H = match c2(inlierIndices);
match_r1_H = match_r1(inlierIndices);
match r2 H = match r2(inlierIndices);
% New Matching Points
xy3 = [match c1 H, match r1 H, ones(numMatches,1)]';
xy4 = [match c2 H, match r2 H, ones(numMatches,1)]';
xx ransac = \overline{\text{affine fit}}(xy3, xy4);
%figure(2)
%visualize match(xy3,xy4,I1,I2);
%title('Matching Points with RANSAC')
%T = maketform('projective',xx ransac);
T = projective2d(H);
img1Transformed = imwarp(I1,T);
%figure(3)
%compositeImg1 = stitch(img1Transformed, I2, H);
%imshow(compositeImg1);
%title('Projective Transform')
figure (4)
%[wholeImg, NewImage, offsets] = draw align image(H, img1Transformed, I2);
save('DAVID MURRAY Project5');
```

Matching Points without RANSAC



Matching Points with RANSAC

