

CSE185

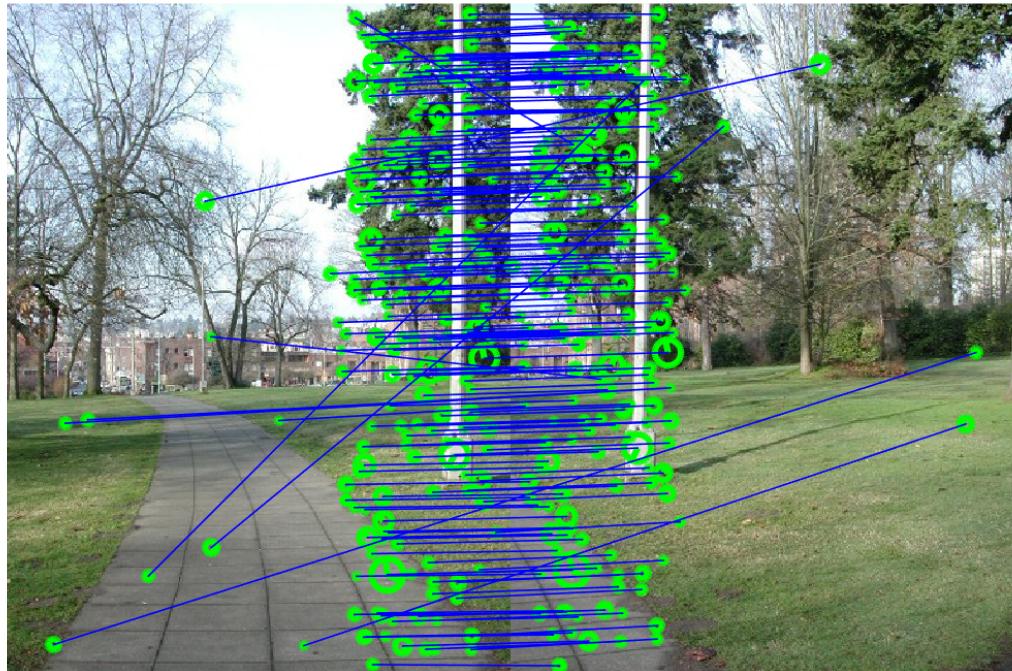
# Introduction to Computer Vision

## Lab 09: Image Stitching

Instructor: Prof. Ming-Hsuan Yang  
TA: Taihong Xiao & Tiantian Wang

# Image Stitching

---



# 4 Steps for Image Stitching

---

1. Feature Extraction (SIFT or Harris)
2. Feature Matching
3. Image Alignment with RANSAC
4. Image Blending/Stitching

# 4 Steps for Image Stitching

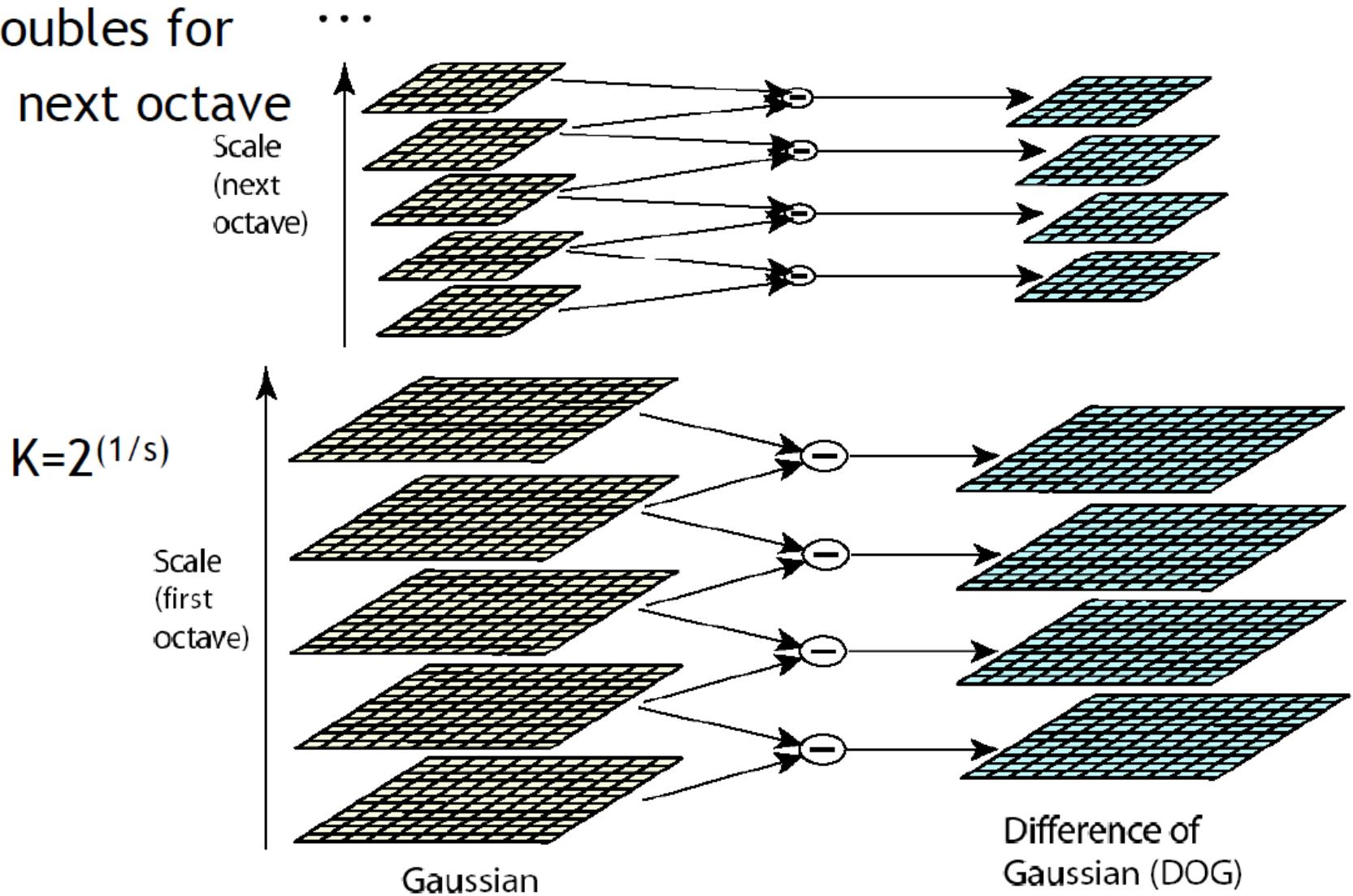
---

- 1. Feature Extraction (SIFT or Harris)**
2. Feature Matching
3. Image Alignment with RANSAC
4. Image Blending/Stitching

# SIFT Features Detector

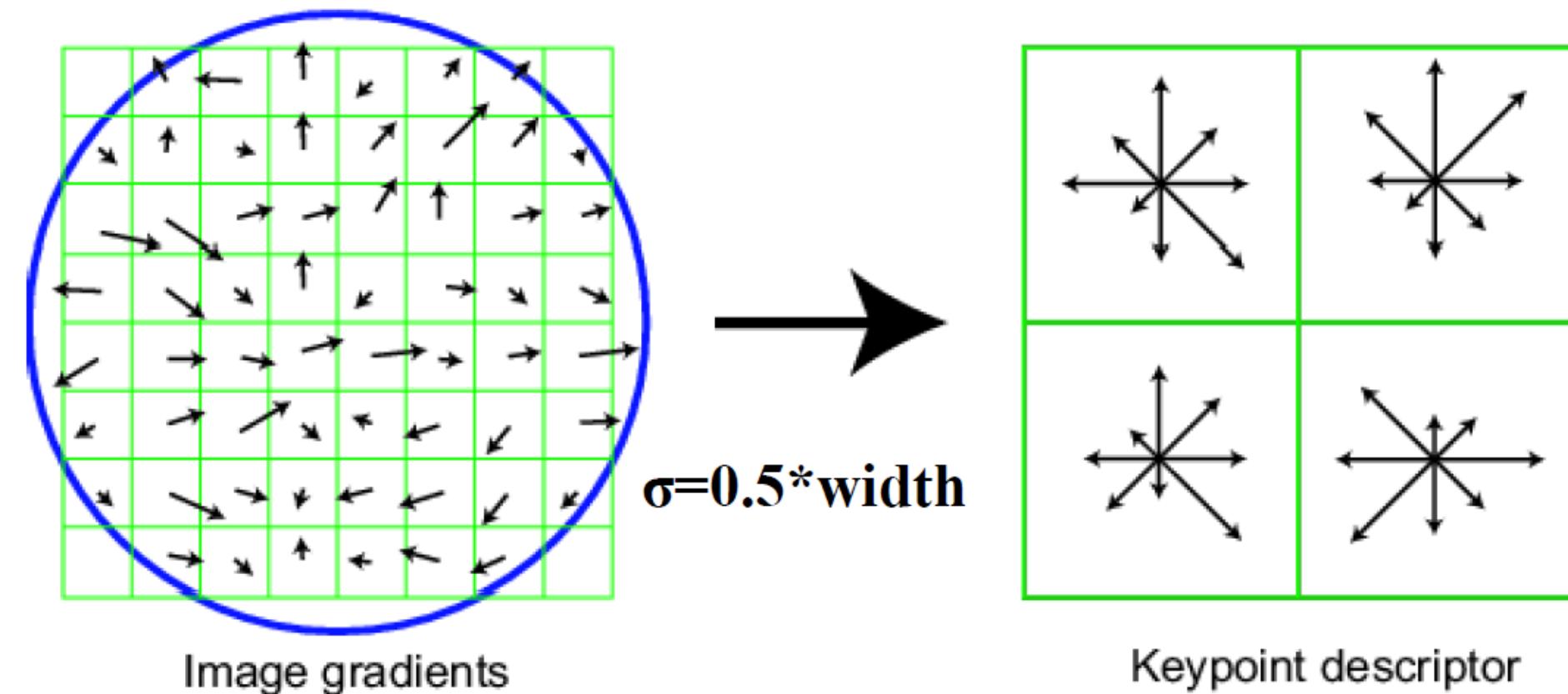
- Use Difference-of-Gaussian to detect multi-scale features

$\sigma$  doubles for  
the next octave



# SIFT Features Descriptor

- Use magnitude and orientation of gradients to describe feature points as 128-D feature vectors



# vlfeat

---

- Use vlfeat toolbox to extract SIFT features
  - download the toolbox and decompress in an folder vlfeat-0.9.20-bin
  - Setup in MATLAB:

```
run('vlfeat-0.9.20-bin/toolbox/vl_setup');
```

- SIFT feature in vlfeat:  
<http://www.vlfeat.org/overview/sift.html>

# Extract SIFT Features

1. Load image as single format
2. convert to gray scale
3. apply `vl_sift()`

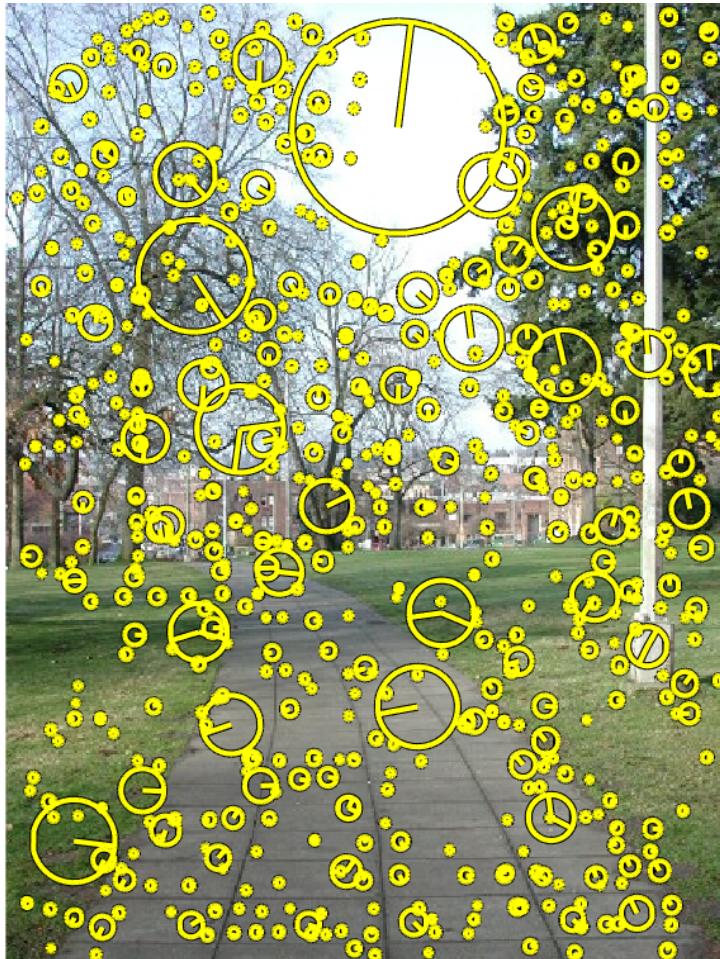
```
img1 = im2single(imread('prtn13.jpg'));  
img2 = im2single(imread('prtn12.jpg'));  
  
%% SIFT feature extraction  
I1 = rgb2gray(img1);  
I2 = rgb2gray(img2);  
[f1, d1] = vl_sift(I1);  
[f2, d2] = vl_sift(I2);  
  
d1 = double(d1);  
d2 = double(d2);
```

f is a  $4 \times N$  feature frame, each column =  $[x, y, scale, orientation]$

d is a  $128 \times N$  descriptor, each column is an 128-D feature vector

# Extract SIFT Features

- Draw feature points with `plot_sift(img, f, d)`



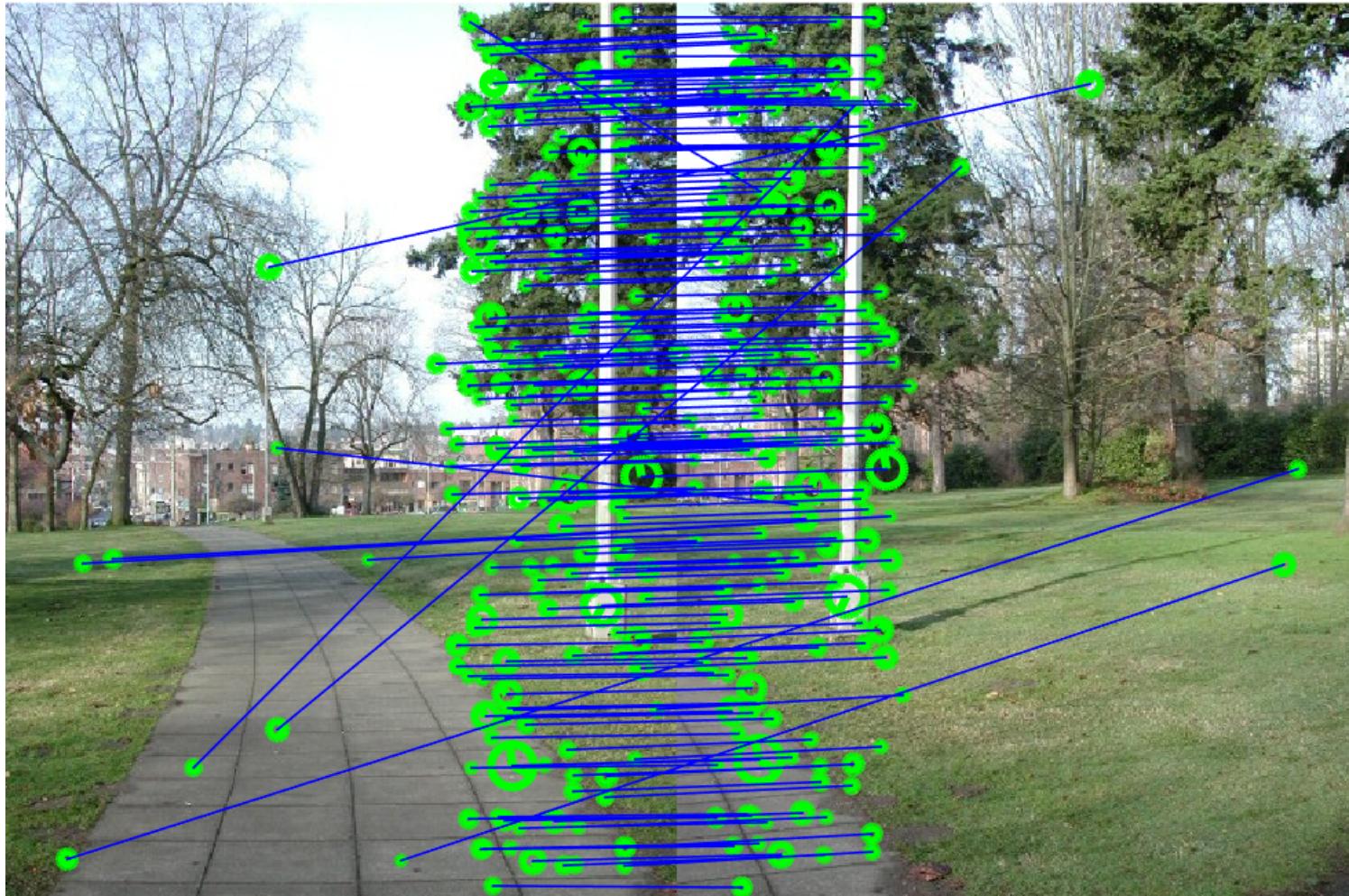
# 4 Steps for Image Stitching

---

1. Feature Extraction (SIFT or Harris)
2. **Feature Matching**
3. Image Alignment with RANSAC
4. Image Blending/Stitching

# Feature Matching

- Find matched features between two images



# Feature Matching

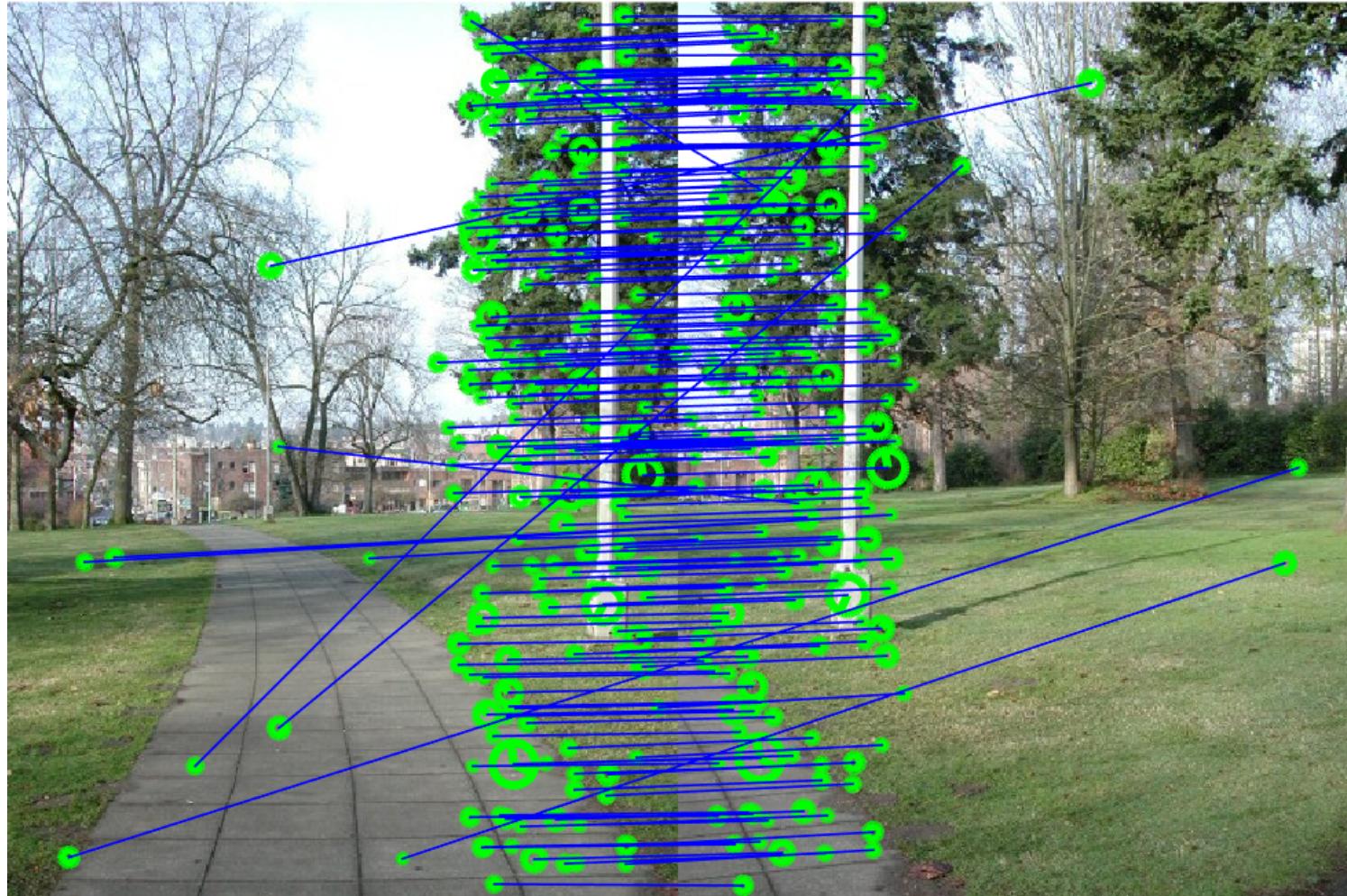
- Use `vl_ubcmatch()`

```
[matches, scores] = vl_ubcmatch(d1, d2) ;
```

- `matches` is a  $2 \times N_{match}$  matrix, each column indicates the matched index of `d1` and `d2`
  - `d1(:, matches(1, 1))` and `d2(:, matches(2, 1))` are the first matched pair
  - `d1(:, matches(1, k))` and `d2(:, matches(2, k))` are the  $k$ -th matched pair
- Plot matched points with `plot_match(img1, img2, f1, f2, matches)`

# Feature Matching

- `plot_match(img1, img2, f1, f2, matches)`



# 4 Steps for Image Stitching

---

1. Feature Extraction (SIFT or Harris)
2. Feature Matching
3. **Image Alignment with RANSAC**
4. Image Blending/Stitching

# Image Alignment

- Assume simple translation model:  $p_1 = p_2 + \begin{pmatrix} tx \\ ty \end{pmatrix}$



image 1

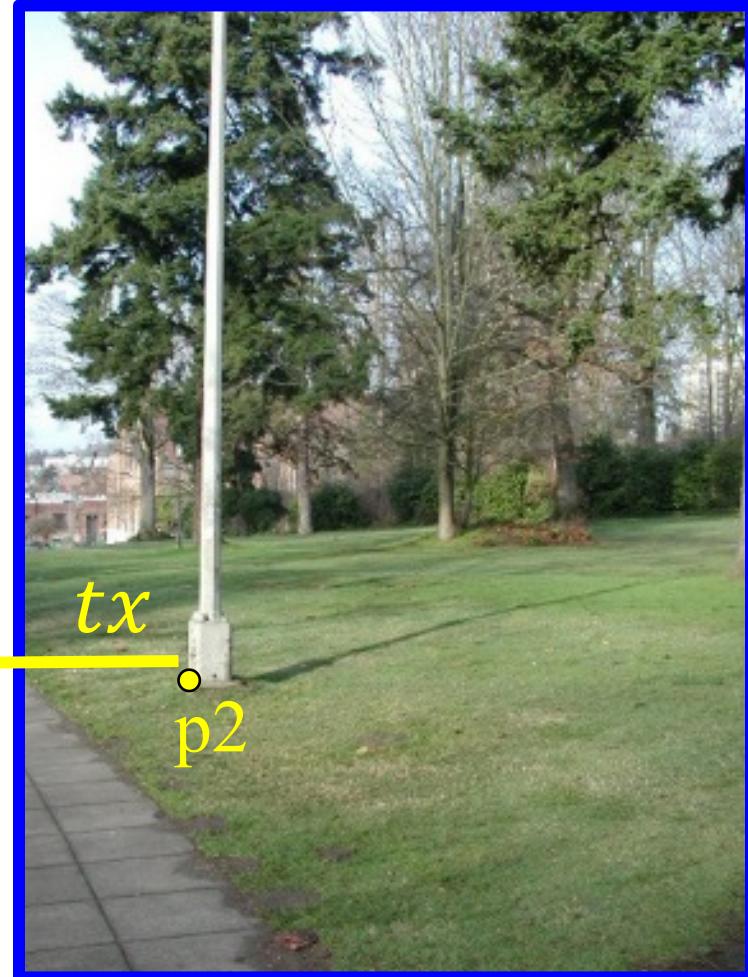


image 2

# Image Alignment

- Assume simple translation model:  $p1 = p2 + \begin{pmatrix} tx \\ ty \end{pmatrix}$

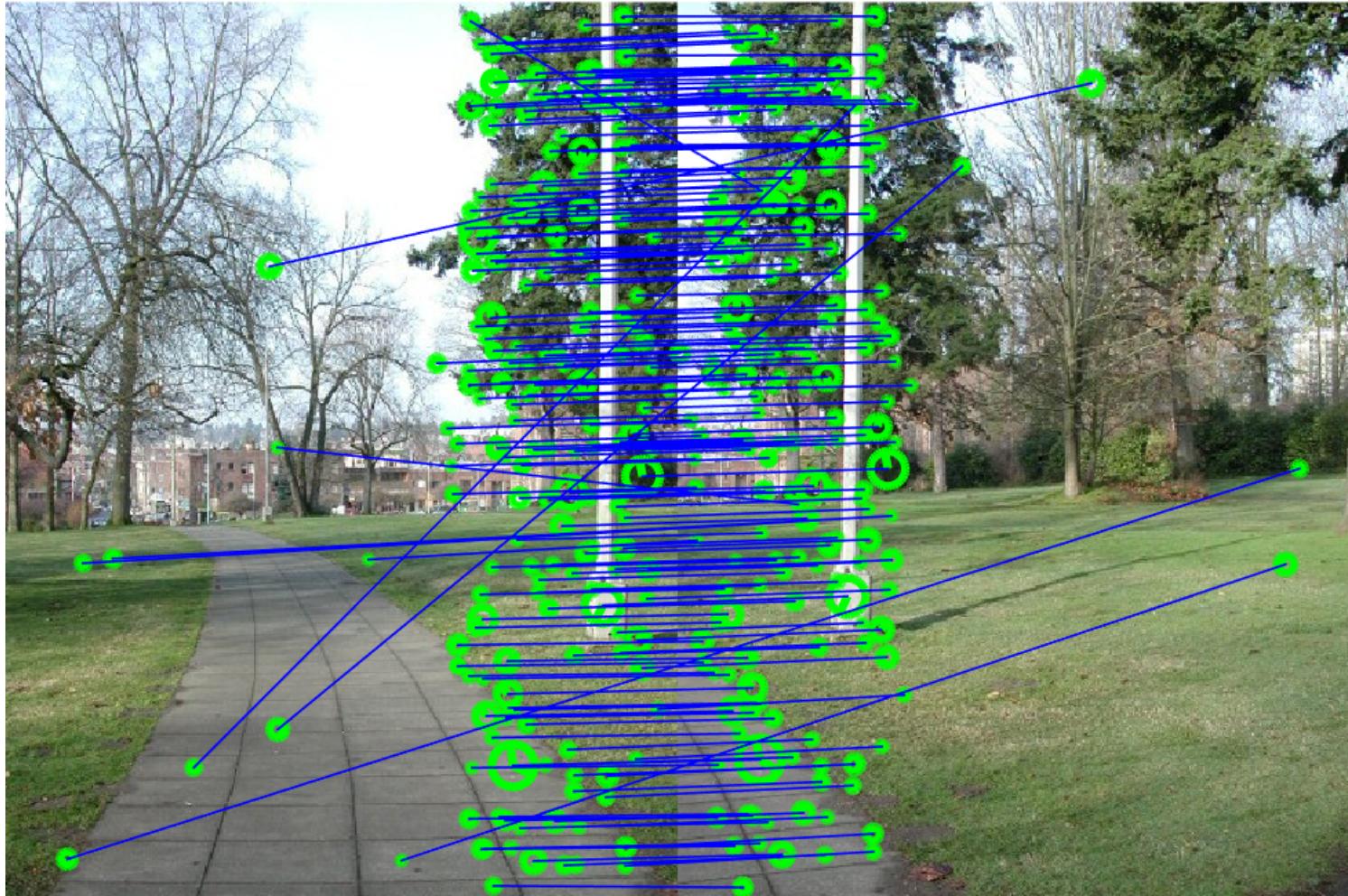


image 1

image 2

# Image Alignment

- Each matching pair can determine a pair of  $(tx, ty)$



# Image Alignment

- Use all matches pair to compute average  $(tx, ty)$ 
  - outliers cause large error



# Image Alignment

---

- Use RANSAC to select the best  $(tx, ty)$



# RANSAC

- Assume total  $N$  match pairs

Run  $k$  times:

1. randomly choose 1 pair (2 feature points)
2. compute  $tx\_0$  and  $ty\_0$
3. for other  $N-1$  pairs:
  - (a) compute  $tx\_1$  and  $ty\_1$
  - (b) if  $(tx\_1, ty\_1)$  is close to  $(tx\_0, ty\_0)$ ,  
count as an inlier

Return the pair  $(tx\_0, ty\_0)$  with the maximal  
#inliers

# How to determine $k$ ?

---

- $n$ : the number of selected points (in our case,  $n = 2$ )
- $e$ : the proportion of outliers
- $P$ : the probability of success rate after  $k$  trials

$$P = 1 - (1 - (1 - e)^n)^k$$

$$k = \frac{\log(1 - P)}{\log(1 - (1 - e)^n)}$$

- If  $n = 2$ ,  $e = 0.1$ ,  $P = 0.999$ , then  $k \cong 4$

# How to count inliers?

- Compute  $(tx, ty)$  from feature frame  $f$

```
p1 = f1(1:2, matches(1, i));  
p2 = f2(1:2, matches(2, i));
```

$f$  is a  $4 \times N$  feature frame, each column =  $[x, y, scale, orientation]$

```
tx = p1(1) - p2(1);  
ty = p1(2) - p2(2);
```

$\text{matches}$  is  $2 \times N_{\text{match}}$  matrix, each column =  $[f1_{\text{index}}, f2_{\text{index}}]$

- A pair  $(tx_1, ty_1)$  is inlier of  $(tx_0, ty_0)$  if

$$(tx_1 - tx_0)^2 + (ty_1 - ty_0)^2 < \delta$$

# RANSAC

- Assume total  $N$  match pairs

Run  $k$  times:

use `randperm()`

1. randomly choose 1 pair (2 feature points)
2. compute  $tx\_0$  and  $ty\_0$
3.  $\#inlier = 0$
4. for other  $N-1$  pairs:
  - (a) compute  $tx\_1$  and  $ty\_1$
  - (b) if  $(tx\_1-tx\_0)^2 + (ty\_1-ty\_0)^2 < \delta$ :  
 $\#inlier = \#inlier + 1$

```
if #inlier > max_inlier:  
    best_tx = tx_0  
    best_ty = ty_0
```

Return `(best_tx, best_ty)`

# 4 Steps for Image Stitching

---

1. Feature Extraction (SIFT or Harris)
2. Feature Matching
3. Image Alignment with RANSAC
4. **Image Blending/Stitching**

# Image Blending/Stitching

- Fix image 1, and shift image 2 by  $(tx, ty)$



image 1

image 2

# Image Blending/Stitching

---

- Simply paste image 2 over image 1



# Image Blending/Stitching

- Fix image 1, and shift image 2 by  $(tx, ty)$

```
output = zeros(H + ty, W + tx, 3);
output(1:H, 1:W, :) = img1;

for y2 = 1:size(img2, 1)
    for x2 = 1:size(img2, 2)

        y1 = y2 + ty;
        x1 = x2 + tx;

        if( y1 >= 1 && y1 <= H + ty &&
            x1 >= 1 && x1 <= W + tx )
            output(y1, x1, :) = img2(y2, x2, :);
    end

end
end
```

# Assignment09

---

- Implement lab09.m
  - use vlfeat for feature extraction and feature matching
  - implement RANSAC
- Adjust  $e$  and  $P$  to see the difference on  $k$  and stitching results.
- Use image pair ‘prtn12.jpg’ and ‘prtn13.jpg’.
- Use parameters  $e = 0.2$ ,  $s = 2$ ,  $p = 0.999$ ;
- Upload the result and **lab09.m**