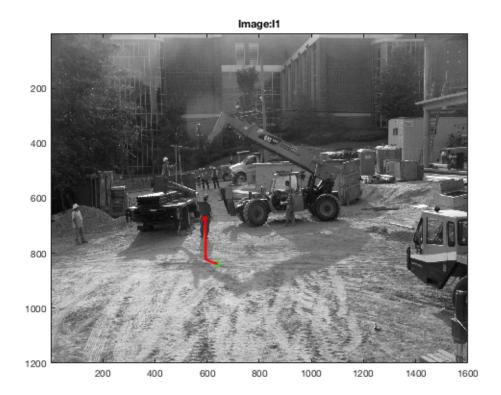
# Image: I1 findTemplate\_I1\_temp2.m

Template: temp2
Parameters:

(c=0.000024,nsteps=200,pos=[640,840])
Final Point:[586.983502,672.585061]





# loseTemplate\_I1\_temp2.m

Template: temp2
Parameters:

Image: I1

(c=0.000024,nsteps=200,pos=[520,600]) Final Point:[519.446550,543.022444]



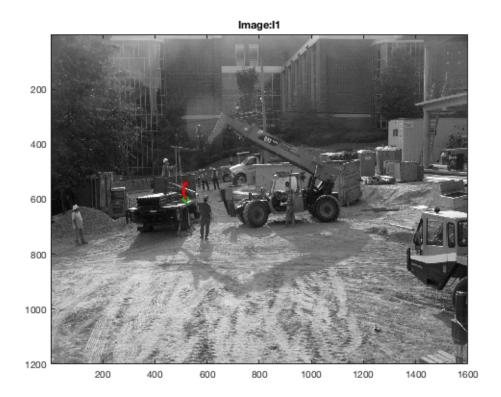


Image: I2
Template: temp2
findTemplate\_I2\_temp2.m

Template: temp2
Parameters:

(c=0.000024,nsteps=200,pos=[780,690])
Final Point:[707.141200,628.116144]



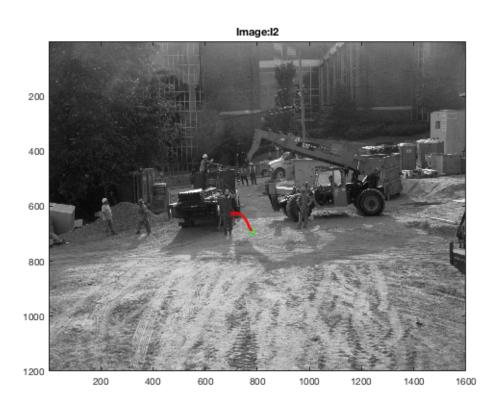


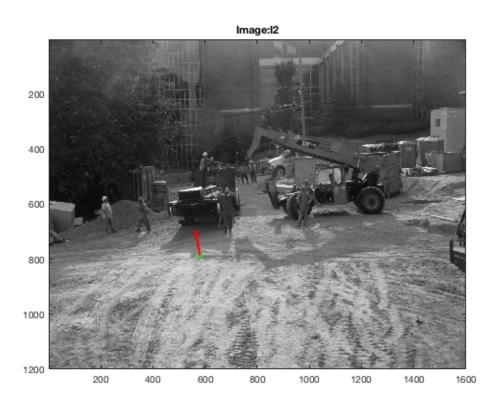
Image: I2
Template: temp2

# loseTemplate\_I2\_temp2.m

Parameters:

(c=0.000024,nsteps=200,pos=[580,790]) Final Point:[564.441330,701.957442]





(c=0.000025,nsteps=200,pos=[520,790]) Final Point:[574.472192,715.056294]

# findTemplate\_I3\_temp1.m





(c=0.000025,nsteps=200,pos=[620,790]) Final Point:[676.103055,809.589928]

# loseTemplate\_I3\_temp1.m





```
용
용
  dPos = gradTempMatchQA(template, image, ipos)
용
용
 Computes the gradient of the template matching score for image matching
  via gradient descent. The current estimate of the template position
  is required. Uses interpolation to do sub-pixel matching on the image
  data, but does not do so for the image gradient information.
용
용
  Input:
                     - the grayscale template patch.
용
   template
용
  image
                            - the image to find the template in.
용
  ipos
                            - the position of the template (centered).
                     the position is in (x,y) coordinates.
용
용
용
 Output:
  dPos
                            - the differential with respect to position.
9
   pCost
                            - the current matching score for ipos (optional).
용
용
용
  Name:
             tmatchT
용
ુ
            Patricio A. Vela, pvela@ece.gatech.edu
 Author:
용
                     2012/02/18
% Created:
% Modified:
              2012/02/18
function [dPos, pCost] = gradTempMatchQA(template, image, ipos)
%==[1] Get the image and template dimensions.
ti = size(template,1);
                                          % Number of rows (size in y dir)
tj = size(template,2);
                                          % Number of cols (size in x dir)
[iM, iN] = size(image);
                                          % Image size rows x cols (y size, x size)
%==[2] Get image patch at the specified location. Make the patch bigger
              by one pixel all around in order to compute the gradients.
if (rem(tj,2) == 0)
                                          % Even size template in x direction.
 xinds = ipos(1) + [(-0.5-tj/2):(tj/2+0.5)];
else
 xinds = ipos(1) + [-(1+(tj-1)/2):(1+(tj-1)/2)];
end
if (rem(ti,2) == 0)
                                          % Even size template in y direction.
 yinds = ipos(2) + [(0.5-ti/2):(ti/2-0.5)];
else
 yinds = ipos(2) + [-(1+(ti-1)/2):(1+(ti-1)/2)];
end
%==[3] Extract the image data from the specified location, plus compute
             the image gradients.
imdat = interp2(1:iN, (1:iM)', image, xinds, yinds');
[gradIx, gradIy] = gradient(imdat);
```

```
%==[4] Compute the gradient (compensating for the extra border).
tdiff = (template - imdat(2:end-1,2:end-1));
%-- Form the G matrix
             a 2x2 matrix from the image gradients.
gradItempX = gradIx(2:end-1,2:end-1);
gradItempY = gradIy(2:end-1,2:end-1);
gradI = [gradItempX(:), gradItempY(:)];
G = gradI' * gradI;
%-- Form the E vector
               a 2x1 vector from the image gradient and matching error.
E = (gradI' * tdiff(:));
%-- Solve for the position update.
              computes the solution to the local quadratic approximation at ipos.
dPos = (G \setminus E);
%==[5] Compute the cost. We have it almost computed, so this is an almost
     free computation.
pCost = sum(tdiff(:).*tdiff(:));
end
```

```
Not enough input arguments.

Error in gradTempMatchQA (line 35)

ti = size(template,1); % Number of rows (size in y dir)
```

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The convergence for the quadratic approach happens much quicker than the linear approach, since there is no need to use the c factor to take small steps. In other words, the approach allows the descent to jump to the local minimum of the quadratic approximation in one step.

The basin of attraction was similar, since similar blurring standard deviations were used for both approaches.

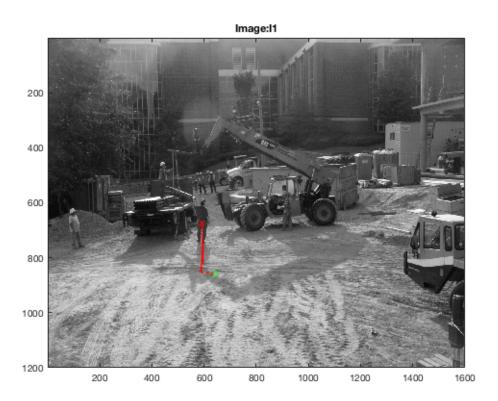
(nsteps=50,pos=[640,860])

Final Point:[592.967259,673.053170]

# findTemplateQA\_I1\_temp2.m

### Template:temp2





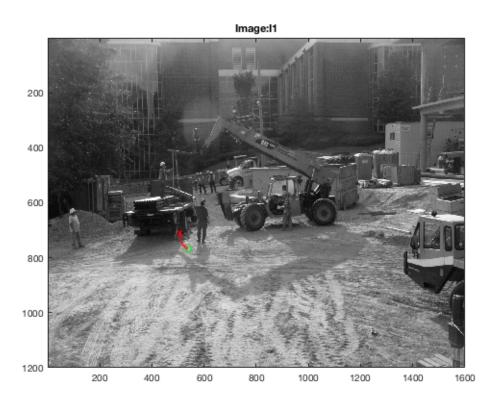
(nsteps=50,pos=[540,770])

Final Point:[507.319135,709.341531]

# loseTemplateQA\_I1\_temp2.m

### Template:temp2





(nsteps=50,pos=[640,860])

Final Point:[685.652961,670.868350]

# findTemplateQA\_I2\_temp2.m

### Template:temp2



### Image:l2



(nsteps=50,pos=[570,730])

Final Point:[564.495365,701.928916]

# loseTemplateQA\_I2\_temp2.m

### Template:temp2



### Image:I2



(nsteps=50,pos=[480,820])

Final Point:[574.484935,715.115290]

# findTemplateQA\_I3\_temp1.m

### Template:temp1





(nsteps=50,pos=[620,790])

Final Point:[676.103055,809.589928]

# loseTemplateQA\_I2\_temp2.m



