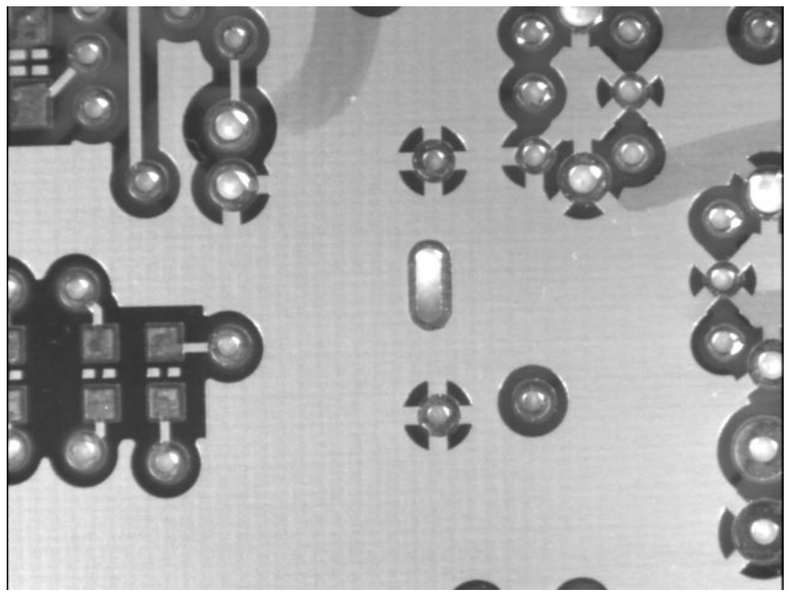
The Problem

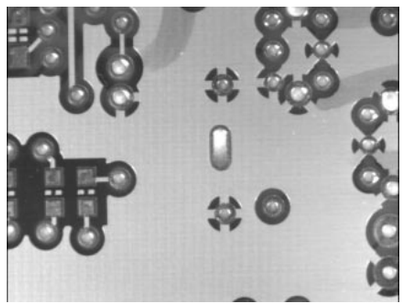
- **Need fast (~10–50 mS) search**
- **Arbitrary Patterns**
- **Non-specialized hardware**
- **Normalized greyscale correlation too slow (or is it?)**
- **Search for target** $M(x, y)$ **in image** $I(x, y)$

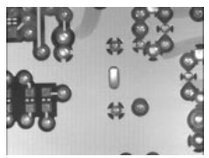
The Approach

- **Manage complexity by using pyramid image representation: speed-up by factor of** $2^{4(k_{\max}-1)}$
  - **Do cheap search at coarse scale, refine at each level of the pyramid**
  - **Use estimates of correlation gradient to guide search at each level**
  - **Keep "good" matches at finest level**

Building the Pyramid





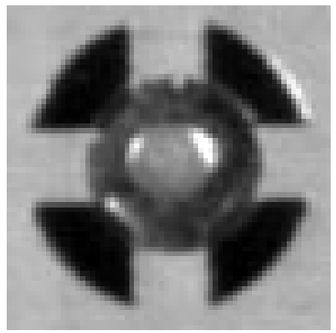


- Image dimensions decrease by factor of 2 at each level
- Each pixel in level *k* is the average of 4 pixels in level *k-1*
- Can be built just using add and shift
- $k_{\max} \leq \log_2 \min(I_w, I_h)$

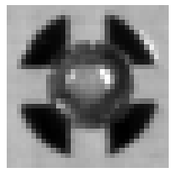
Tuning the Pyramid

- How many levels to use in pyramid?
- Can build pyramid "too deep" (1×1 at top? Pathological?)
- Use analysis of target to decide: "worst case" scenario
  - Construct "maximum height" pyramids for target and shifted versions of target
  - Use correlation between original and shifted versions to determine worst possible scores
  - Use scores to limit pyramid depth

Target Representation



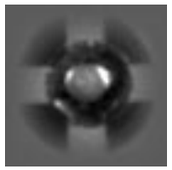
Original Target



Target Pyramid Level 1

Target Pyramid Level 2

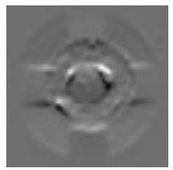
Model



Windowed Model

Window Function

X Component of Gradient



Y Component of Gradient

Correlation Gradient

- **At each level the location estimate is refined using steepest descent search (4 terms to compute)**

$$C(u,v)=\frac{\iint I(x,y)(MW)(x-u,y-v)dx\,dy}{\sqrt{\iint I^2(x,y)W(x-u,y-v)dx\,dy}}$$

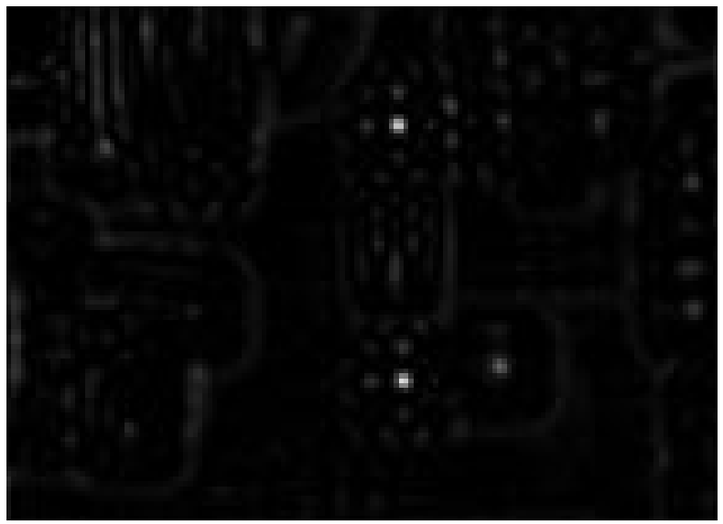
$$\nabla C^2(u,v)=-2\frac{\iint I(x,y)(MW)(x-u,y-v)dx\,dy}{\iint I^2(x,y)W(x-u,y-v)dx\,dy}\times\iint I(x,y)\nabla(MW)(x-u,y-v)dx\,dy$$

$$+\left[\frac{\iint I(x,y)(MW)(x-u,y-v)dx\,dy}{\iint I^2(x,y)W(x-u,y-v)dx\,dy}\right]^2\iint I(x,y)\nabla W(x-u,y-v)dx\,dy$$

The Search

- Do "inexpensive" correlation at top-level of pyramid
- Identify "candidates" from peaks in top-level correlation
  - For each candidate, descend through pyramid (coarse-to-fine), revising location estimate
  - Use correlation gradient estimate to minimize steps at each level
  - Reject candidates whose score drop below threshold

Top Level Correlation



Top Level
Correlation Surface

- Inexpensive due to reduced image size at top of pyramid
- May lead to false matches, so call peaks "candidates"
- Inhibit candidate peaks after use

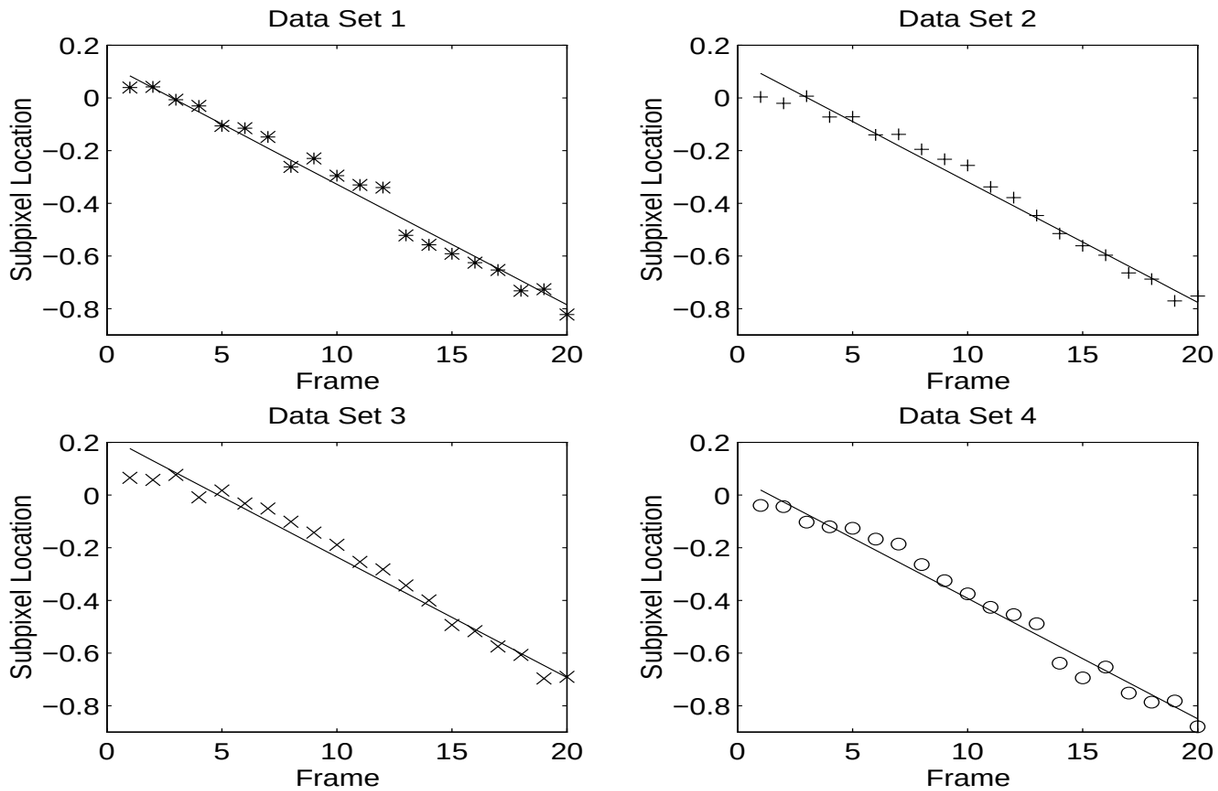
Subpixel Localization

- We can locate a target with better than 1 pixel resolution
- Model correlation surface in 3x3 neighbourhood with a biquadratic surface: $\hat{C}(x,y)=\alpha x^2+\beta y^2+\gamma x+\delta y+\xi xy+\varphi$
- Fit model parameters via least squares (cheap/fast)
- Solve for maximum:

$$\frac{\partial \hat{C}}{\partial x}=0 \Rightarrow 2\alpha x_c+\gamma+\xi y_c=0 \qquad \frac{\partial \hat{C}}{\partial y}=0 \Rightarrow 2\beta y_c+\delta+\xi x_c=0$$

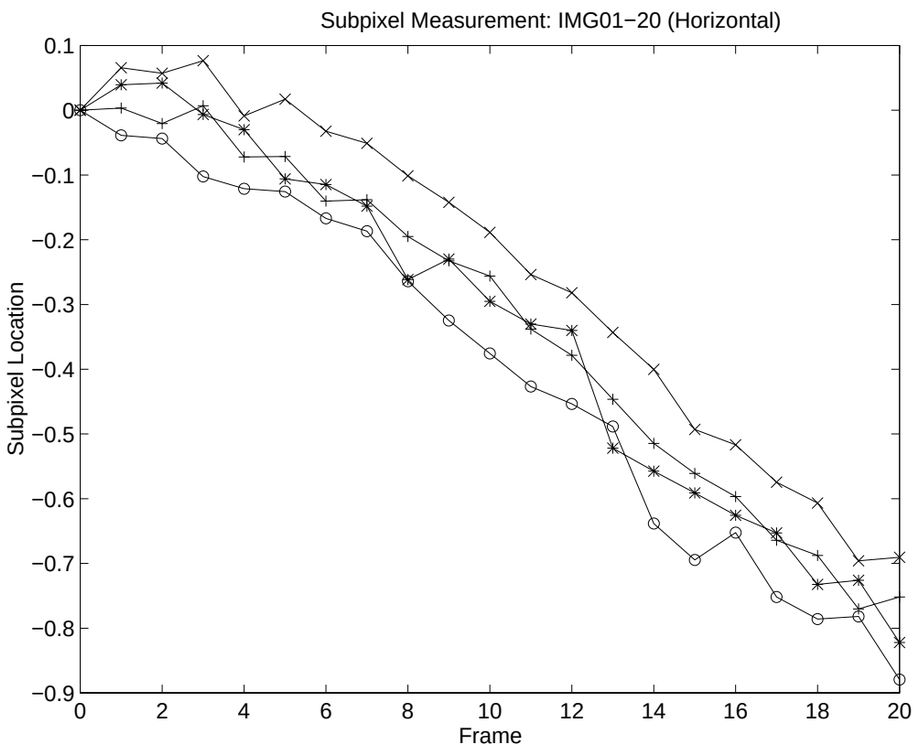
- Simple computation, much done offline in advance

•Subpixel Results

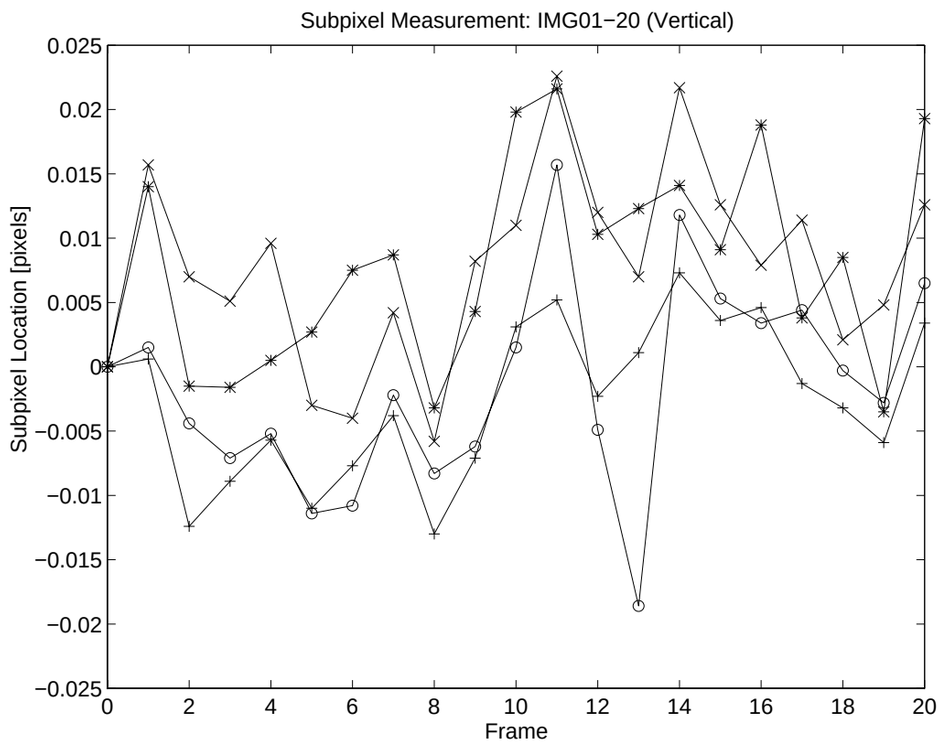


- Search region moved in approximately 1/20th pixel increments horizontally with camera stationary
- Target detected with subpixel localization, results plotted

Subpixel Results

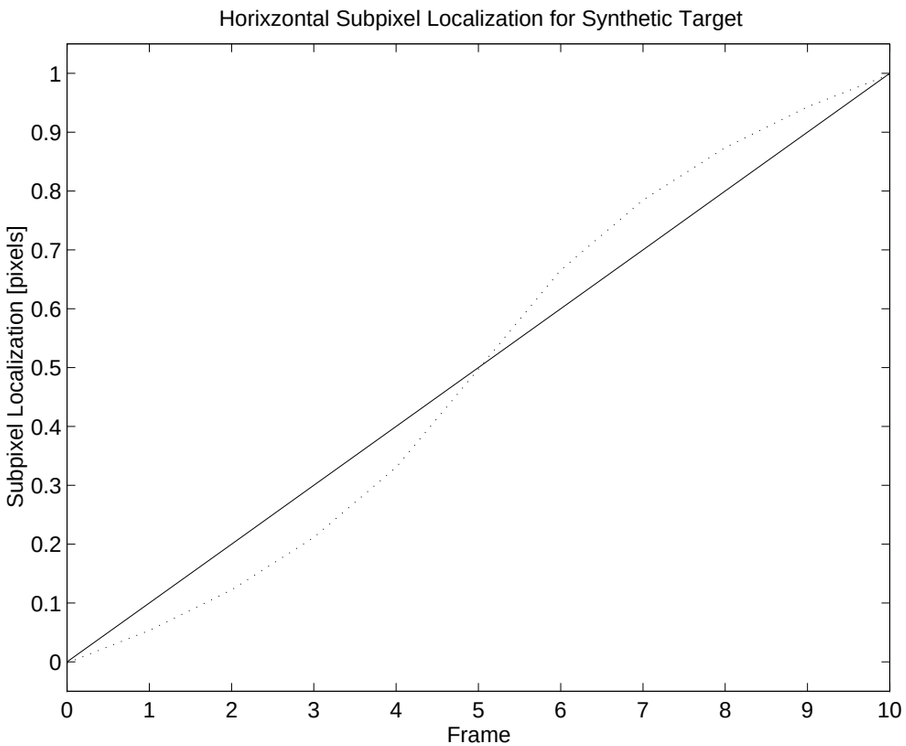


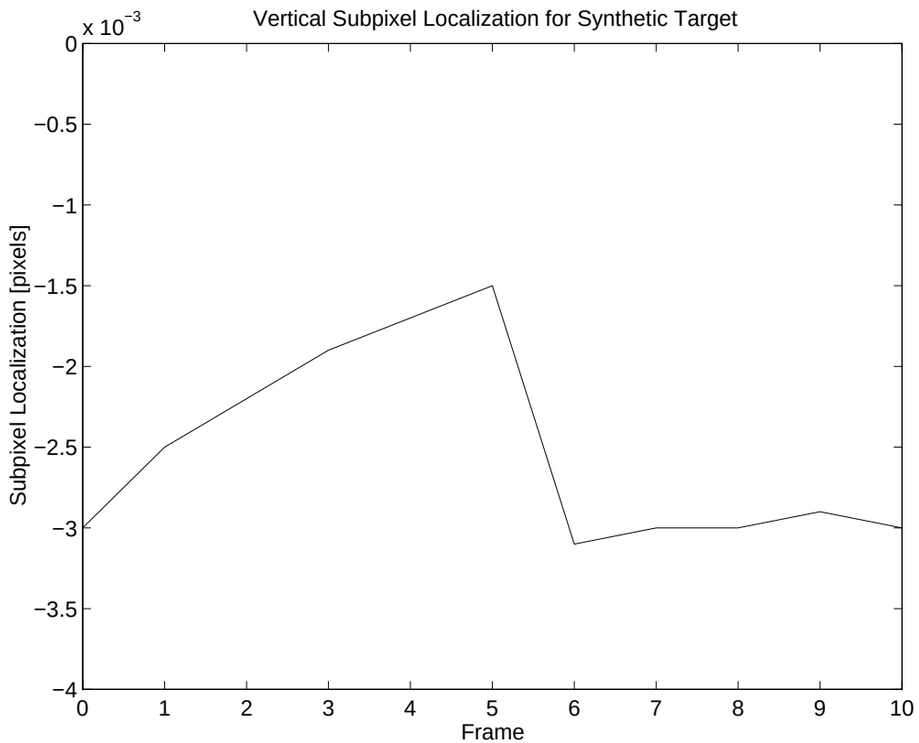
Composite of Horizontal Data



Composite of Vertical Data

Subpixel Results: Synthetic Data





Results

- **10 to 50 mS search times on 200 MHz PC running Windows NT, 64MB RAM**
- **Testing with accept threshold set to 0.95:**
  - **False positives typically less than 0.1% (varies with accept threshold)**
  - **False negative rate similar (rises if accept threshold set too high)**
- **Search time rises if many candidates**
- **Larger images have similar search times; pyramid just gets deeper**

Applications

- **Machine Vision inspection & registration**
- **Searching for images on the Internet (insensitive to watermark removal)**
  - **Many images to search, so speed important!**
  - **Searching for "abused images" (defaced trademarks) on Internet**

Future Directions

- Size invariance (pyramid representation well suited to this)
- Rotation invariance
- Occlusion
- Don't care pixels

Conclusions

- **Novel use of pyramid representation in correlation image search**
- **Use of "worst case" analysis to decide depth of target pyramid representation**
- **Use of correlation gradient estimate to guide location refinement at each level of pyramid**
- **Implementation of simple but effective subpixel localization**