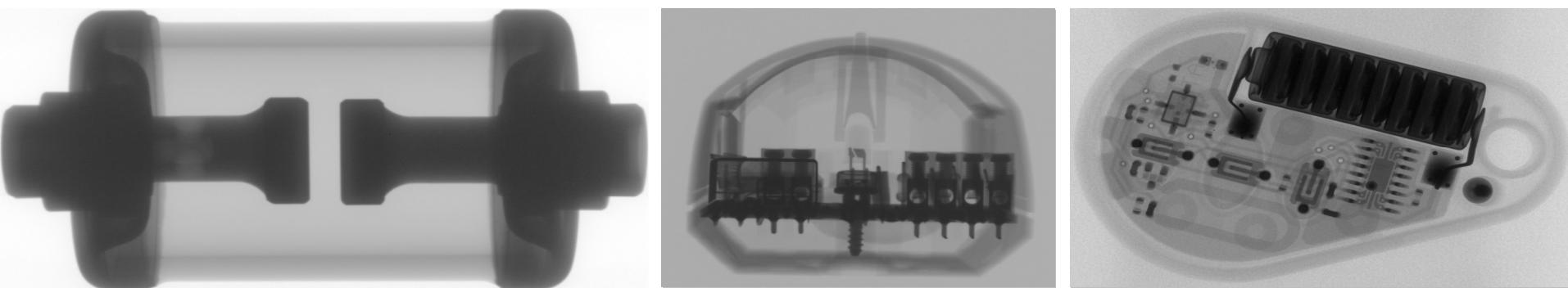


Exceptional service in the national interest



Exploration of Available Feature Detection and Recognition Systems and their Performance on Radiographs

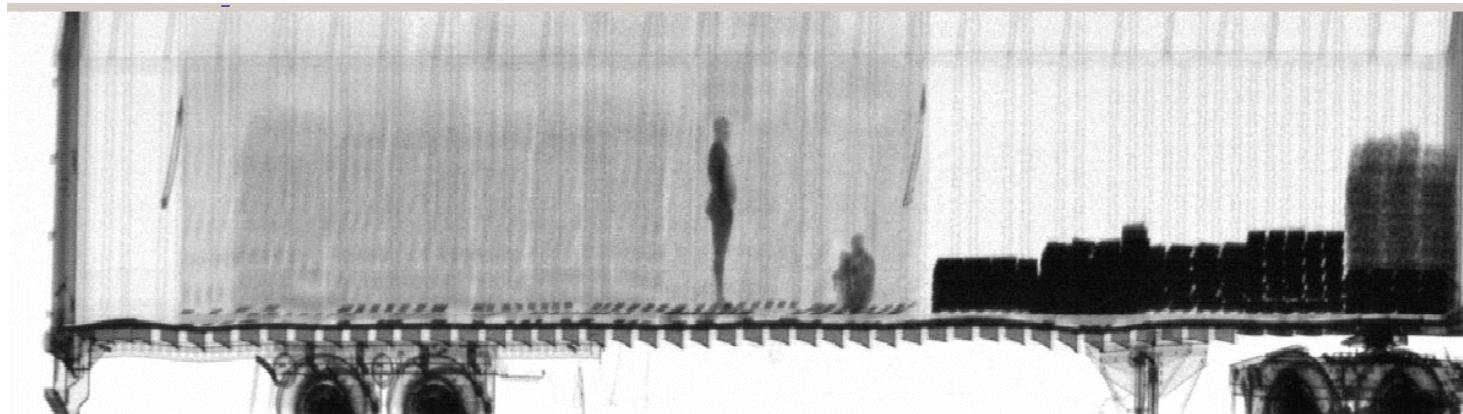
Andrew C. Wantuch, Joshua A. Vita, Edward S. Jimenez, and Iliana E. Bray
Sandia National Laboratories, Software Systems R&D



Sandia National Laboratories is a multi-program laboratory managed and operated by Sandia Corporation, a wholly owned subsidiary of Lockheed Martin Corporation, for the U.S. Department of Energy's National Nuclear Security Administration under contract DE-AC04-94AL85000. SAND NO. 2011-XXXXP

Goal

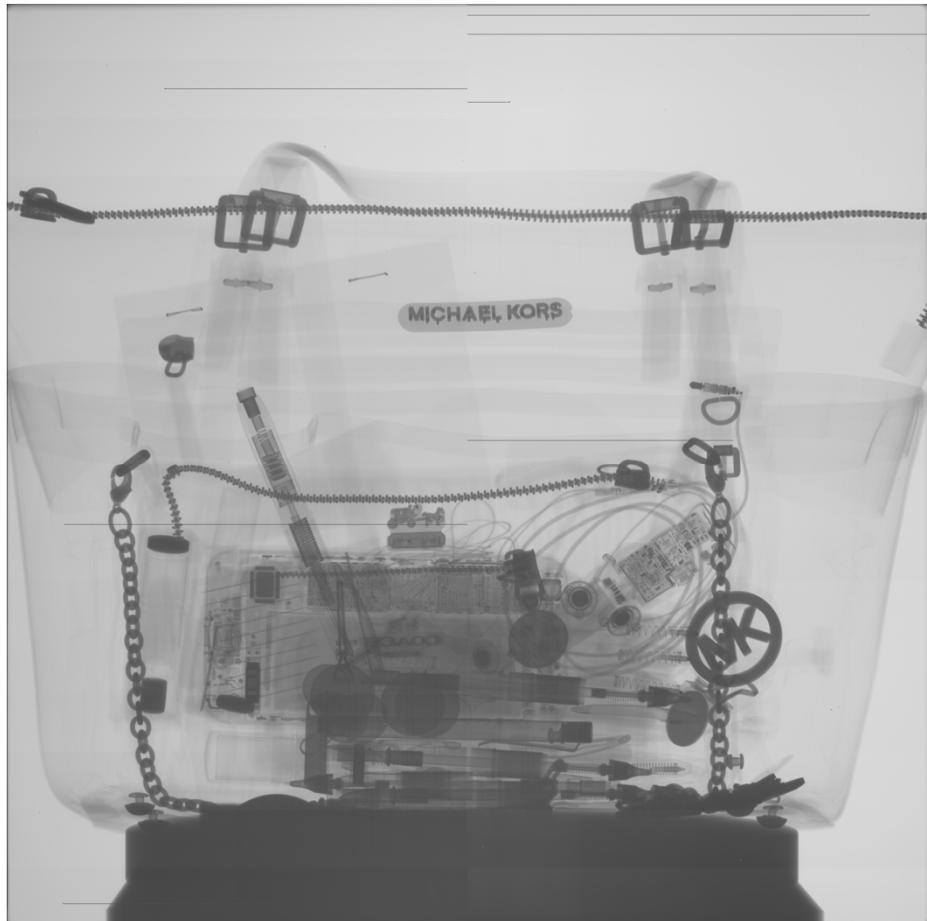
- Image recognition using radiographs captured in the wild.
- Potential applications include:
 - Checkpoint Security
 - Component Recognition
 - Nondestructive Evaluation
 - Reverse Engineering



Source: https://upload.wikimedia.org/wikipedia/commons/d/d7/VACIS_Gamma-ray_Image_with_stowaways.GIF

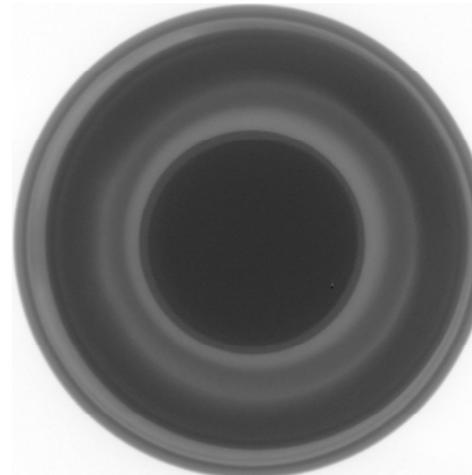
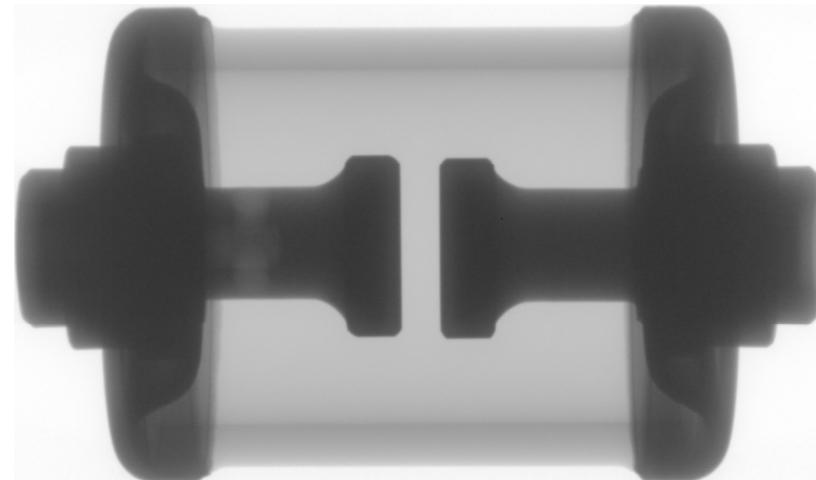
Challenge - Radiographs

- Object Recognition in X-rays
 - Occlusion
 - Perspective
 - Noise



Challenge - Radiographs

- Object Recognition in X-rays
 - Occlusion
 - **Perspective**
 - Noise



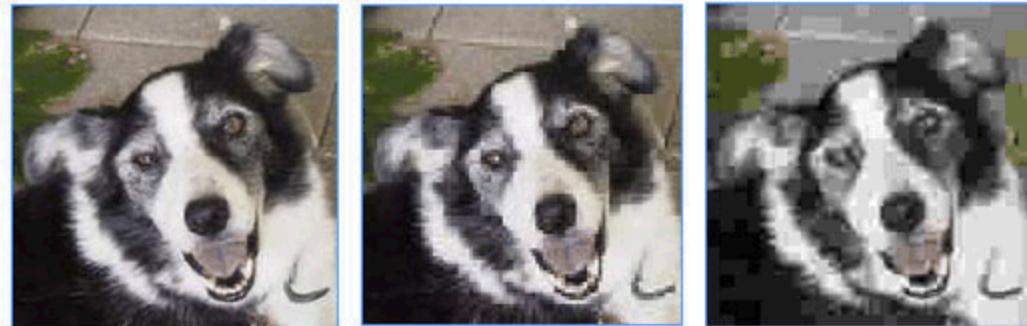
Challenge - Radiographs

- Object Recognition in X-rays
 - Occlusion
 - Perspective
 - **Noise**



Challenge - Database

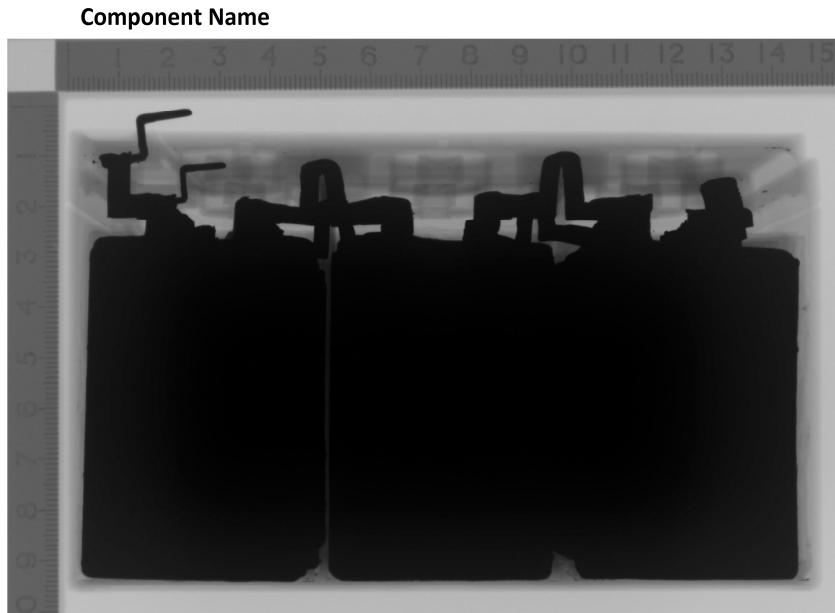
- Database Constraints
 - **JPEG images - lossy compression**
 - Inconsistent/insufficient energy
 - Unclean/modified images
 - Limited/inconsistent views
 - Similar Objects
 - Subcomponents
 - Useless Images



- Image source: <http://cdn.mos.cms.futurecdn.net/26db74f7c66961368a34408c33f00f18.jpg>

Challenge - Database

- Database Constraints
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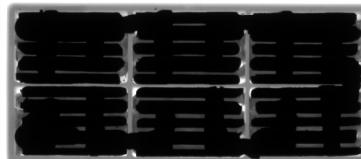
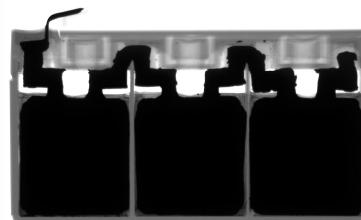


Challenge - Database

■ Database Constraints

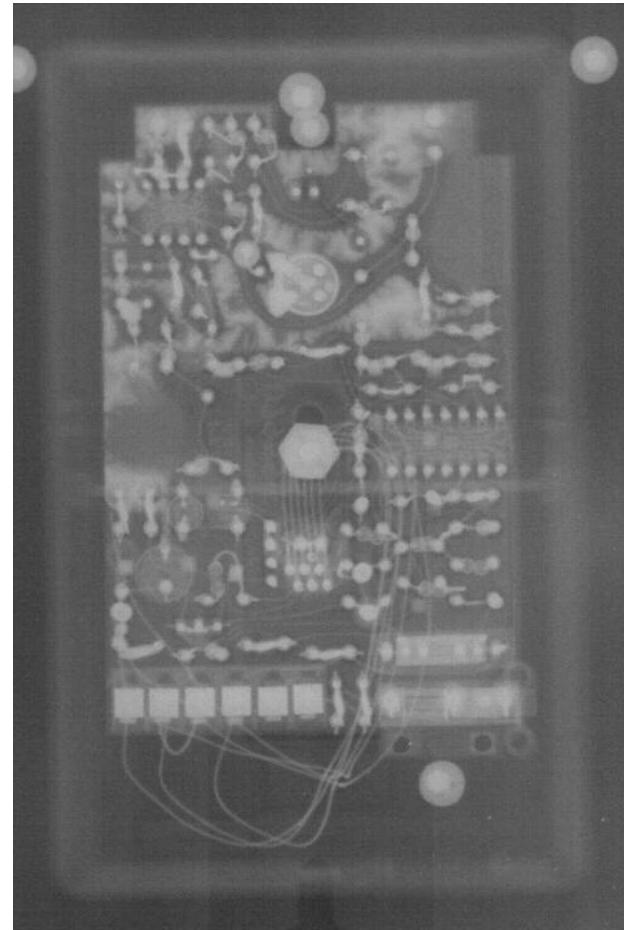
- JPEG images - lossy compression
- Inconsistent/insufficient energy
- **Unclean/modified images**
- Limited/inconsistent views
- Similar Objects
- Subcomponents
- Useless Images

Component Name



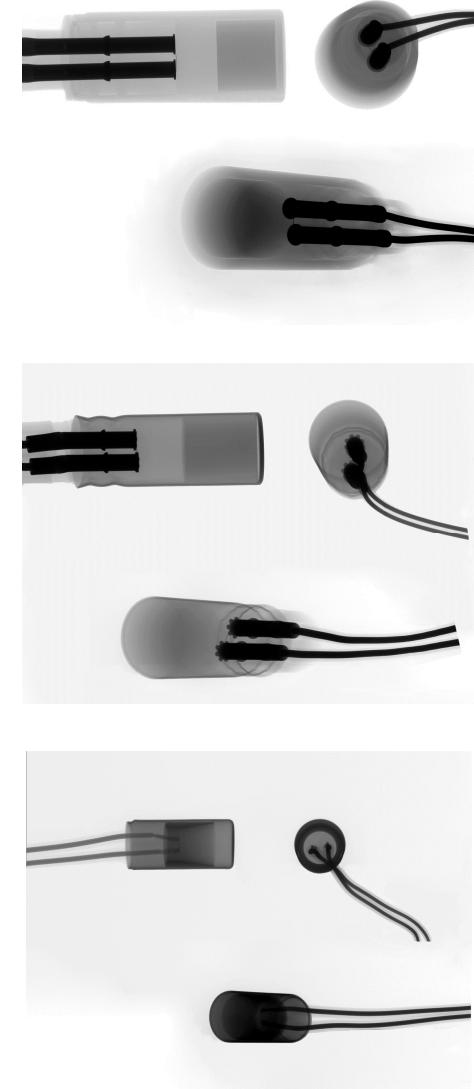
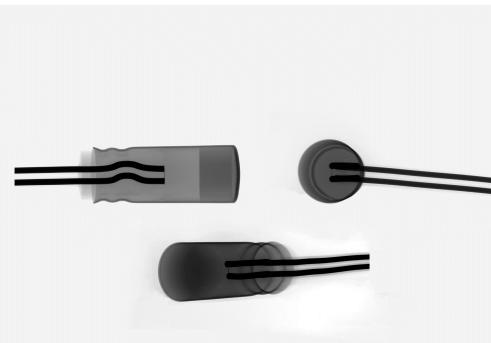
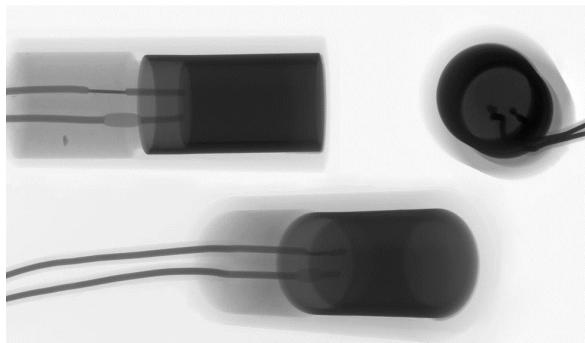
Challenge - Database

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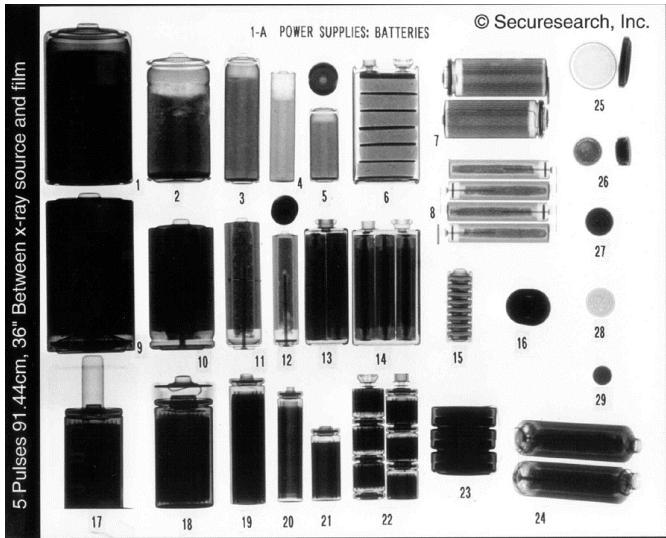
Challenge - Database

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Challenge - Database

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Challenge - Database

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 - Similar Objects
 - Subcomponents
 - **Useless Images**



So why THIS Database?

- Contains images of certain types of components
- ~17K images
- Metadata
 - Type
 - Main Material
- Continually updated

Background

- Bray et al.
 - “Exploring the feasibility of traditional image querying tasks for industrial radiographs” (2015)
 - Layering approach
- David G. Lowe
 - “Distinctive image features from scale-invariant keypoints” (2004)
 - Scale-invariant feature transform (SIFT)
 - Scale, Noise, Illumination
- Bay et al.
 - “SURF: Speeded Up Robust Features” (2007)
 - Faster than SIFT, but less accurate

SIFT

- Scale invariant feature transform
- Interest points
 - Convolved with Gaussian filters at various scales (expensive)
 - Minima/maxima Difference of Gaussians at different scales define points
- Advantages
 - Identify objects in clutter
 - Partial Occlusion
- Disadvantage
 - Very slow due to computational complexity
 - Performance degradation with blurring

■ Source: <http://artofthehome.com/wp-content/uploads/2011/04/sifting-flour.jpg>



Surf

- Speeded Up Robust Features (Bay et. Al.)
- Based on SIFT
- Generally several times faster than SIFT
- Three main parts:
 - Interest Point Detection
 - Local Neighborhood Description
 - Matching



- Source: http://www.surfingindia.net/files/surf-files/u9/India_Surf_Tours_-_17__1_.jpg

- Square filters
 - Approximate Gaussian Smoothing
 - Detect Scale Invariant Feature Points
- Blob Detector
 - Interest points found at various scales
 - Hessian Matrix to get points of interest
 - At point $p=(i,j)$, the Hessian at p and scale σ

$$H(p, \sigma) = \begin{bmatrix} L_{ii}(p, \sigma) & L_{ij}(p, \sigma) \\ L_{ij}(p, \sigma) & L_{jj}(p, \sigma) \end{bmatrix}$$

Slide 17

JESJ1 Jimenez, Edward Steven Jr, 8/24/2016

JESJ2 Jimenez, Edward Steven Jr, 8/24/2016

- Local Neighborhood Descriptor

- Provides unique and robust description of an image feature
- Balance between computational complexity and robustness/accuracy
- Based on the sum of Haar Wavelet Responses

- Matching

- Compare Descriptors from different images
- Matching pairs → Matching Features

Harris-Stephens Algorithm

- Corner Detection Algorithm
- Less focused on extracting image features, but rather detecting edges
- Corners are defined as interest points the weighted sum of squared differences is high
- Not to be confused with Laplace-Harris which is more of a blob detection algorithm
- Tested as many objects in the database are rectangular in shape
- Sharp edges present in database as well
- Will fail on circular objects?

MSER

- Maximally Stable Extremal Regions
- Blob Detection algorithm
 - Find similarities from images with different viewpoints
- Advantages
 - Invariance to affine transformations (objects wrt position of image acquired)
 - Multi-Scale Detection (x-ray magnification)
 - Scale and Rotation invariant
- Disadvantage
 - Blurring (not relevant for our application)

Approach

- Compare performance of out-of-the-box algorithms
 - Time
 - Accuracy
- Matlab
 - Computer Vision Toolbox
 - SURF
 - MSER
 - Harris
 - VLFeat
 - SIFT
- Python
 - OpenCV
 - SIFT

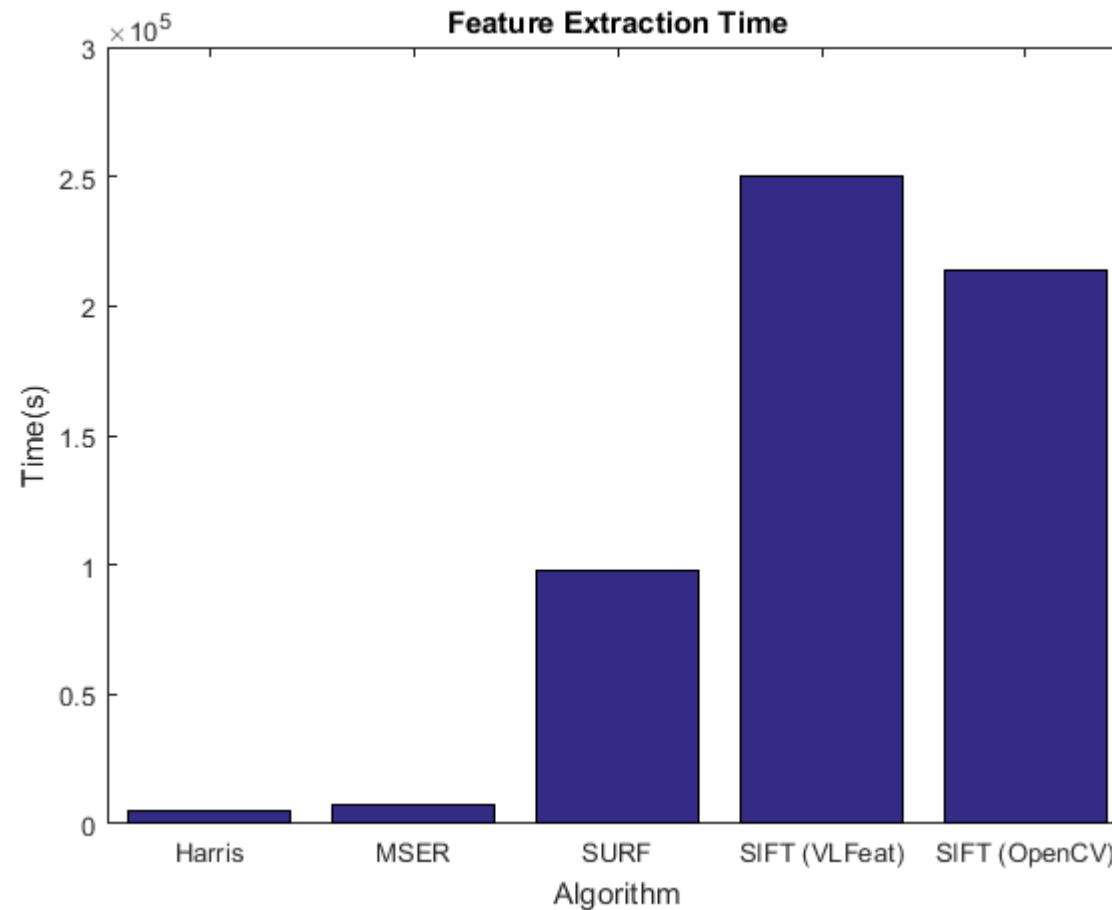
Implementation

- Running on a Dell PowerEdge R920
 - 4 Intel Xeon E7-4820 v2 CPUs at 2.0 GHz (64 Cores Total)
 - 512 GB DDR3 RAM
- Matlab version R2016a
 - Computer Vision System Toolbox
 - Image Processing Toolbox
 - VLFeat version 0.9.20
- Python version **3.5.2**
 - OpenCV version 2.4.13
- Used 3250 of ~17K images
 - 325 objects with 10 images each

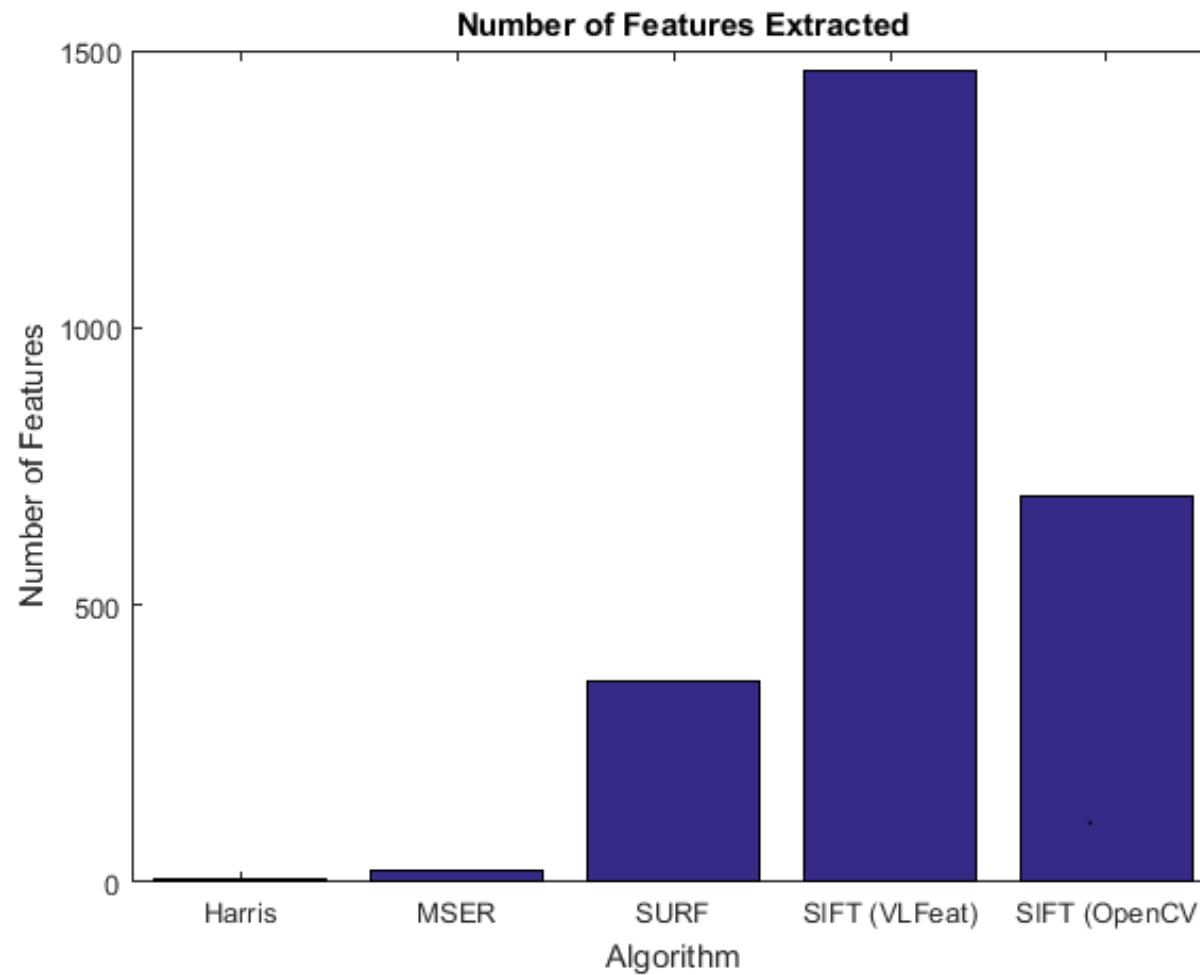


Source: http://en.community.dell.com/cfs-file/__key/communityserver-wikis-components-files/00-00-00-01-62/PowerEdgeR920.png

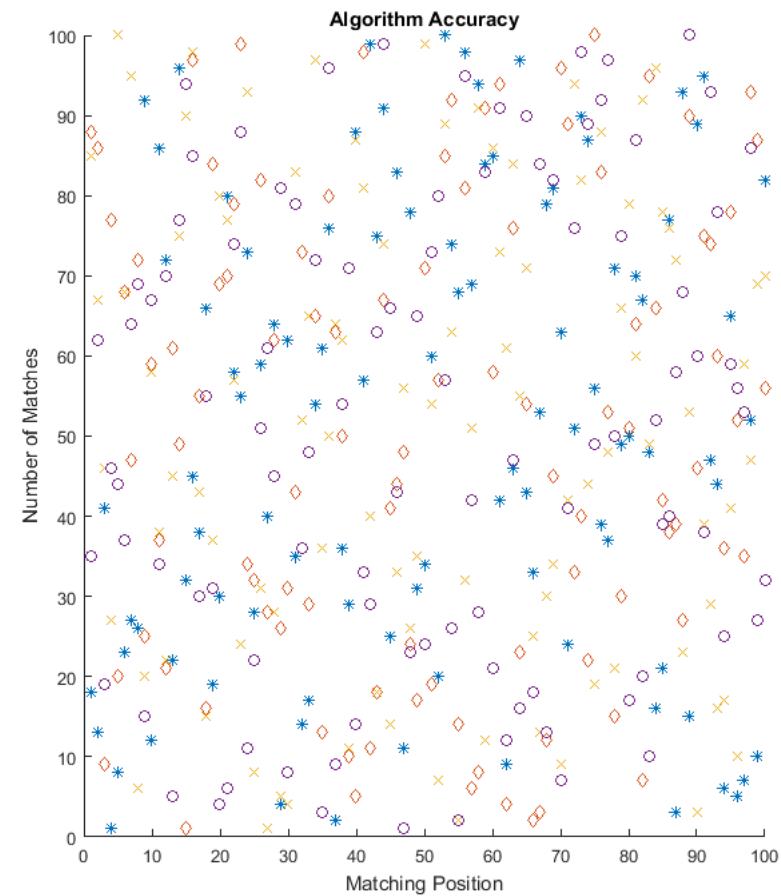
Results



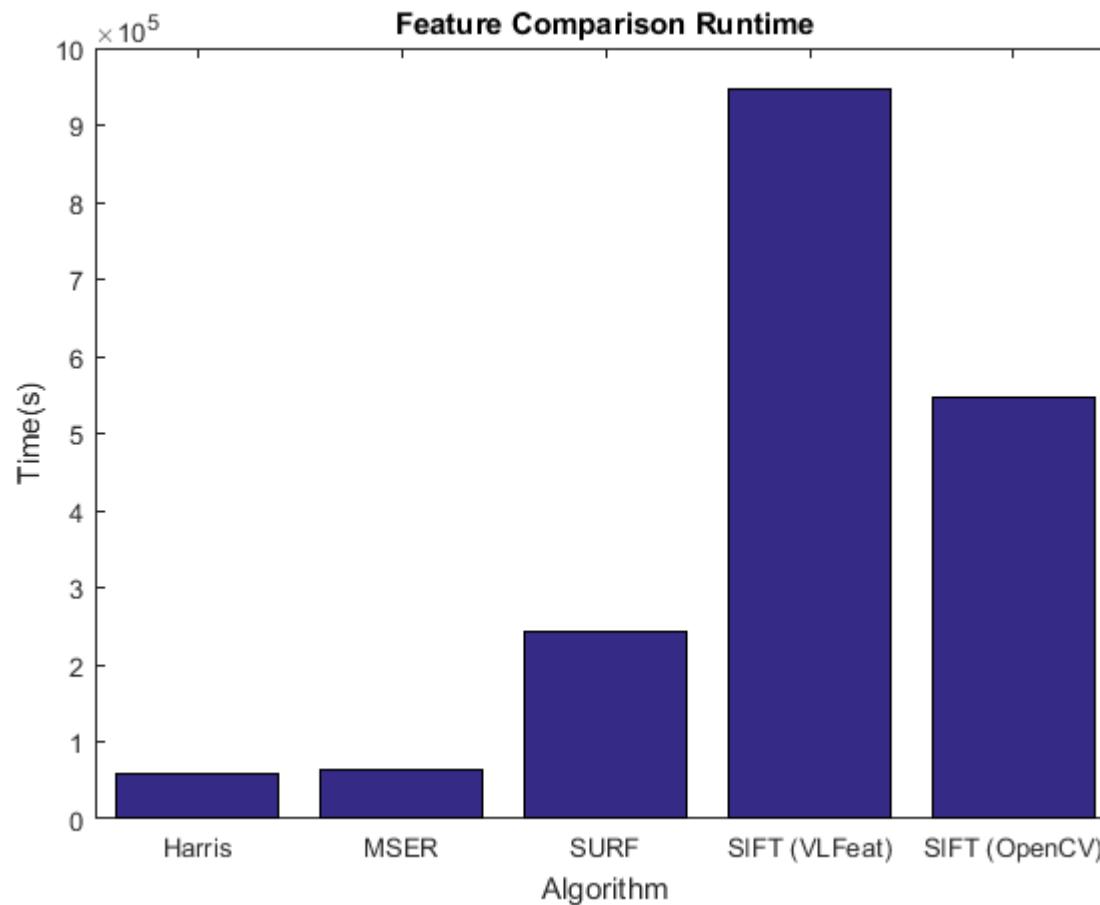
Results



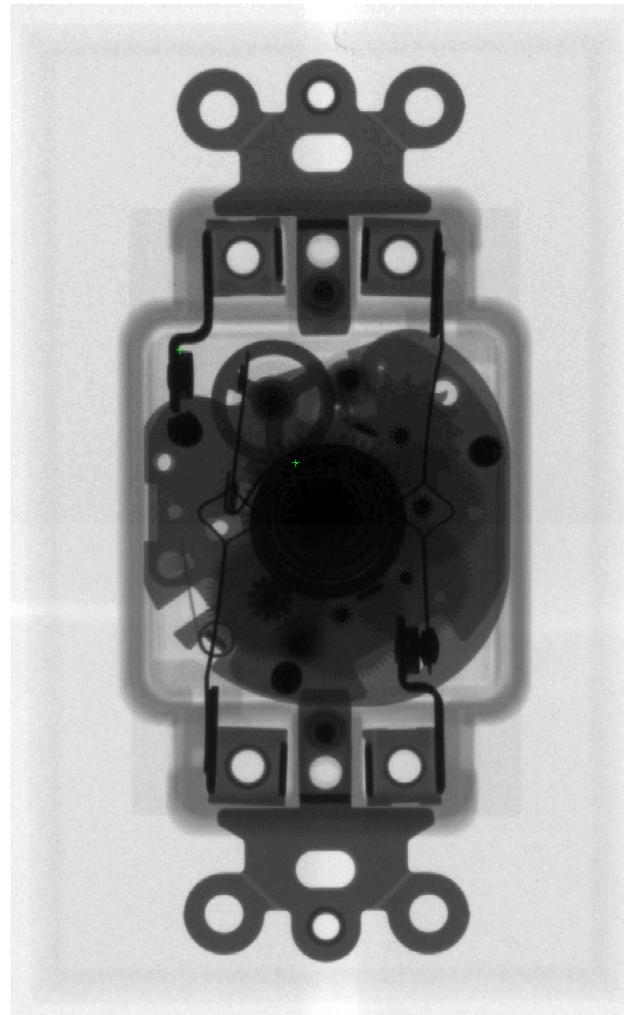
Results



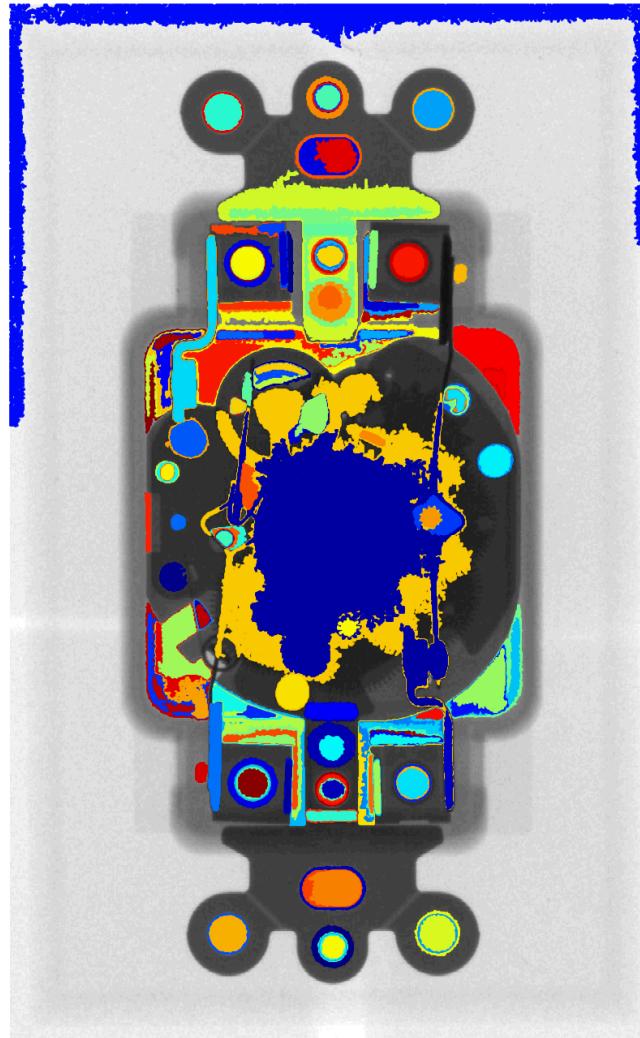
Results



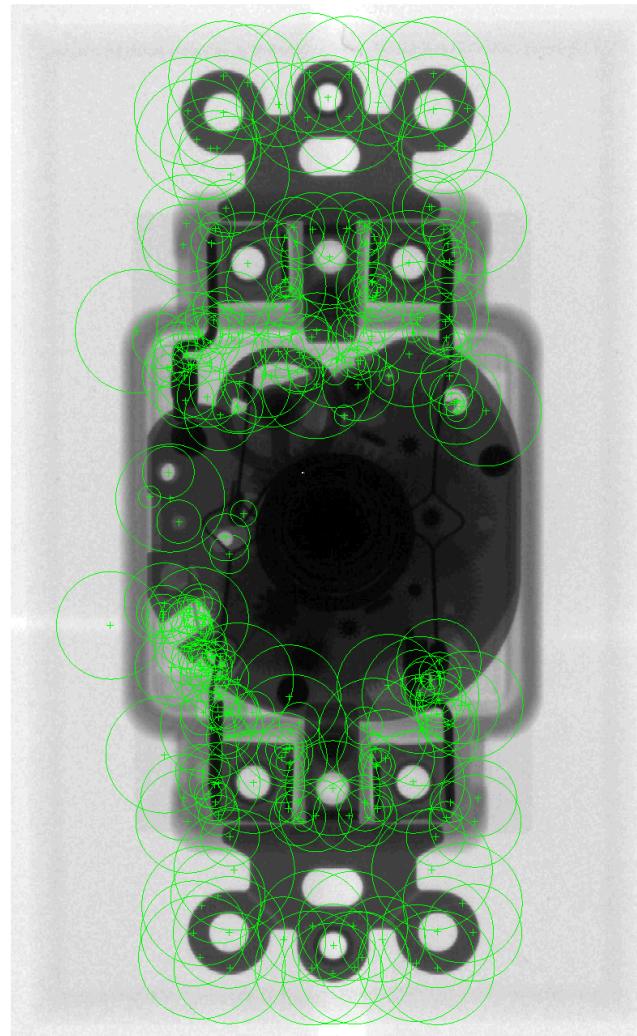
Results - Harris



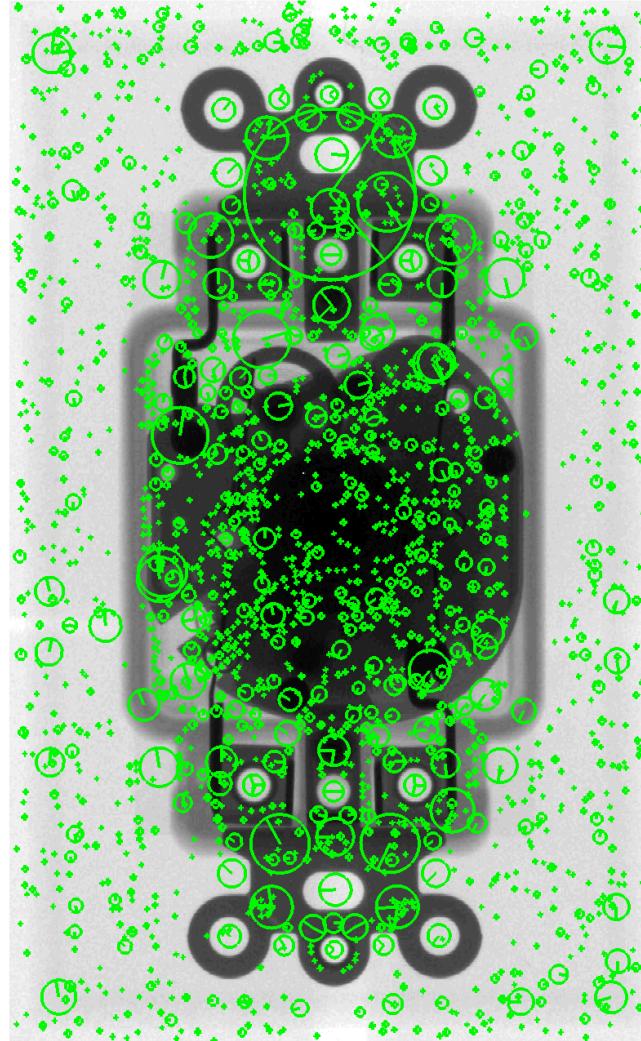
Results - MSER



Results - SURF



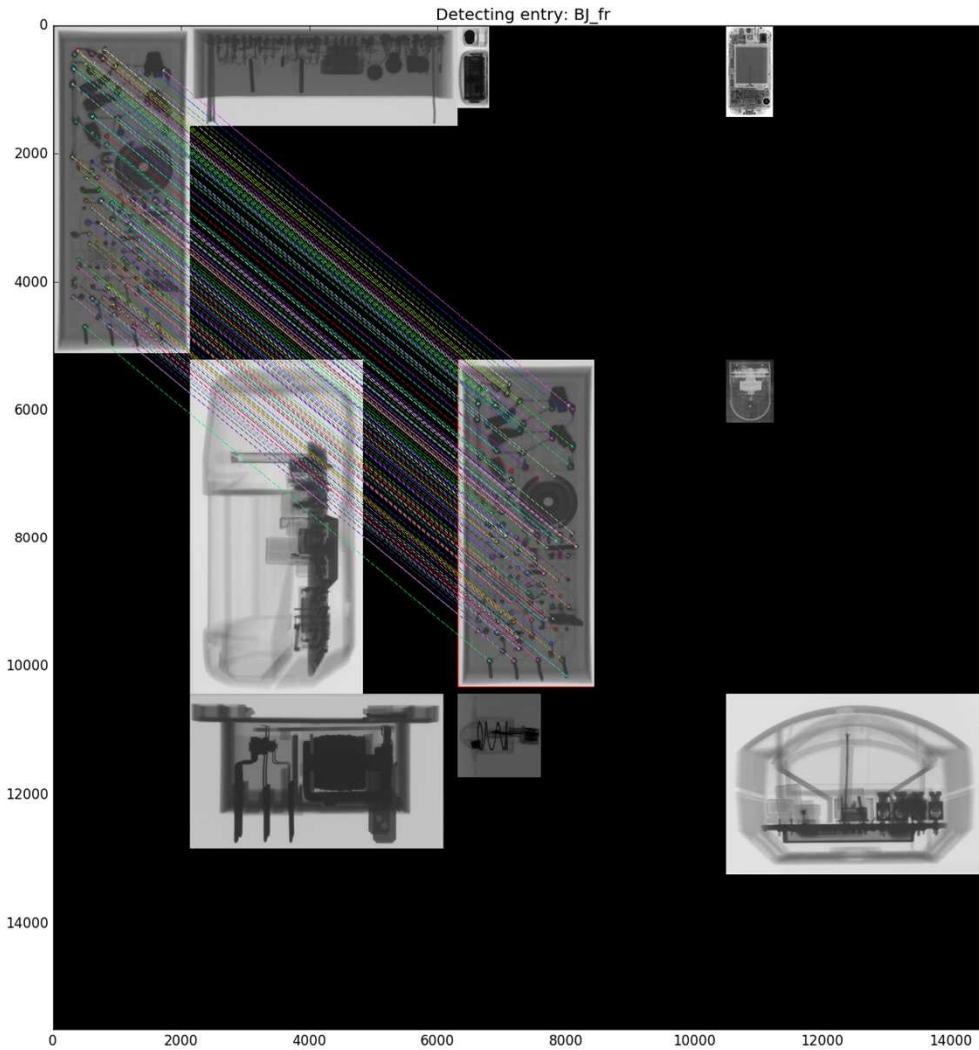
Results - SIFT



Results

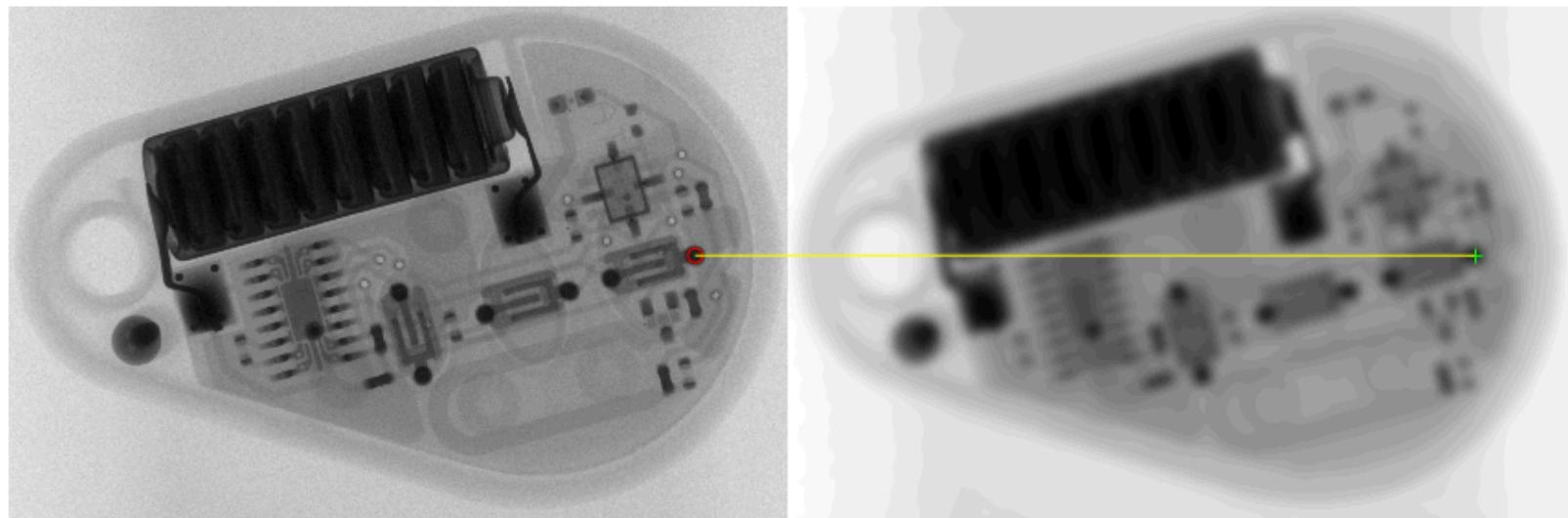
| Algorithm | Average # of features extracted |
|-----------|---------------------------------|
| Harris | 4 |
| MSER | 10 |
| SURF | 120 |
| SIFT | 255 |

Results

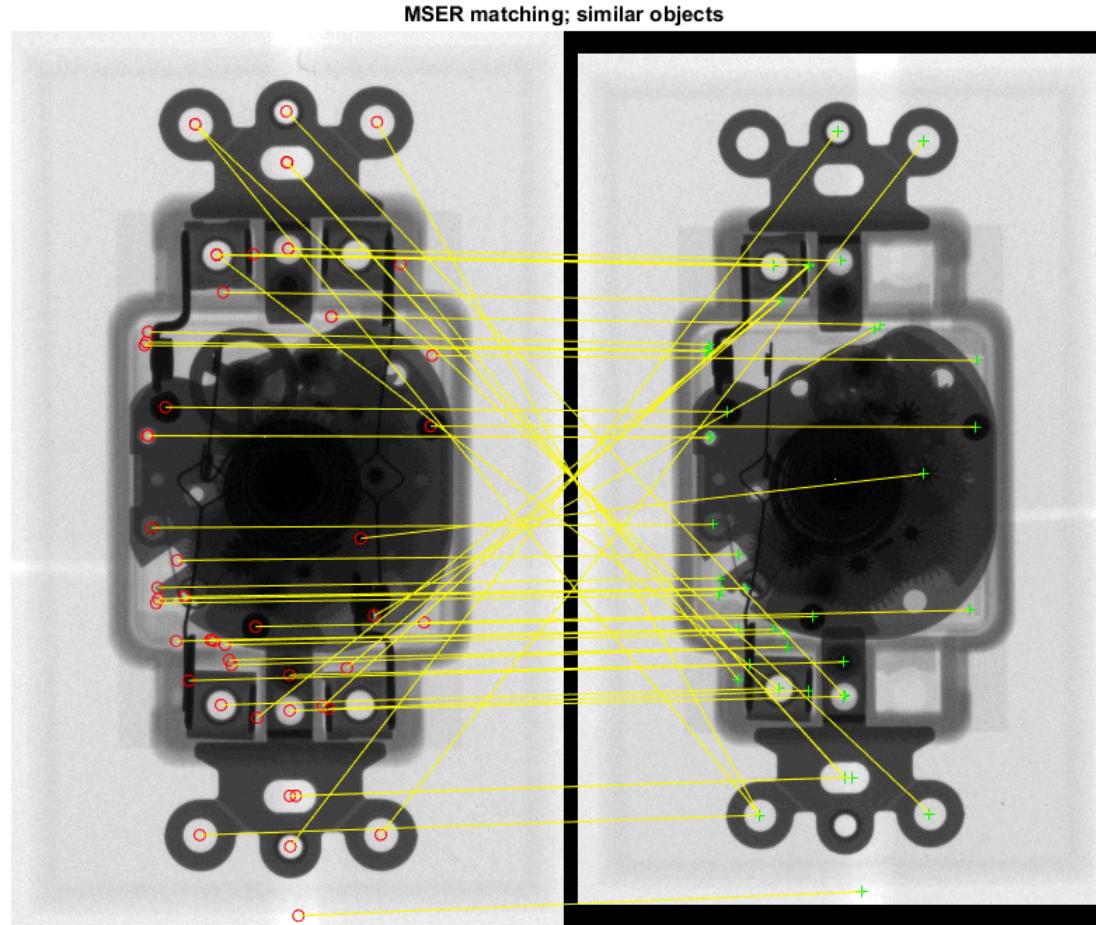


Results

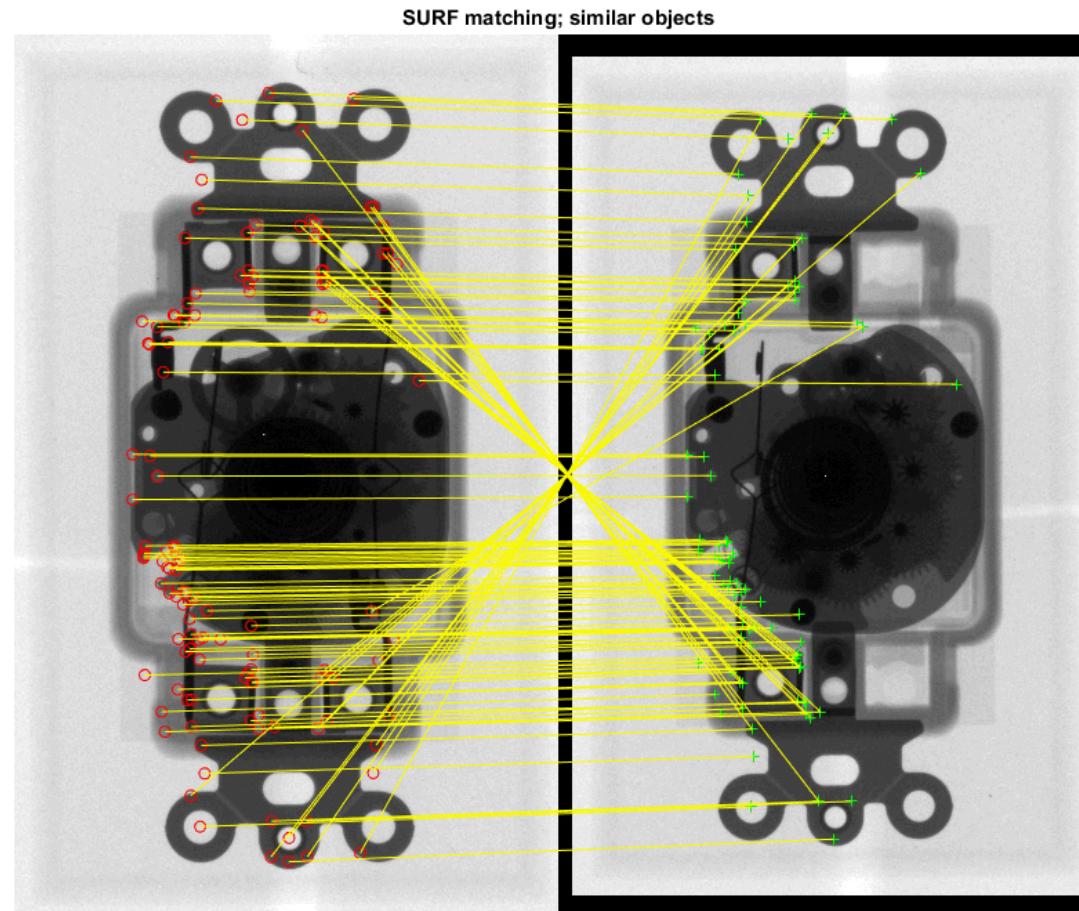
- Blurring



Results



Results



Results

| Algorithm | Our Results | Original Results |
|-----------|-------------|------------------|
| SIFT | 25% | 55% |
| SURF | 23% | 50% |
| MSER | 20% | 40% |
| Harris | 15% | 25% |

Conclusion

- Accuracy **Sift > SURF > MSER > Harris**
- Performance is a big challenge
 - All perform slowly
 - $||\text{Sift}|| > ||\text{SURF}|| > ||\text{MSER}|| > ||\text{Harris}||$
- There are shortcomings and there is room for improvement
- Our particular database makes computer vision challenging
 - Better quality data may yield better results
- Many of the shortcomings of the various algorithms are valid in the x-ray imaging modality as well as many of the advantages
- These algorithms do hold a lot of promise in the x-ray modality for the same reasons they are valid for photography

Future Work

- Narrow down which algorithms work best for radiographs
- Investigate more algorithms based on SIFT
- Create hybrid algorithm which will work for real-world images