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Object Recognition in X-ray Radiographs

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Introduction

- Objective
 - Develop the ability to perform image-based queries on a database of a collection of component radiographs
 - Matching under various conditions (noise, occlusion, out-of-plane rotations, etc.)
- Obstacles
 - Most computer vision work done with visible-light photographs in mind
 - Difficulties posed by X-ray penetration
 - Occlusions
 - Lack of color information
 - Working with our database
 - JPG compression
 - Extraneous images
 - No labelling

Background

- Database preparation (past work)
 - Removing irrelevant images
 - Cleaning radiographs
- Feature detectors/descriptors
 - SIFT
 - SURF
 - MSER
- Image segmentation
 - Global Otsu Method

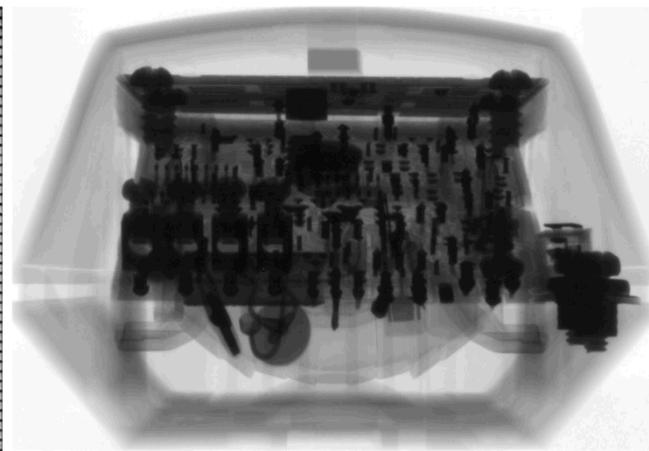
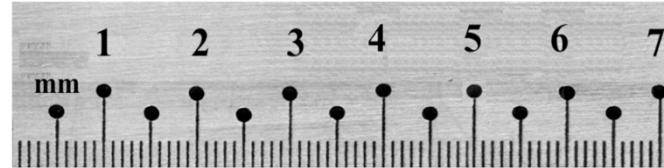
Background – Database preparation

- Not originally intended for object recognition
- Example images from the database

Visonic Duet



Bosch DS840T RZ



Background – Database preparation

- Bray et al. (2015)
 - Removed duplicate data using commonalities in the naming scheme.
 - Separated photographs from radiographs by checking color/greyscale information.
 - Segmented out extraneous objects from the radiographs (rulers, text, etc.).
 - Database was manually checked to verify that it was successfully cleaned.

Background – SIFT/SURF

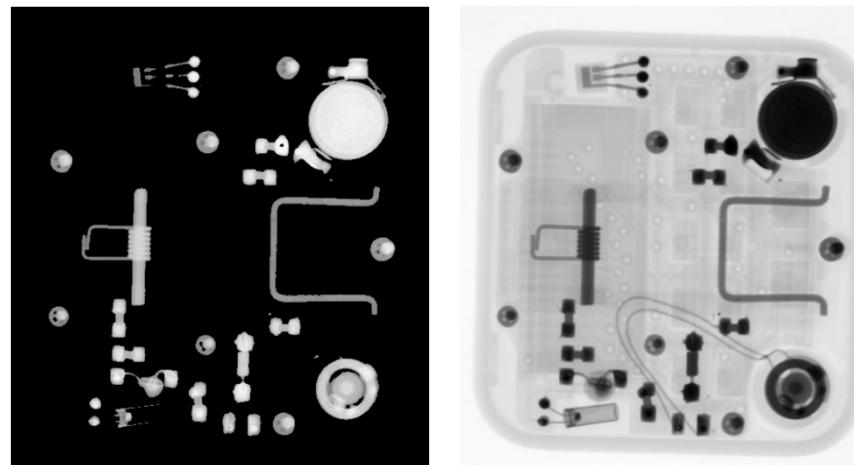
- SIFT – Scale Invariant Feature Transform
 - David G. Lowe (2004)
 - One of the most widely used feature detectors/descriptors
 - Invariant to:
 - Affine transformations (scale, translation, etc.)
 - Noise
 - Rotation (slightly)
- SURF – Speeded Up Robust Features
 - Inspired by SIFT
 - Specifically designed to be significantly faster than SIFT
- SIFT: accurate
- SURF: fast

Background – MSER

- MSER – Maximally Stable Extremal Regions
 - Used for “blob” detection
 - Detects regions of uniform intensity values
 - Described using enclosing ellipses
- Can be particularly in radiographs since intensity values are essentially the ONLY information provided.
- Possible implementations combine an MSER detector with a SIFT descriptor

Background – Segmentation

- So far, segmentation only intended to remove background
- Future work will look into separating various layers
- Global Otsu Method
 - Nobuyuki Otsu
 - Histogram cluster-based image thresholding
 - Attempts to minimize the in-class variance when segmenting the histogram



Background

- SIFT



- SURF



- MSER



Approach

- Experiment
 - Test available software
 - MATLAB: Image Processing Toolbox, Computer Vision Toolbox
 - OpenCV (Python implementation)
 - VLFeat
- Apply
 - Determine ideal processing and recognition techniques for our specific database
- Expand
 - Experiment with advanced methods specific to X-rays
 - Research and develop alternative methods

Approach - Experiment

- MATLAB
 - Image Processing Toolbox
 - MATLAB's system for reading/writing/editing images.
 - Contains many standard image processing and image analysis tools.
 - In our case, used mainly for filtering, segmentation, and morphological operations.
 - Computer Vision Toolbox
 - Contains implementations of the most common computer vision techniques.
 - Includes functions for object recognition, tracking, facial recognition, image registration, and various other tools.
 - We used this primarily for feature extraction and matching.

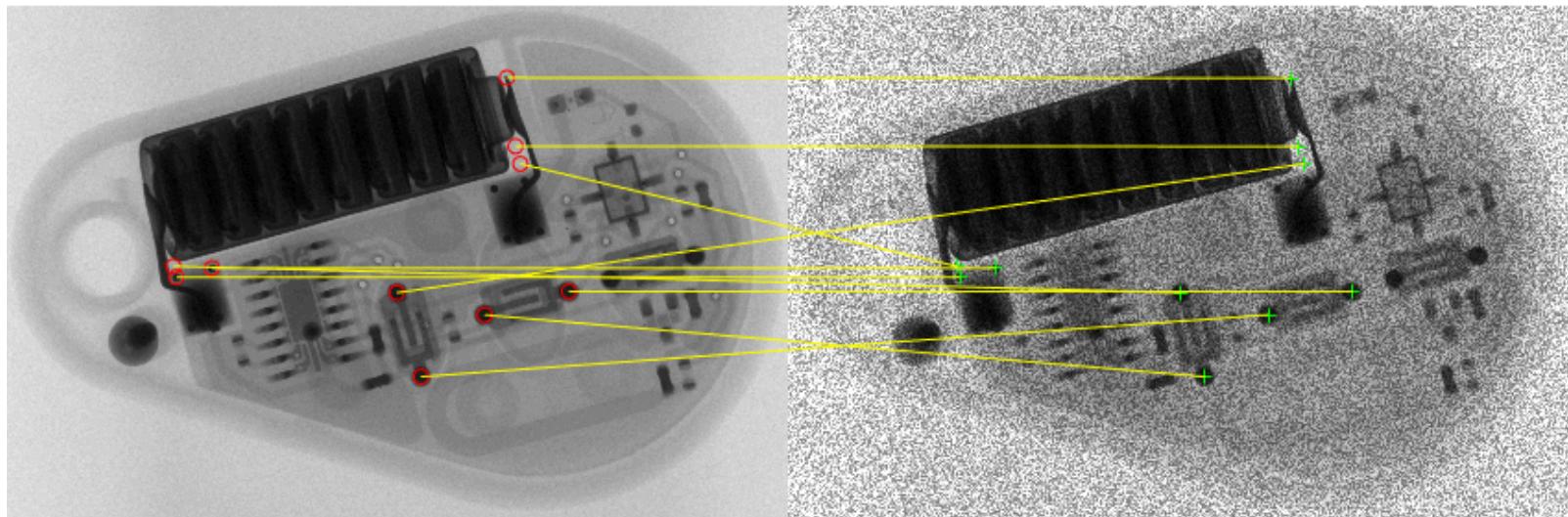
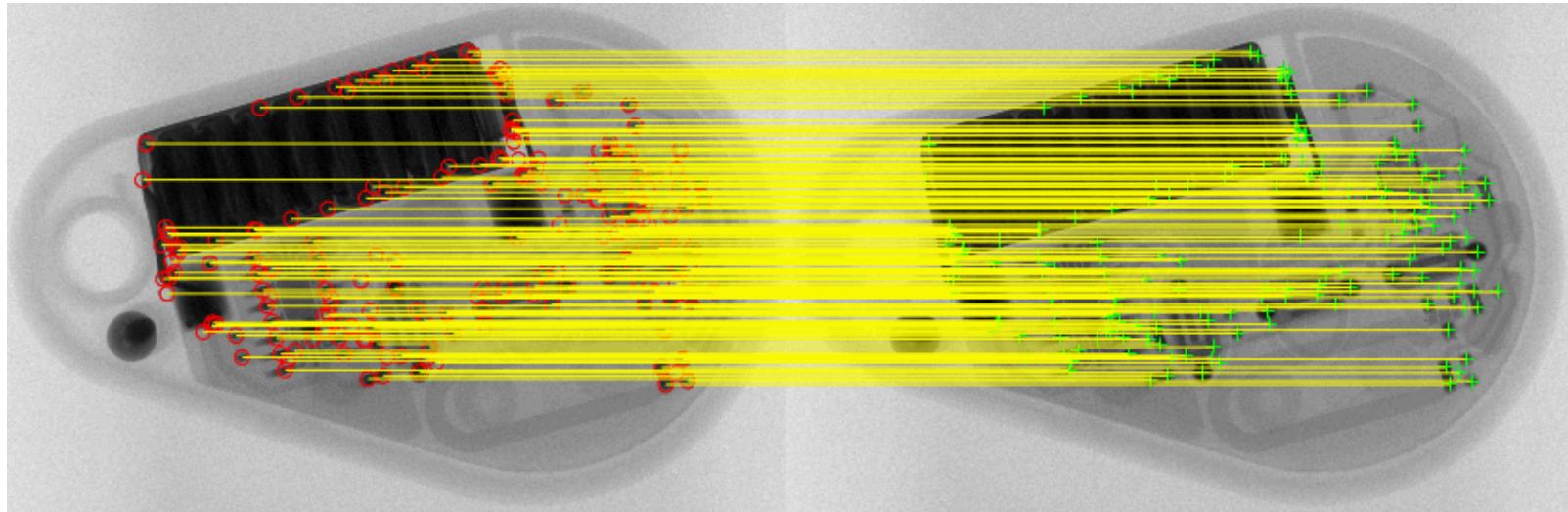
Approach - Experiment

- OpenCV
 - Open source
 - Written in optimized C/C++
 - Has C++/C/Java/Python implementations
 - Extremely large range of tools
 - Image processing
 - Computer vision
 - Machine learning
 - Many more

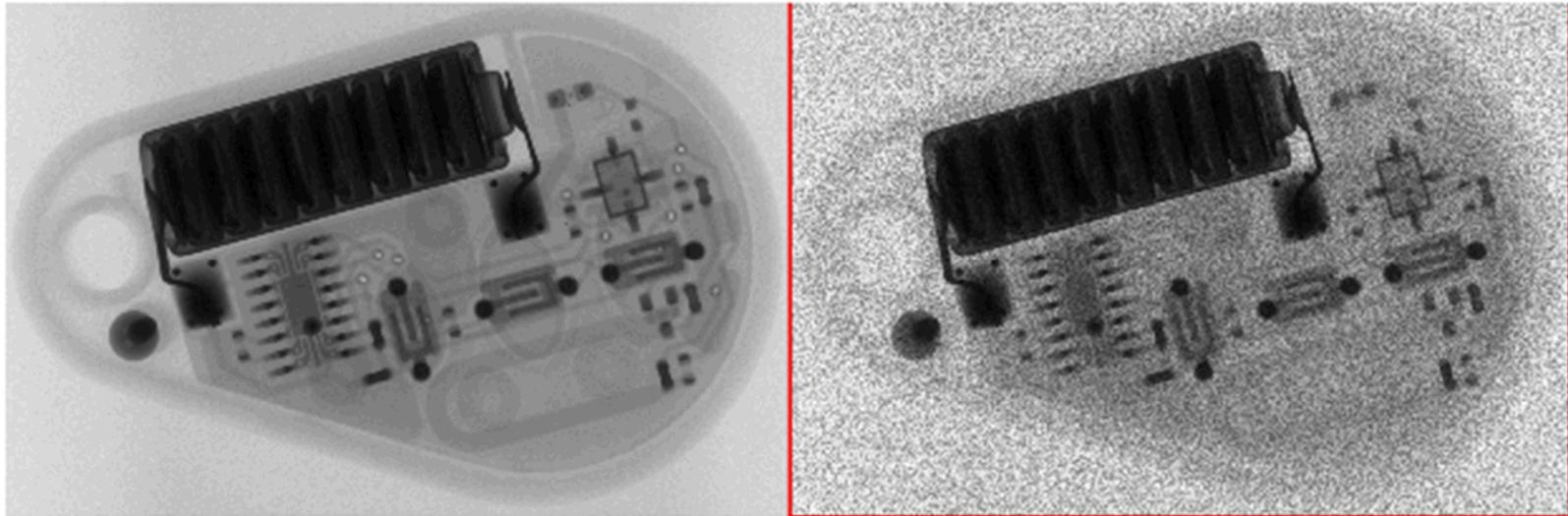
Approach - Experiment

- VLFeat
 - Open source
 - Written in C; interfaces in MATLAB/Python
 - Toolset less exhaustive than MATLAB/OpenCV
 - Supports some feature detectors/descriptors
 - Small number of image processing functions
 - Does have some statistics and data analysis tools
 - **Implements SIFT features**
 - By default, MATLAB and OpenCV (OpenCV > 2.4) don't have SIFT features
 - We specifically used this for the MATLAB implementation of SIFT features

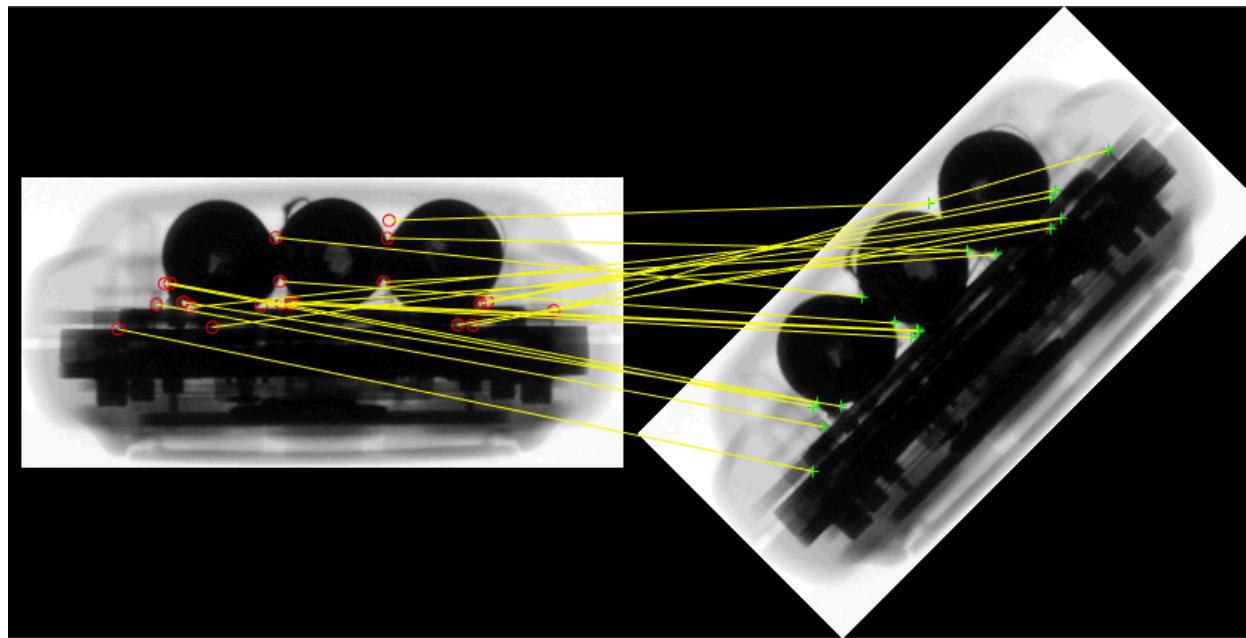
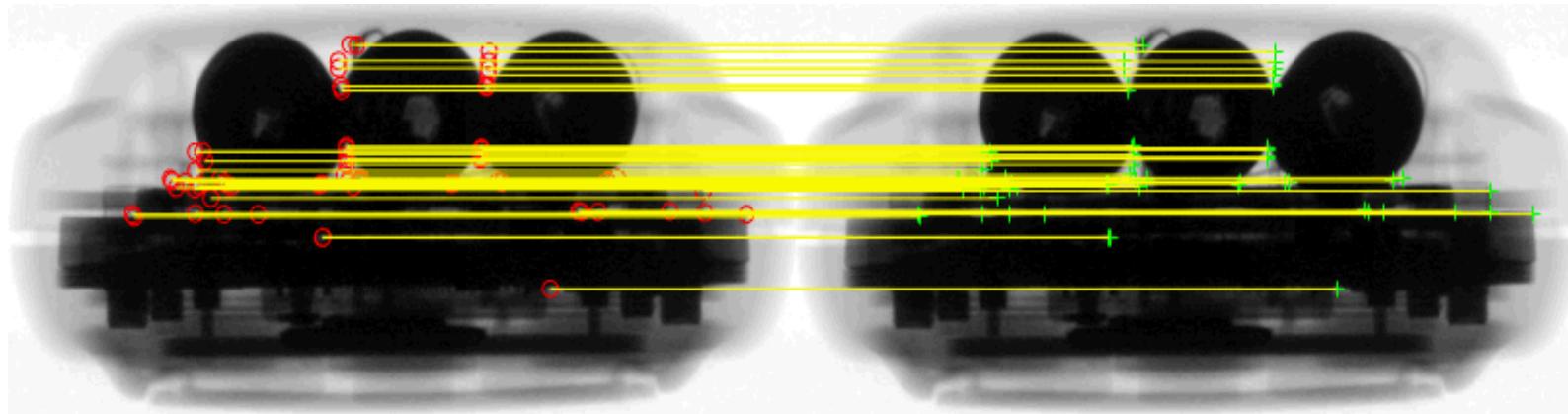
Results – Noise (SIFT features)



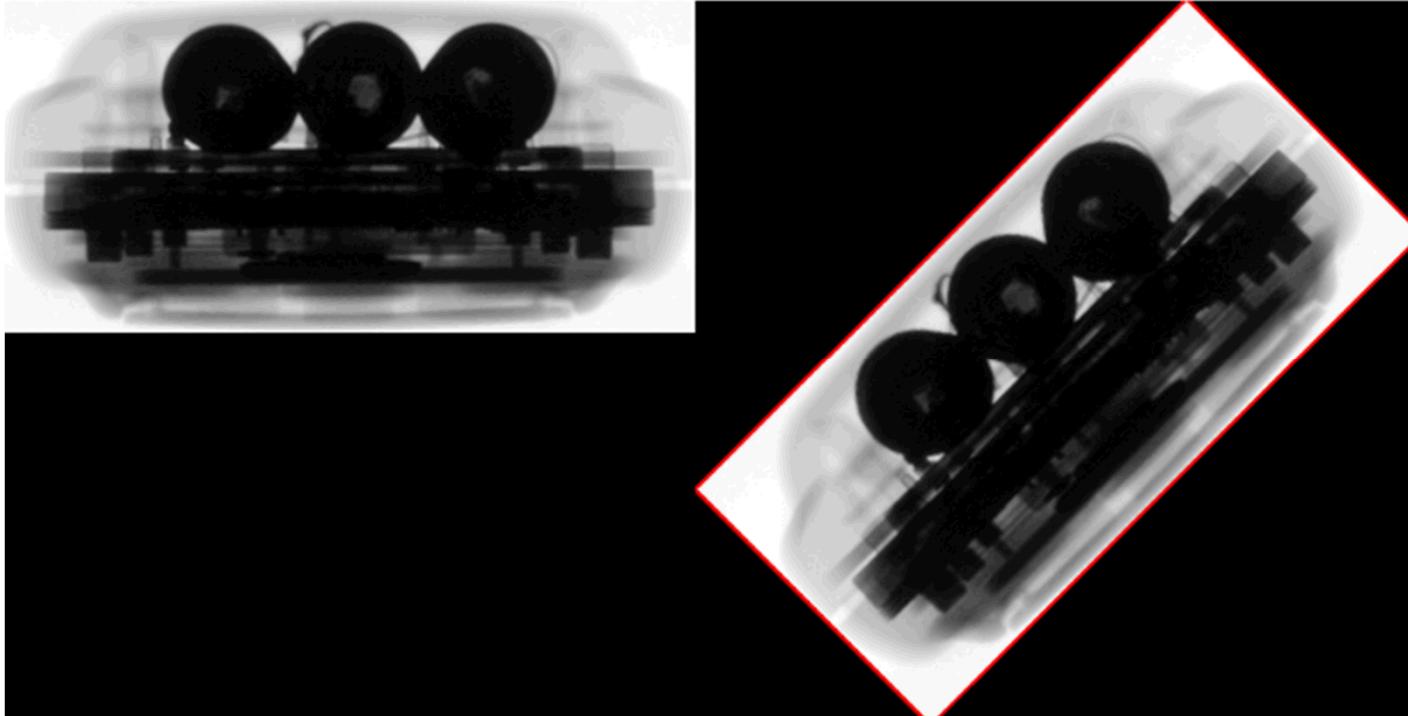
Results – Noise (SIFT features)



Results - Rotation (SIFT features)

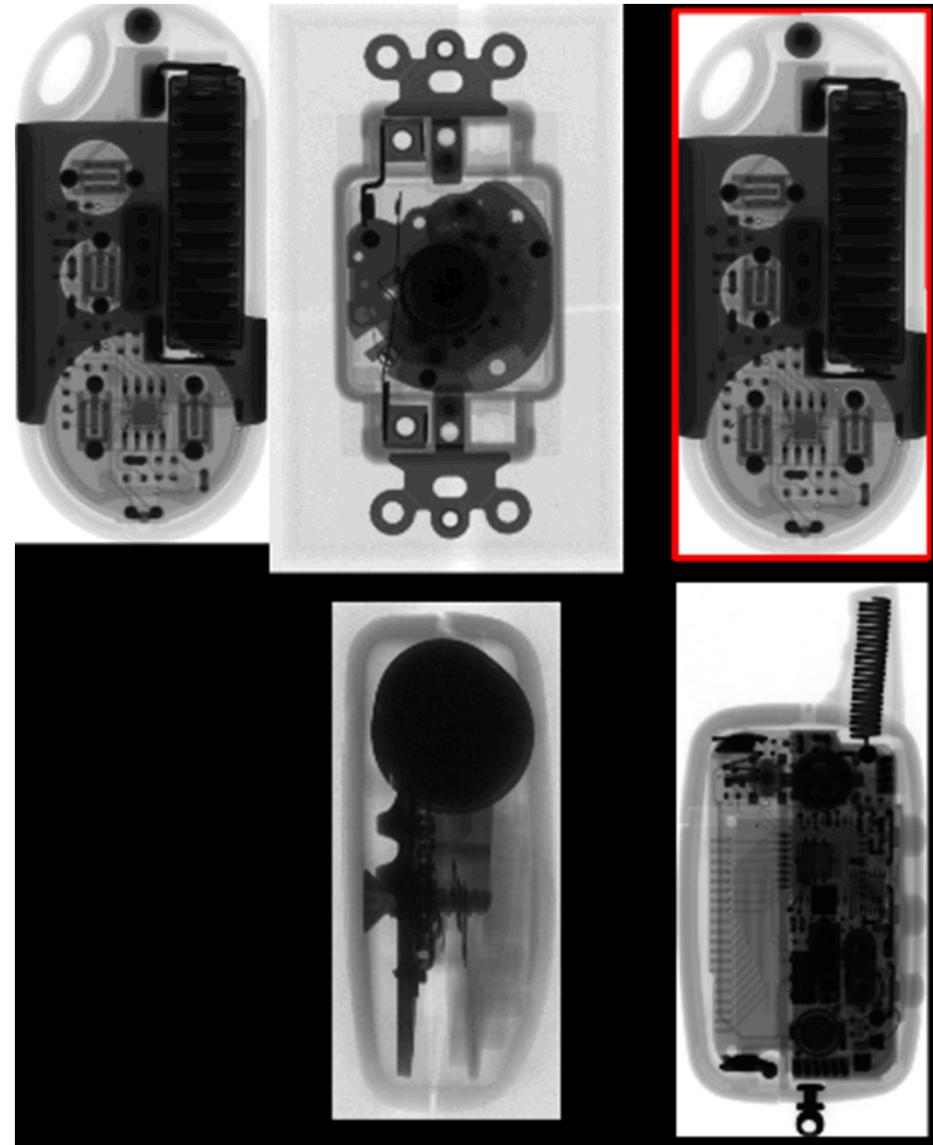


Results - Rotation (SIFT features)



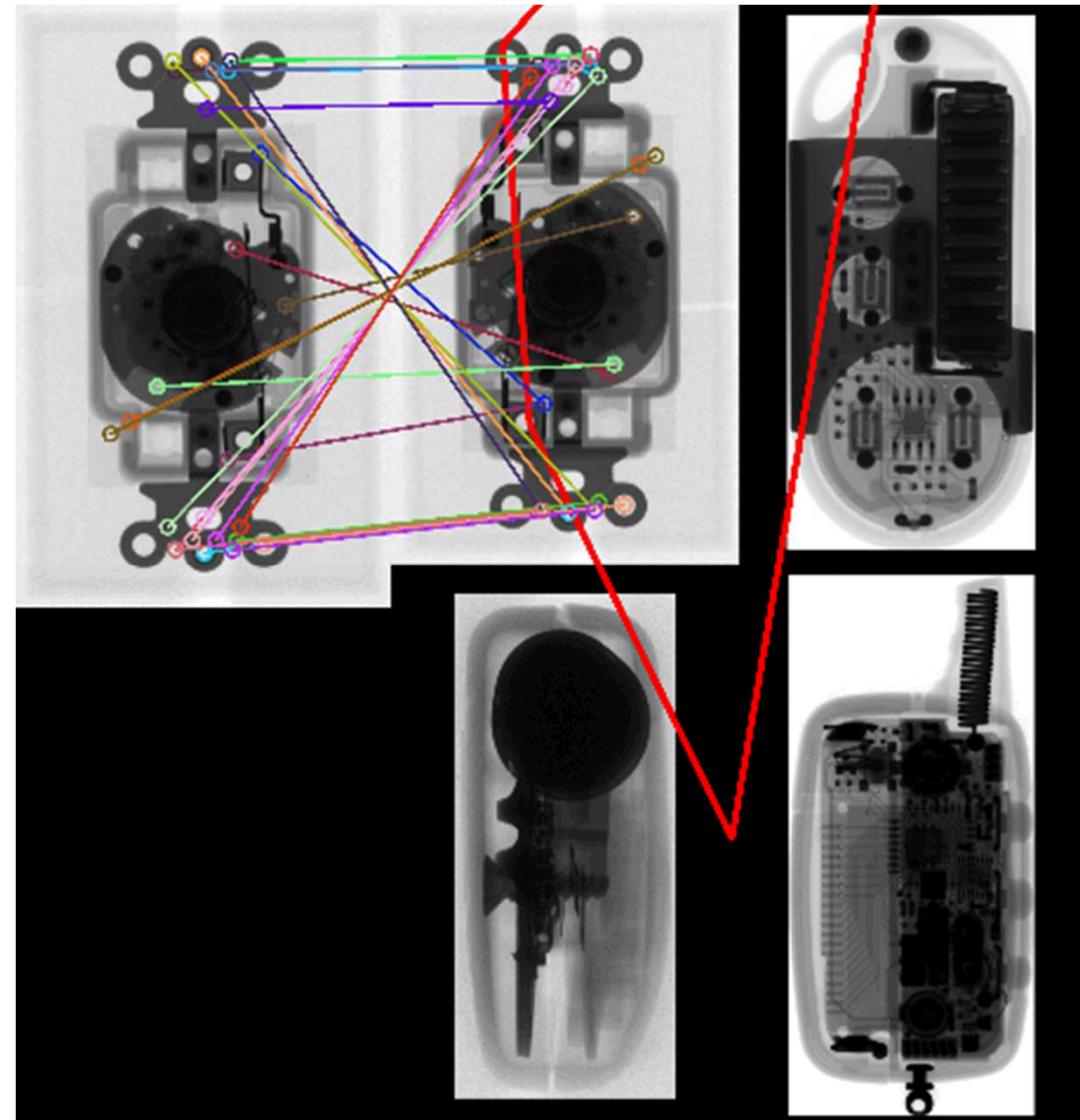
Results – Groups (SIFT features)

- Good match



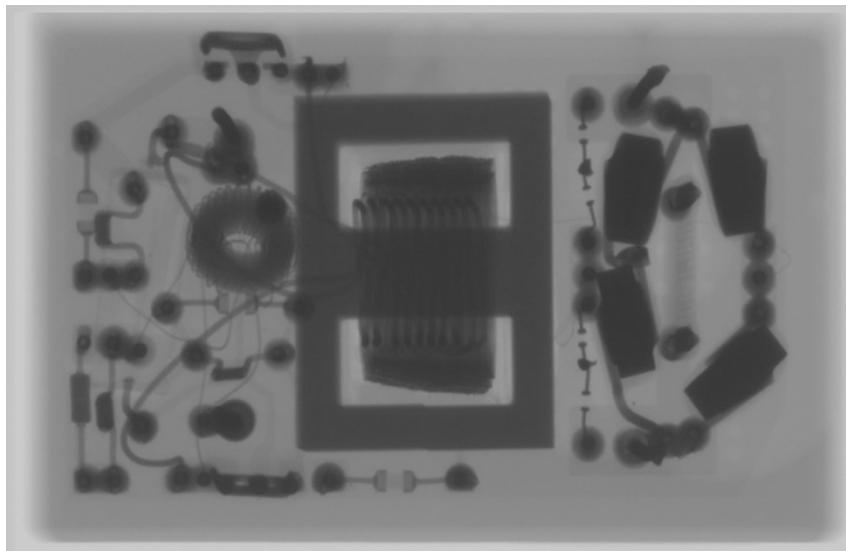
Results – Groups (SIFT features)

- Skewed match
- Matched features drawn to emphasize cause of poor bounding box calculation

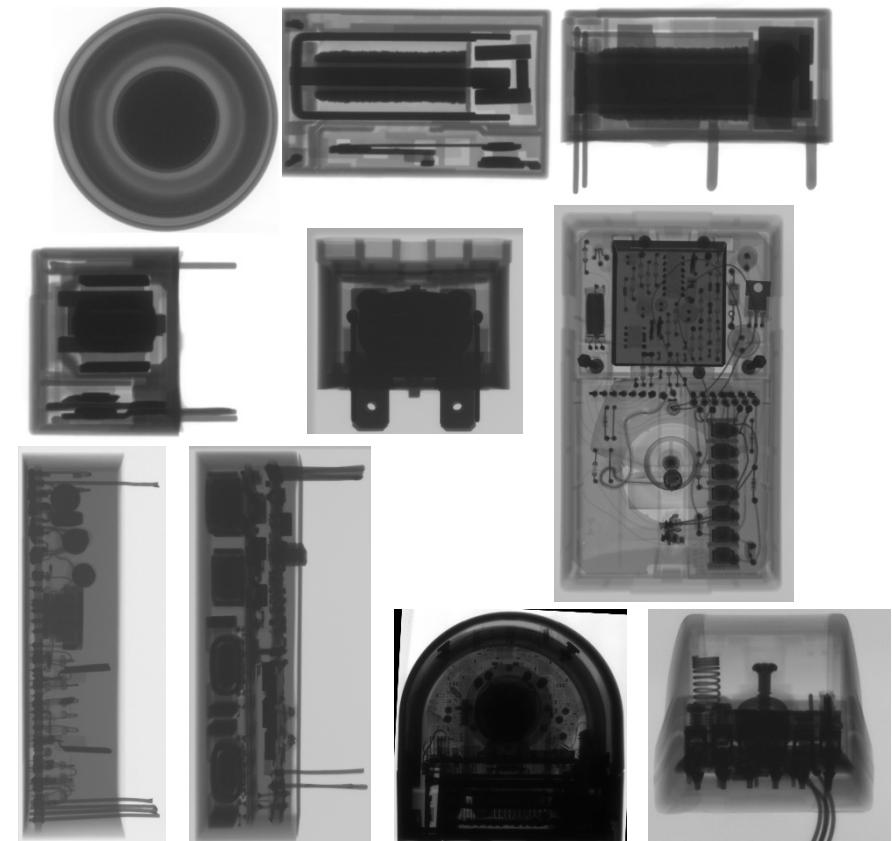


Results

Test image



Top matches (not to scale)



Discussion

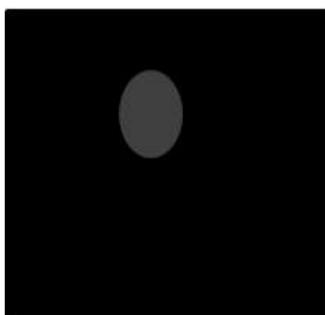
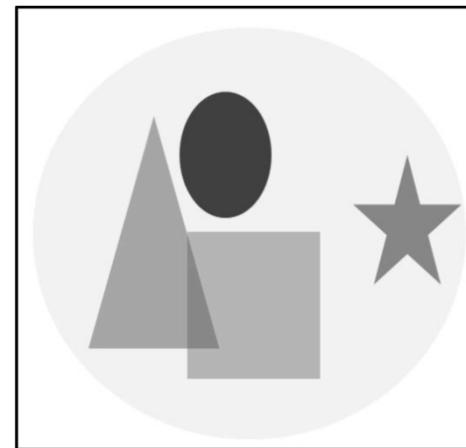
- Strengths of algorithm
 - Handles simpler obstacles easily (noise, rotation, scaling)
 - Can positively detect objects even in groups
 - Segmentation removes background extremely well
- Weaknesses of algorithm
 - Can't match “similar” objects
 - Struggles with more difficult obstacles (occlusion, out-of-plane rotation, etc.)
 - Segmentation unable to handle transmission overlap

Future work – Bag-of-Words

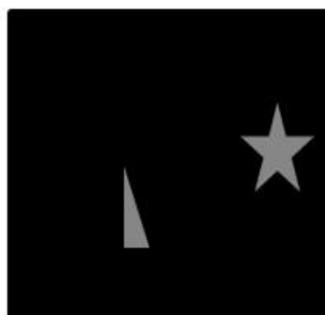
- Bag-of-Words
 - Initially designed for language processing and text analysis
 - Creates a simplified version of an object
 - e.g breaking down a paper about Christmas into key words such as “presents”, “tree”, “ornaments”, “lights”, etc.
 - Extended to computer vision
 - Instead of [document] and [key words], uses [image] and [features]
 - Pros
 - Able to match “similar” objects
 - Can detect multiple objects in a single image
 - Cons
 - Requires labelled images for machine learning

Future work – Segmentation

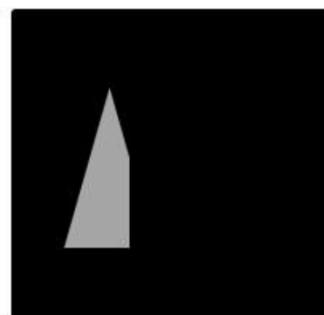
- X-ray penetration introduces segmentation problems that aren't present in photographs
- From Bray et al.:



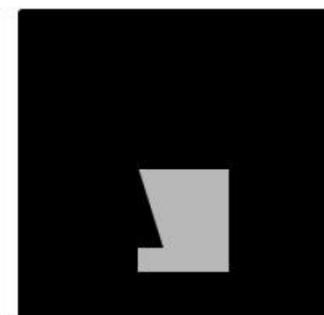
Layer 1
1 Object



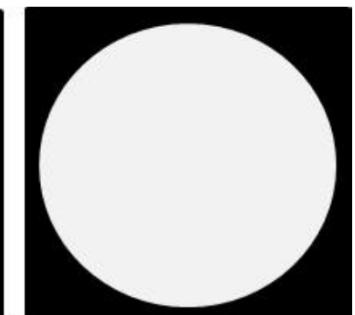
Layer 2
2 Objects



Layer 3
1 Object



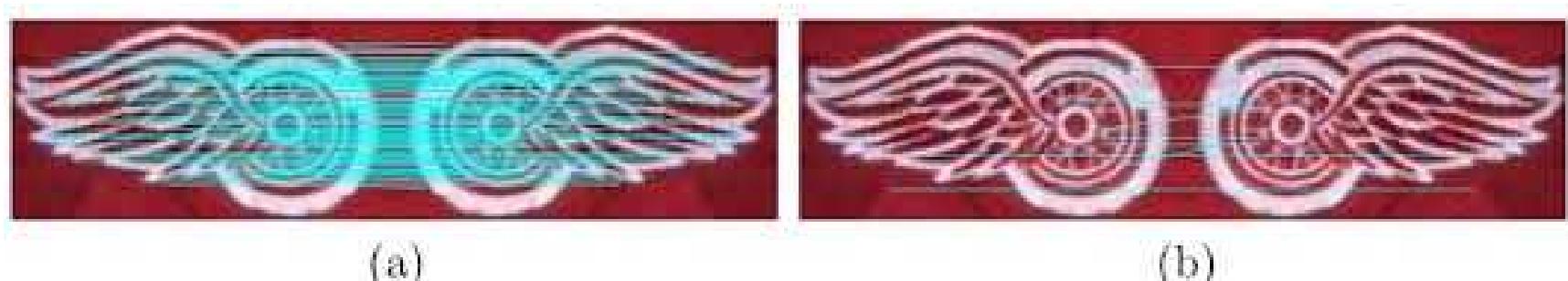
Layer 4
1 Object



Layer 5
1 Object

Future work - MIFT

- MIFT
 - Guo et al. (2010)
 - Variation on SIFT that is also invariant to mirror reflections
 - A comparison of MIFT (a) vs. SIFT (b)



- Just one example of how improvements to feature descriptors can greatly increase performance