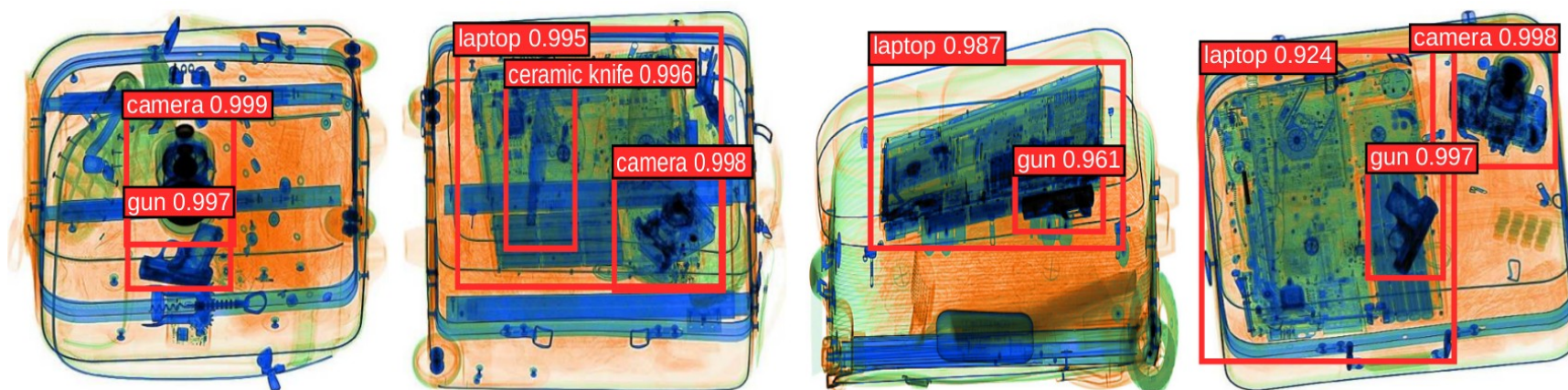


Automatic Prohibited and Illicit Item Detection in X-ray and Computed Tomography Security Screening

– a research snapshot



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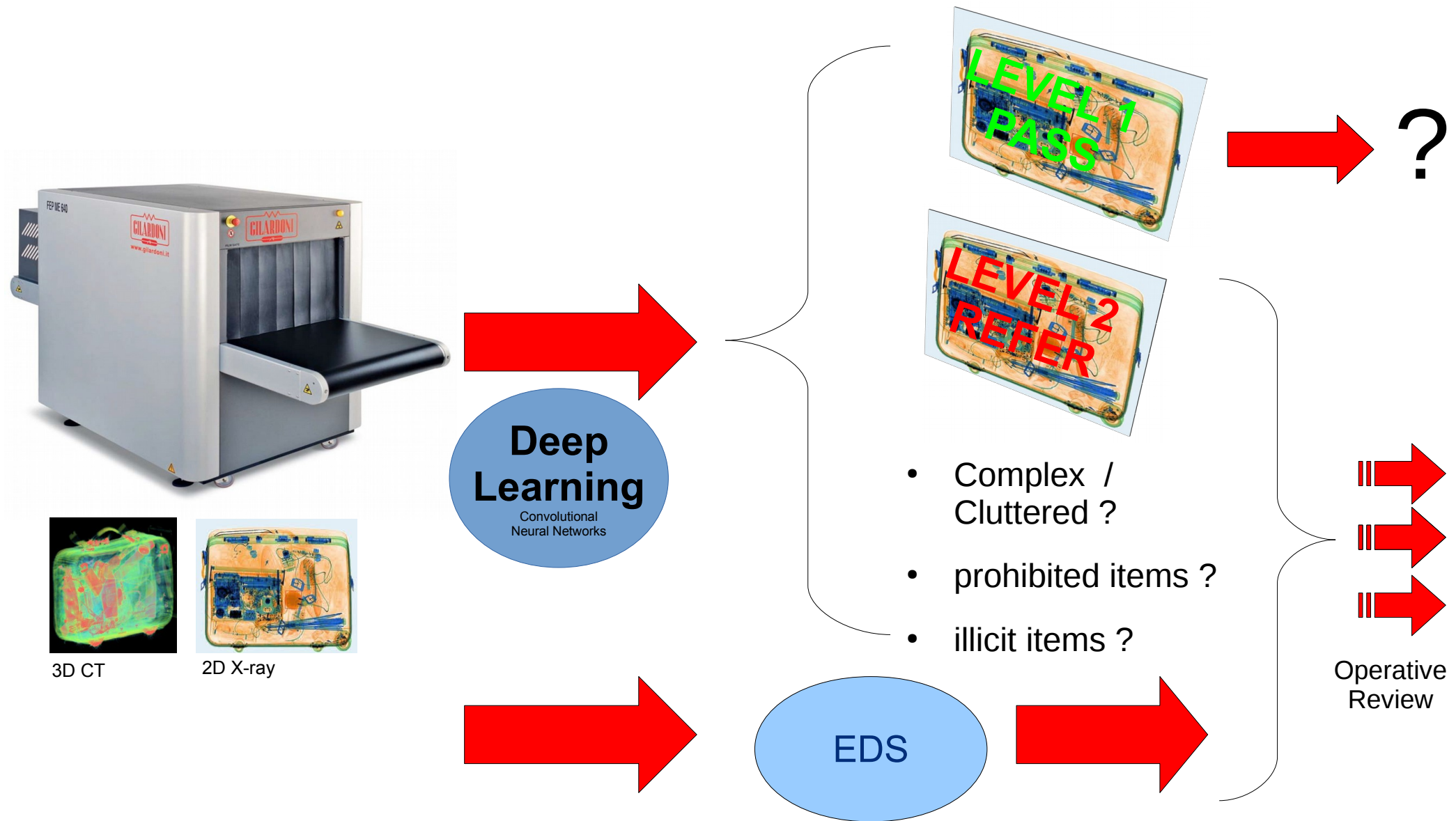
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So What ? / Who Cares ?

- *Space: Baggage & Parcel Inspection*
(carry on and hold, extensible to freight)
- *Problem:*
 - **Prohibited Item Detection** (by shape/material ... guns / knives / other - ?)
 - **Anomalous Item Detection** (by knowing what is abnormal - ?)
- *Solution:*
 - **3rd party, world-leading automatic object detection & classification algorithms**
 - using 2nd / 3rd generation deep learning techniques
- Results: **~98%+, < 1 sec., FP <1%, invariant** (on firearms detection, > 95% for other)
- TRL: 6
- Contact me: toby.breckon@durham.ac.uk

Concept of Operation





- **3rd oldest** university in England (1832)

- **World leading** university (top 100)

- **UK ranking: top 5**
Engineering & Computer Science

- **Engineering & Computer Science**

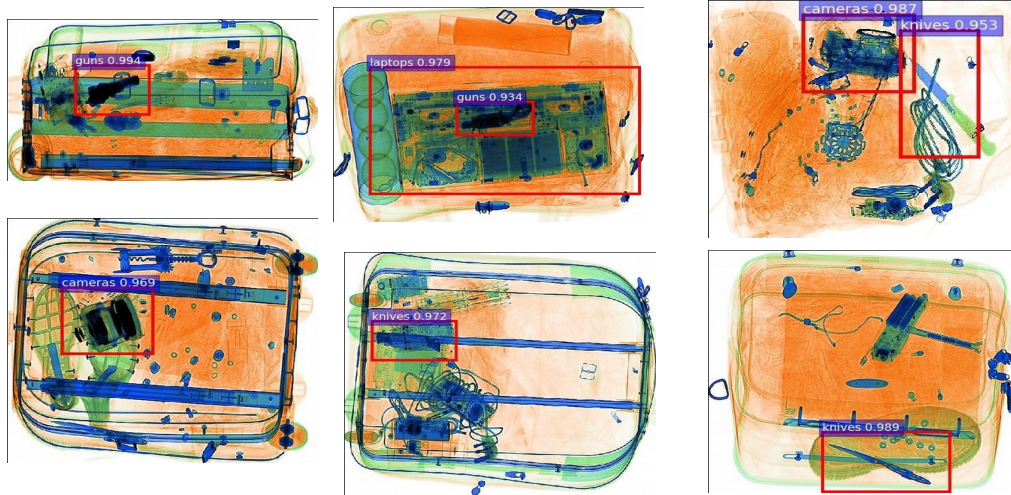
- Nvidia Research Centre
- Intel Parallel Computing Centre



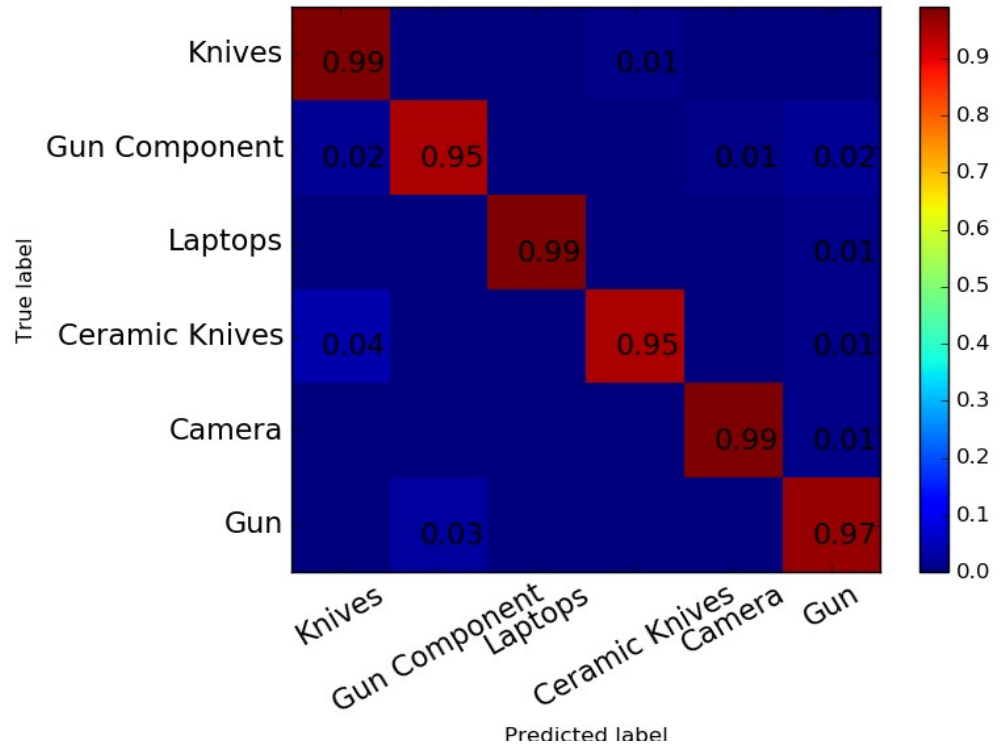
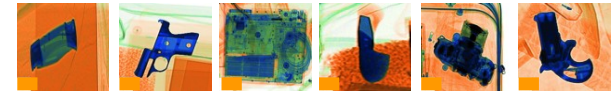
- **Within X-ray Security:**
 - 12 years experience
 - threat detection, threat image projection, anomaly detection



Deep Learning for Object Detection in 2D X-ray



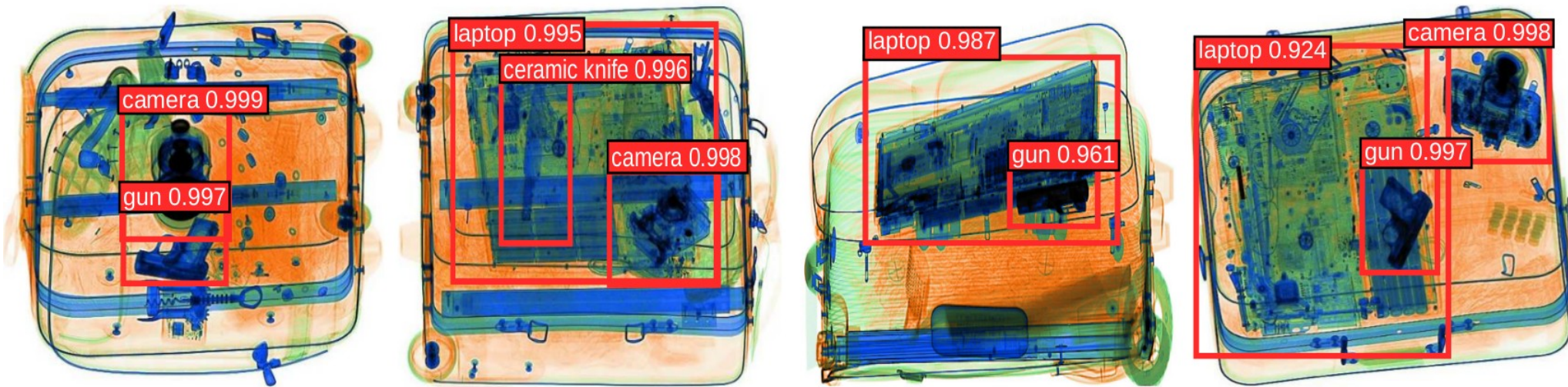
Method	Class	True +	False +
[Akçay et al. 2016]	Firearm	98.62	0.21



- 1st generation deep net approaches

- 95% (True+) over 6 object categories
- established X-ray training via transfer learning (which everyone uses now) [Akçay et al. 2016]

Deep Learning for Object Detection in 2D X-ray

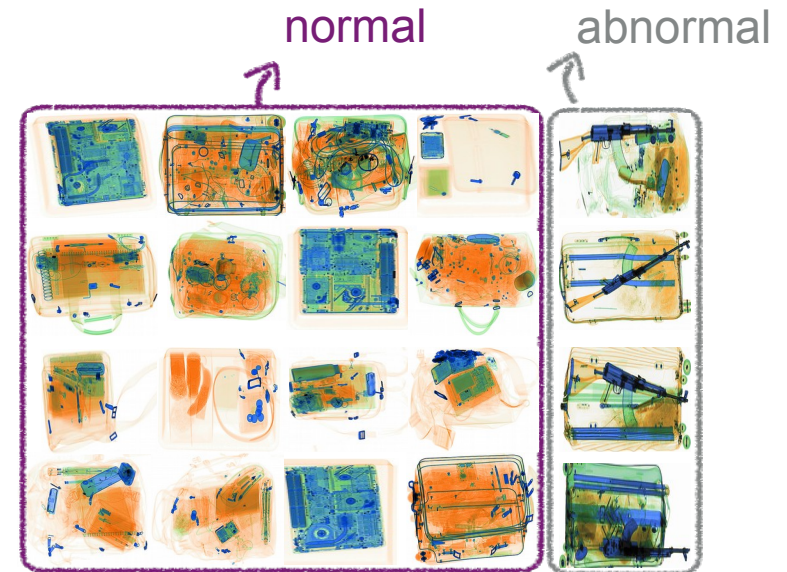
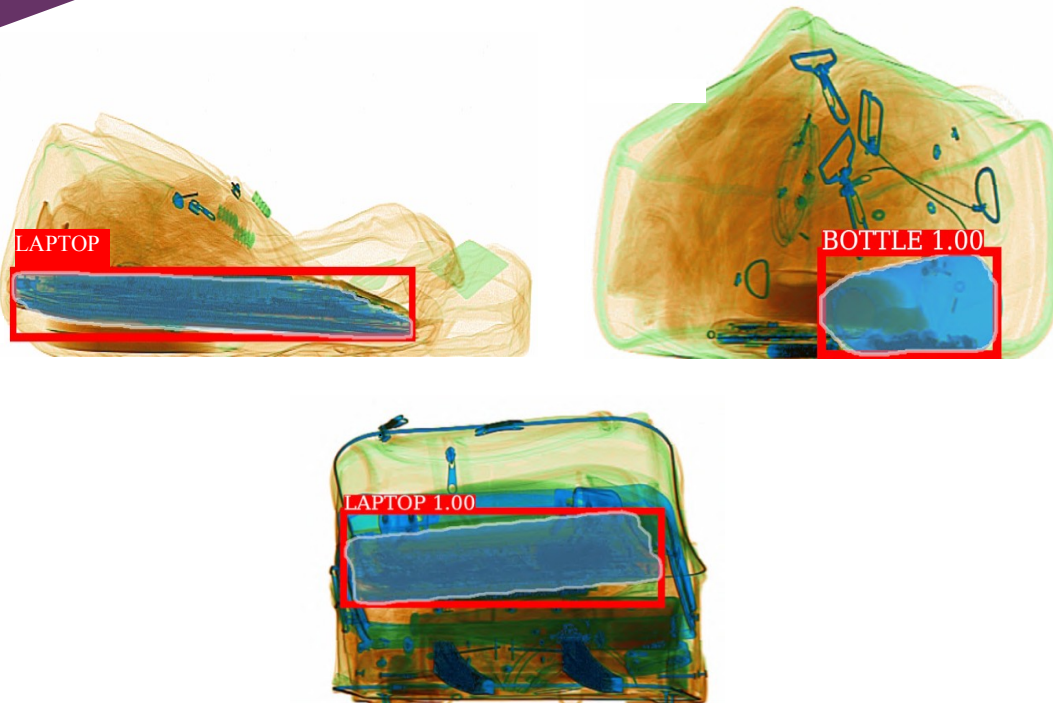


Method	Class	True +	False +
[Akçay et al. 2017 / 2018]	Firearm	99.5+	< 0.5

- **2nd generation deep net approaches**

- high PD (true+), low PFA (false+) [Akçay et al. 2017]
- leading global results; UK government test dataset [Akçay et al. 2018]

Deep Learning for Anomaly Detection in 2D X-ray



Method	Durham X-ray Dataset (Dbf3)				HMG (FFOP)
	gun	gun-parts	knife	overall	full-weapon
AnoGAN	0.598	0.511	0.599	0.569	0.703
Eff GAN	0.614	0.591	0.587	0.597	0.712
GANomaly	0.747	0.662	0.520	0.643	0.882

Statistic: AUC

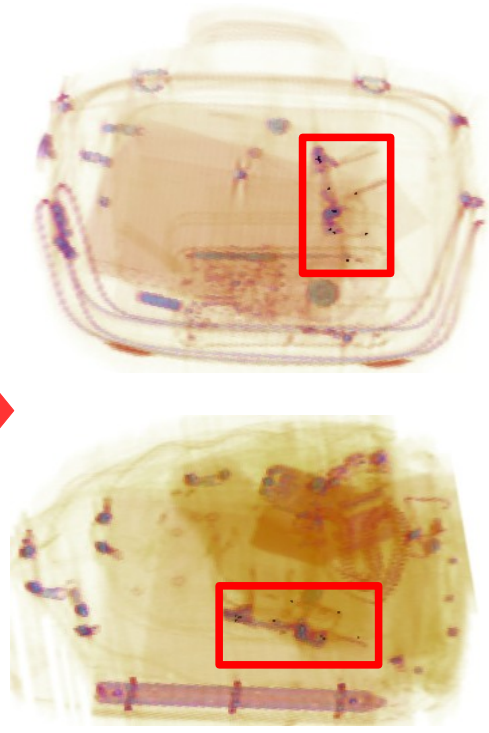


- 3rd generation deep net approaches

- need normal-only training data - GANomaly [Akçay et al. 2018]
- use of object-wise and component-wise anomalies [Gaus et al. 2019 + in press]

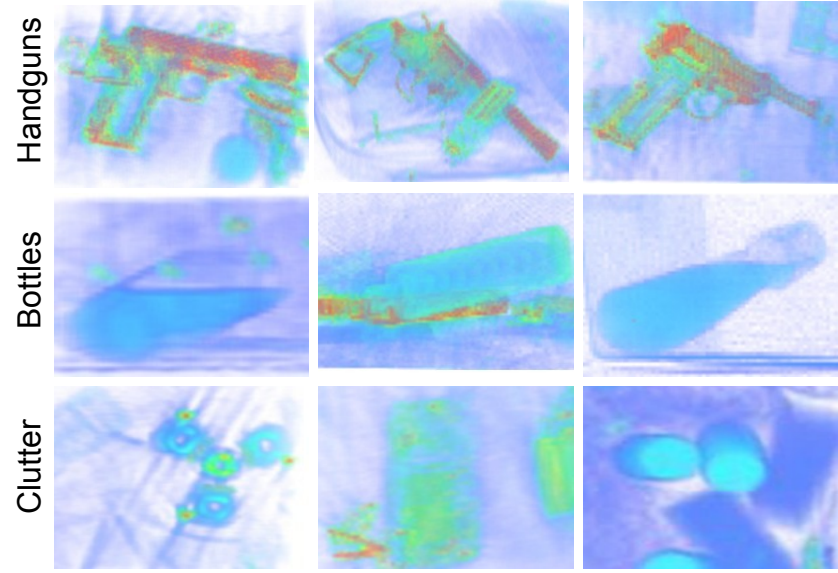
Also available in 3D CT ...

[prior work]



Single signature feature-point based **detection**: ~90% detection

Images [Flitton, Breckon, Megherbi - 2010]

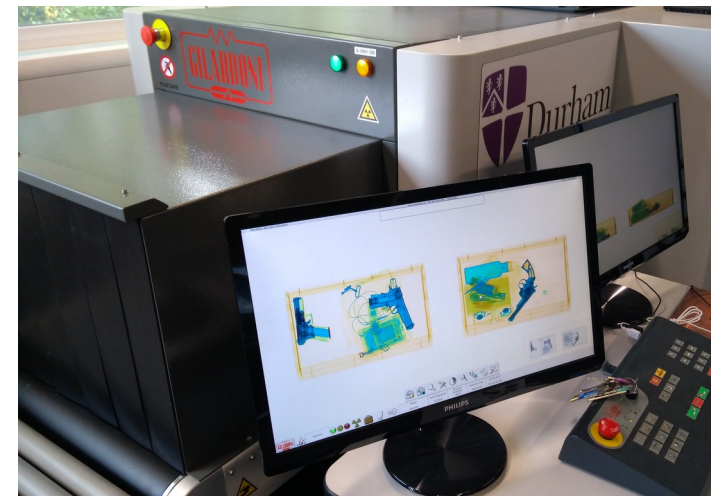


“bag of visual words” generalized signature **classification** : ~98+% detection, low FP (<1%)

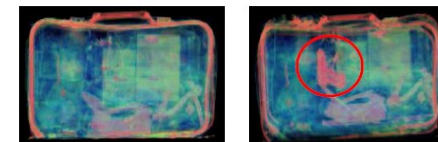
Images - [Mouton, Breckon, 2014] [Mouton, Breckon 2015] [Flitton, Breckon 2015] [Flitton, Breckon 2012]

Working with:

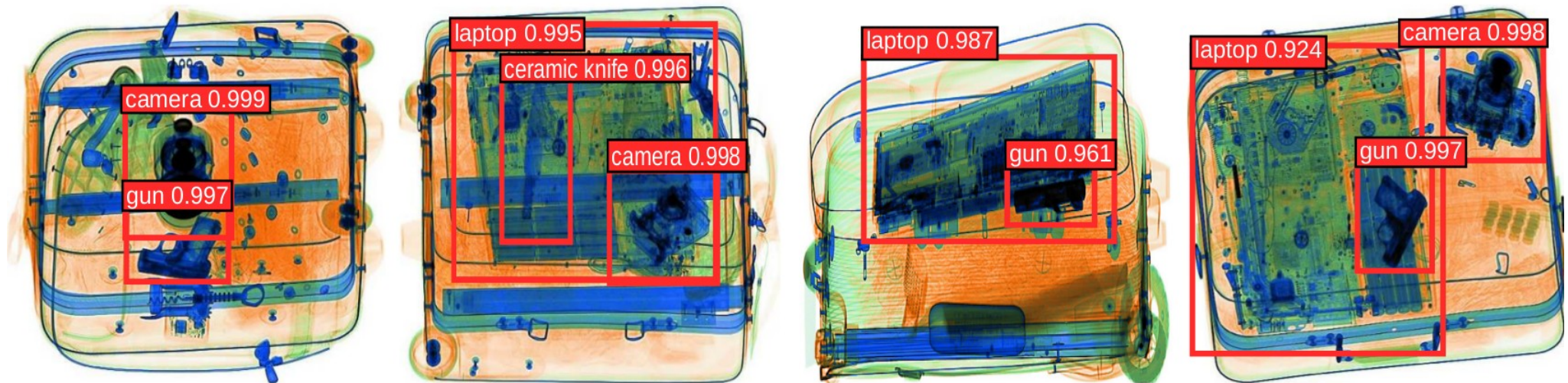
Experience in the Field ...



- **Training Data:**
 - CT : ~800-1,000+ bags
 - 2D X-ray : UK gov. + our own on-site X-ray scanner (~100,000+ images)
- **Funding:** 2007 → 2019+
 - *Today:* 10+ years, 10+ projects and 25+ publications later
- **Publications:** *“never unreasonably withheld”*
 - published in leading conference / journal venues
 - wider impact in generalized 3D object recognition + medical CT
- **Algorithm Deployment:** 3D TIP solution



Automatic Prohibited and Contraband Item Detection in X-ray and Computed Tomography Security Screening – *a research snapshot*



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References : Latest Technical Publications

X-ray Detection:

[On Using Deep Convolutional Neural Network Architectures for Automated Object Detection and Classification within X-ray Baggage Security Imagery](#) (S. Akcay, M.E Kundegorski, C.G. Willcocks, T.P. Breckon), In IEEE Transactions on Information Forensics & Security, IEEE, Volume 13, No. 9, pp. 2203-2215, 2018..

[Transfer Learning Using Convolutional Neural Networks For Object Classification Within X-Ray Baggage Security Imagery](#) (S. Akcay, M.E. Kundegorski, M. Devereux, T.P. Breckon), In Proc. International Conference on Image Processing, IEEE, 2016. (to appear)

Anomaly Detection: [Skip-GANomaly: Skip Connected and Adversarially Trained Encoder-Decoder Anomaly Detection](#) (A. Akcay, A. Atapour-Abarghouei, T.P. Breckon), In Proc. Int. Joint Conference on Neural Networks, IEEE, 2019.

[GANomaly: Semi-Supervised Anomaly Detection via Adversarial Training](#) (S. Akcay, A. Atapour-Abarghouei, T.P. Breckon), In Proc. Asian Conference on Computer Vision, Springer, 2018.

3D CT Overview: [A Review of Automated Image Understanding within 3D Baggage Computed Tomography Security Screening](#) (A. Mouton, T.P. Breckon), In Journal of X-Ray Science and Technology, IOS Press, Volume 23, No. 5, pp. 531-555, 2015.

3D CT Detection & Segmentation:

[Materials-Based 3D Segmentation of Unknown Objects from Dual-Energy Computed Tomography Imagery in Baggage Security Screening](#) (A. Mouton, T.P. Breckon), In Pattern Recognition, Elsevier, Volume 48, No. 6, pp. 1961–1978, 2015.

[Object Classification in 3D Baggage Security Computed Tomography Imagery using Visual Codebooks](#) (G.T. Flitton, A. Mouton, T.P. Breckon), In Pattern Recognition, Elsevier, Volume 48, No. 8, pp. 2489–2499, 2015.

[3D Object Classification in Baggage Computed Tomography Imagery using Randomised Clustering Forests](#) (A. Mouton, T.P. Breckon, G.T. Flitton, N. Megherbi), In Proc. International Conference on Image Processing, IEEE, pp. 5202-5206, 2014