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Anomaly Detection for Security Imaging

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DSDS'16: *Anomaly Detection for Security Imaging*

Anomaly detection provides the **final line of defence** – identifying oddities that we have never seen before, but which a human operator would spot.

DSDS'16: *Anomaly Detection for Security Imaging* **Non-intrusive inspection of cargo**



'Before I say
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tion"

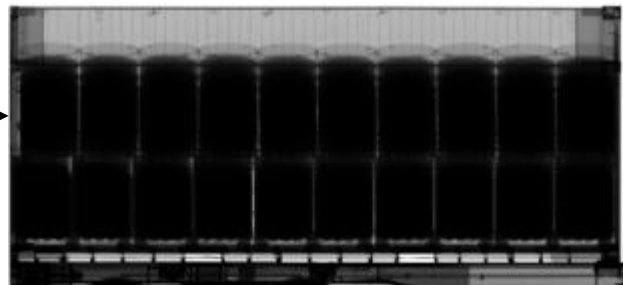
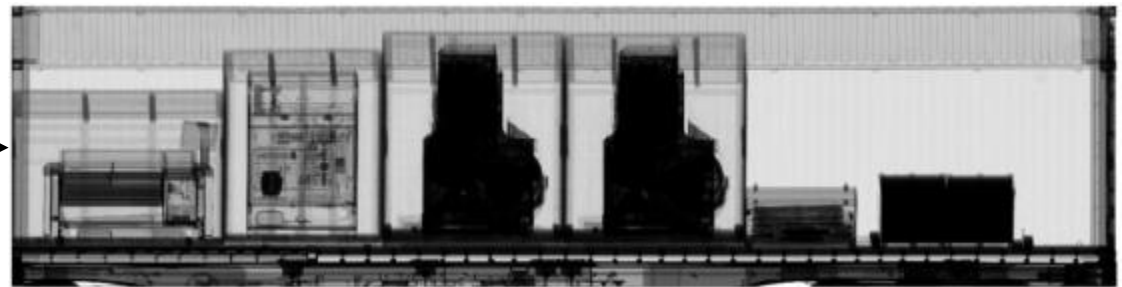
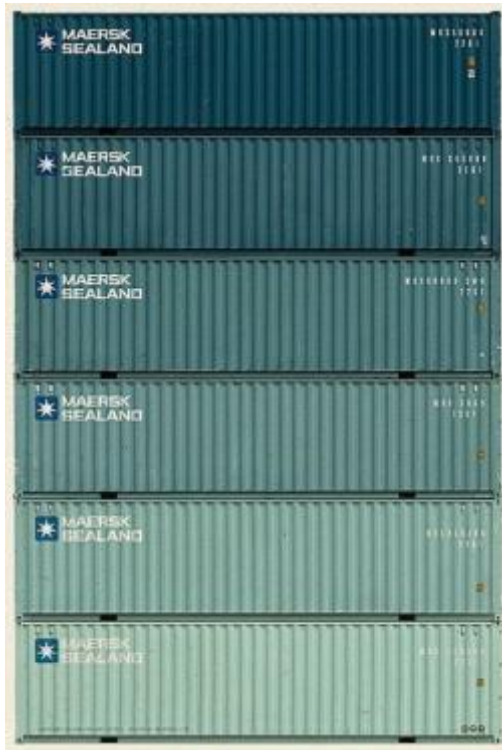
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Minimise false searches; maximise true searches



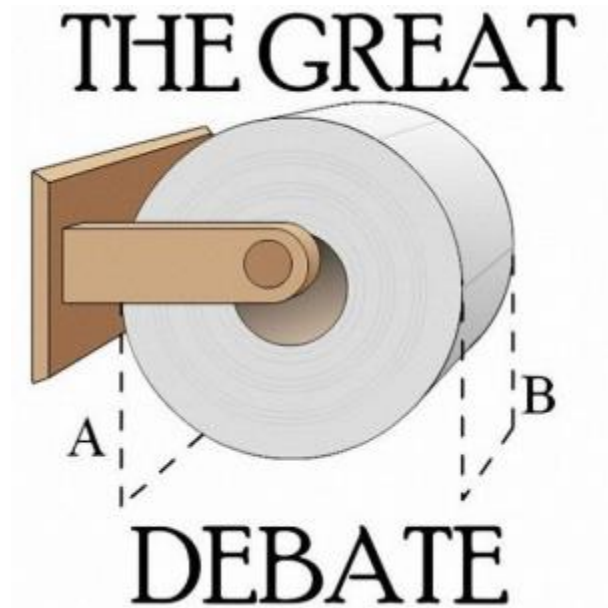
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Appreciable visual challenge for humans



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Existing approaches to automated image analysis



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Existing approaches can be ineffectual

1. Threat data is rare



2. X-ray smart adversaries



3. Never-before-seen-threats



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Inspired by the practice of customs officers, we are developing algorithms to discover visual anomalies, in X-ray images, that are **historically atypical with respect to expected patterns.**

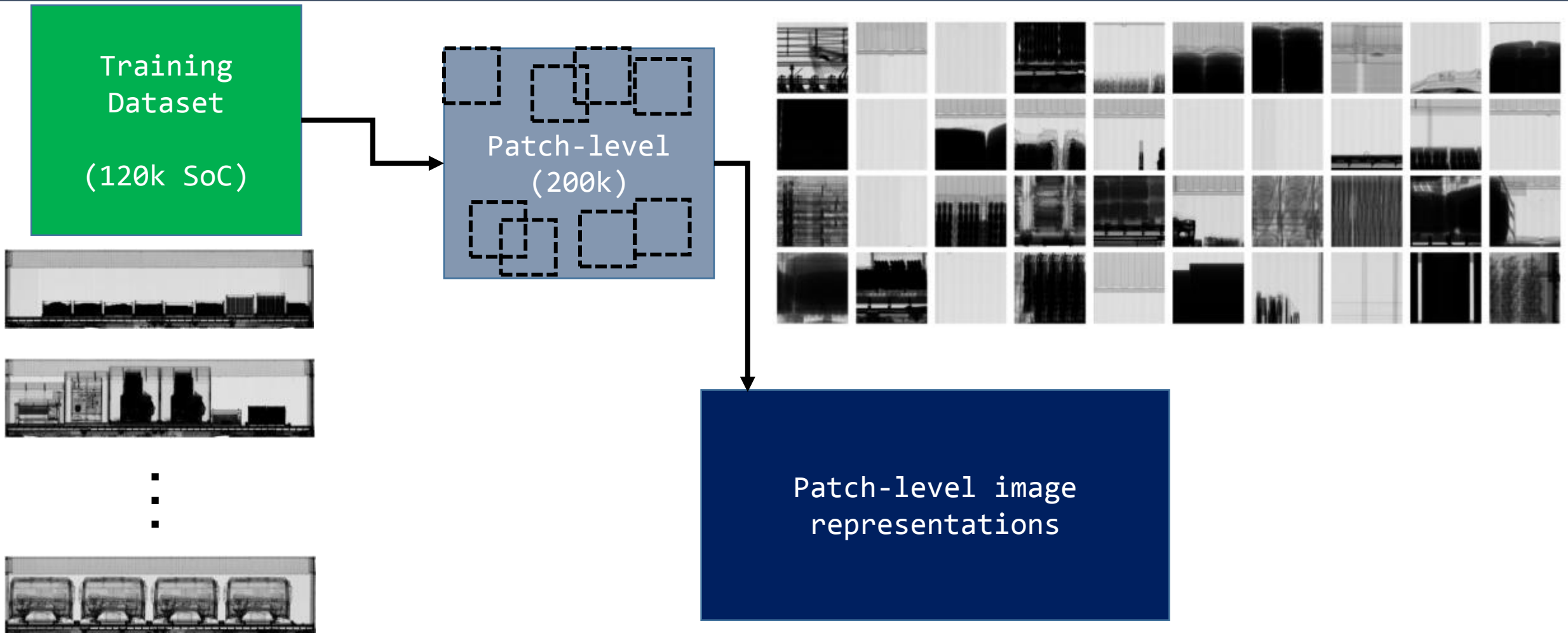
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Our anomaly detection framework is based on:

1. Image representations
2. Anomaly detection relative to those representations

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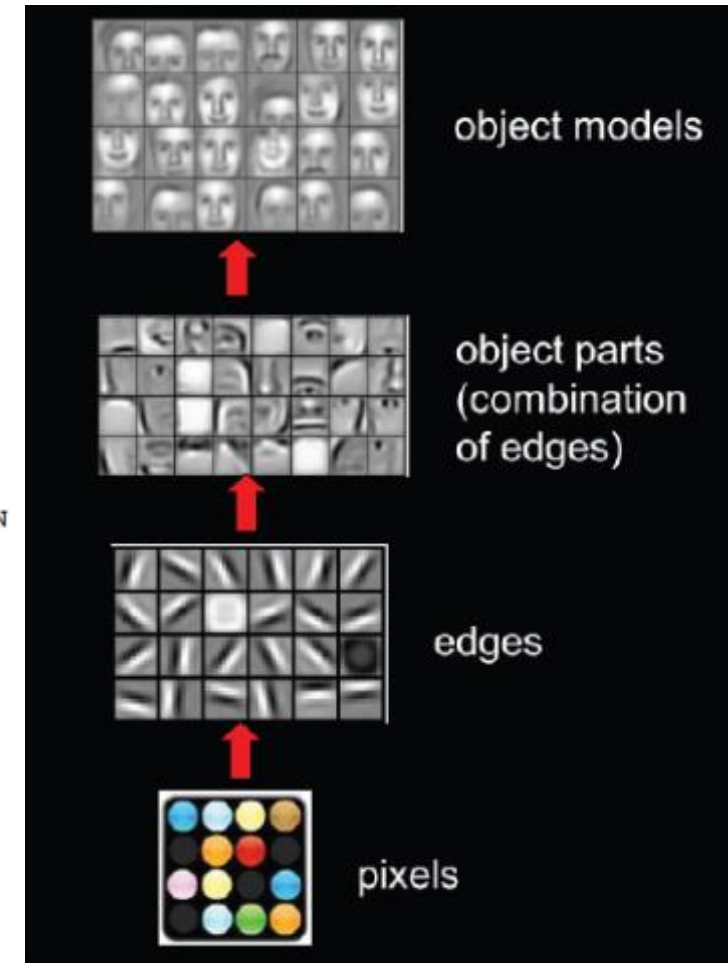
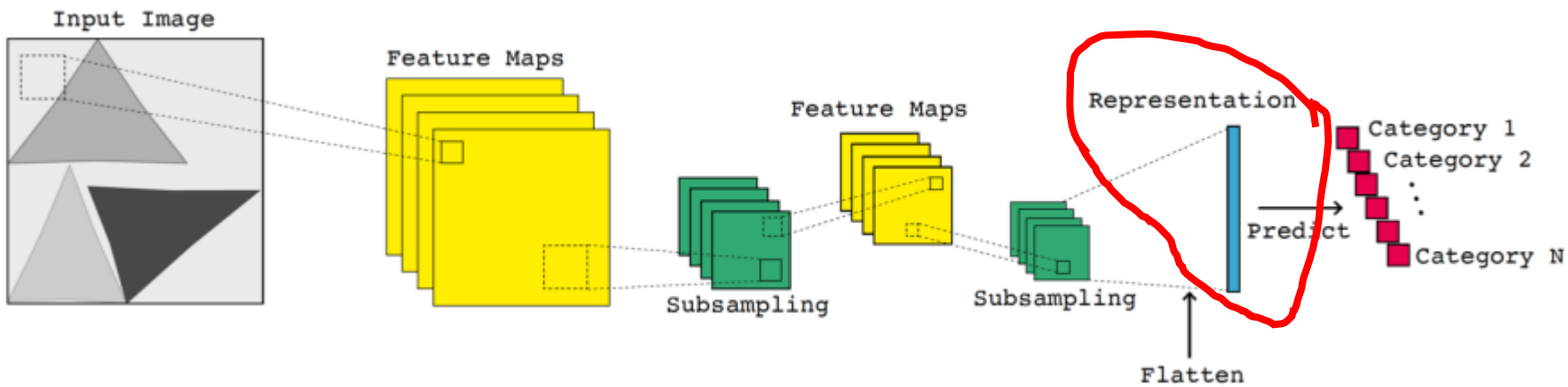
Generation of normal samples



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Patch-level image representations

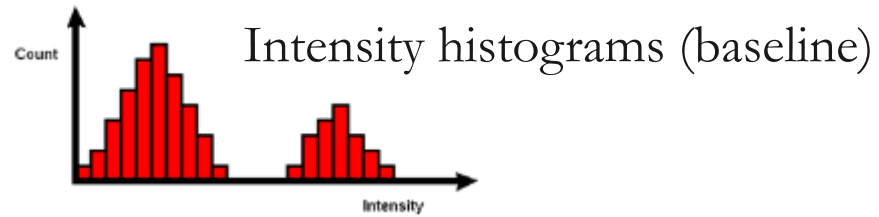
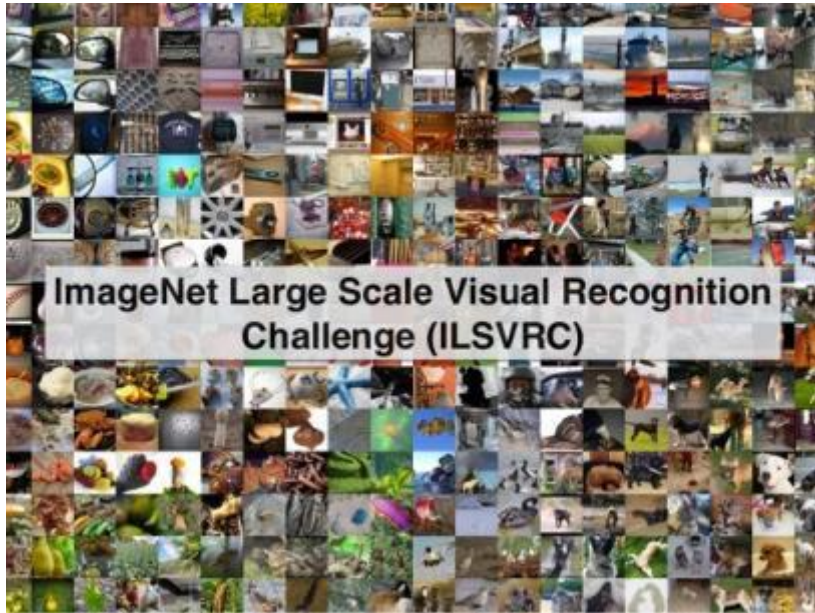
Convolution neural network (CNN)



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Image representations we use...

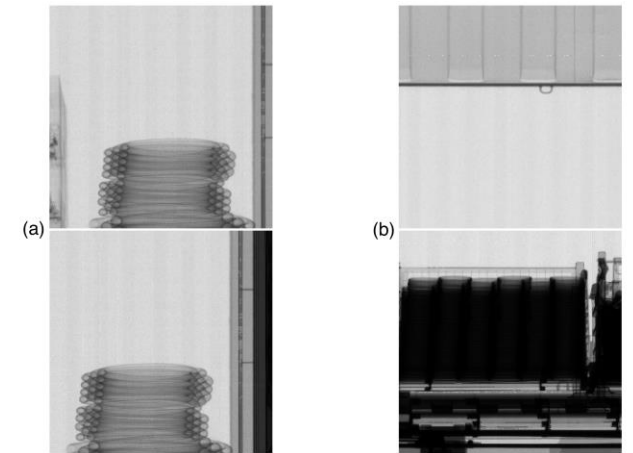
CNN trained on ImageNet



CNN trained to detect cars

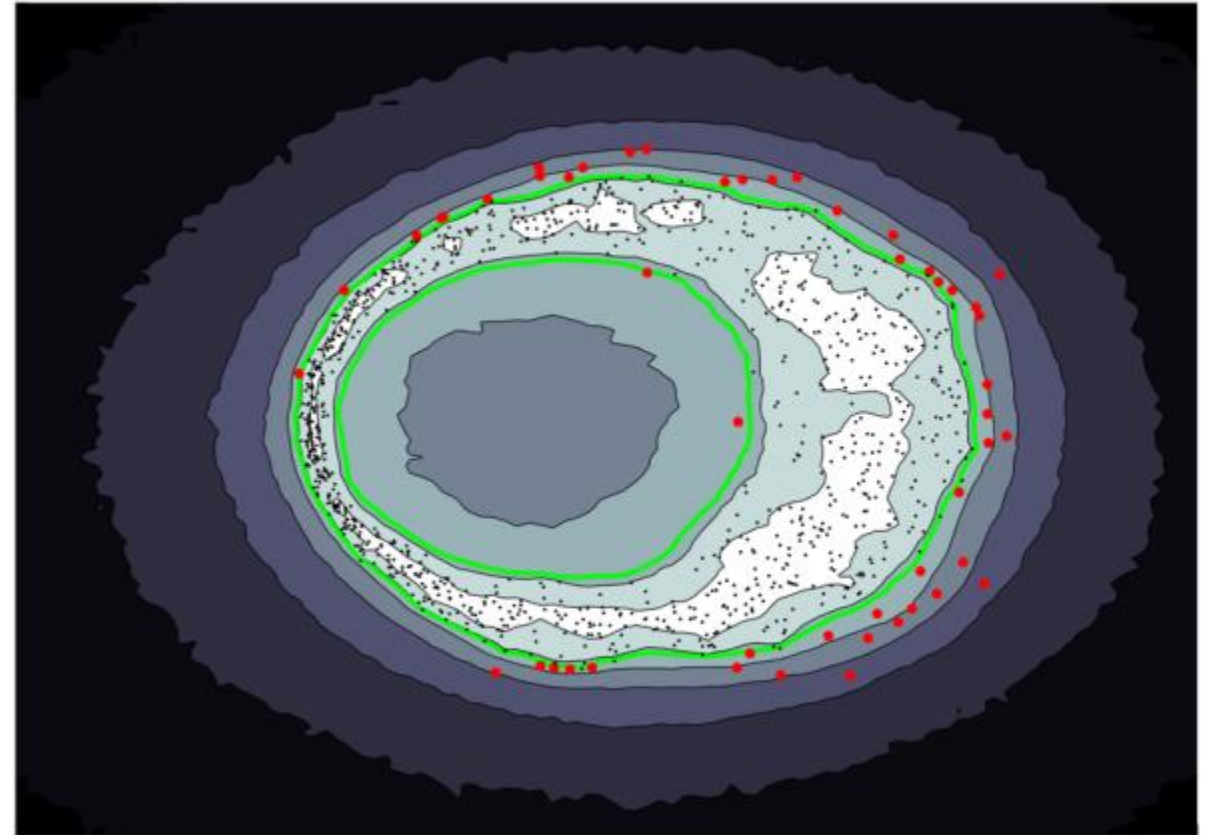
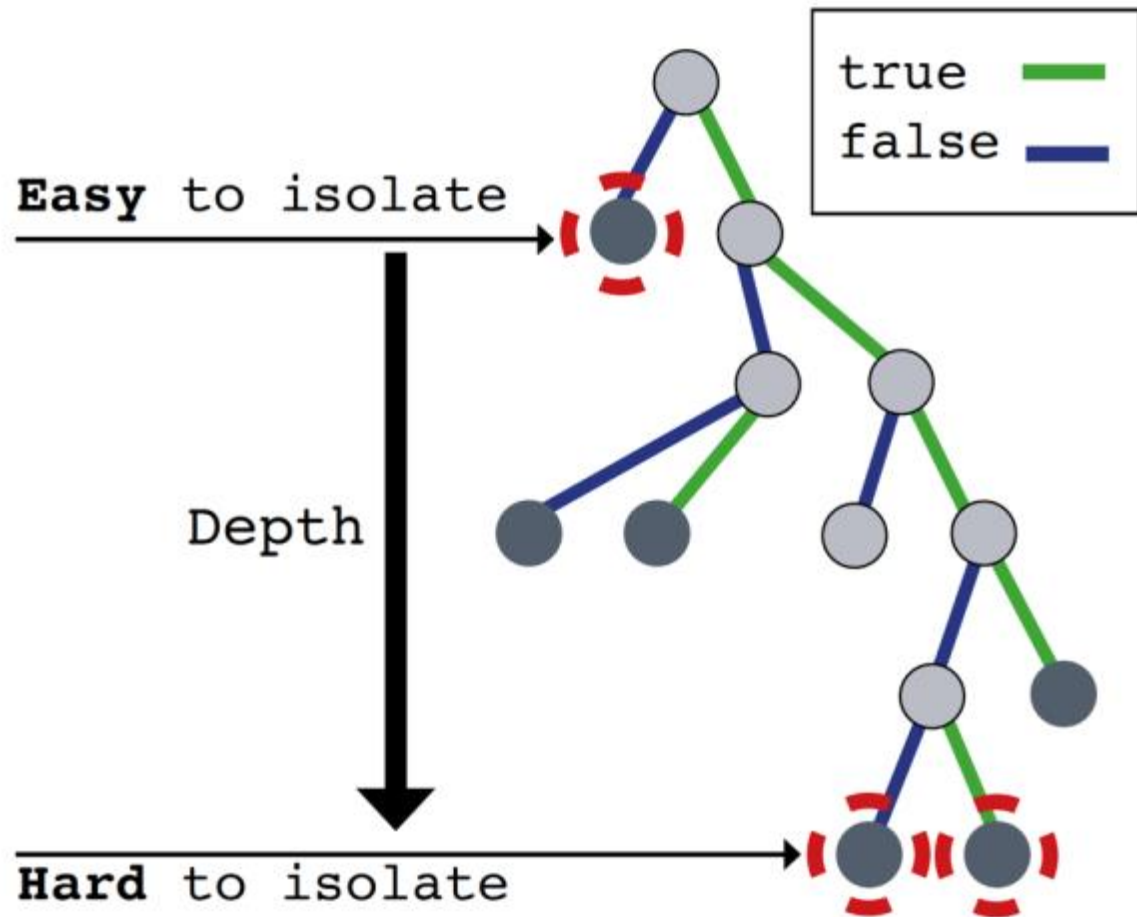


CNN trained on auxiliary task

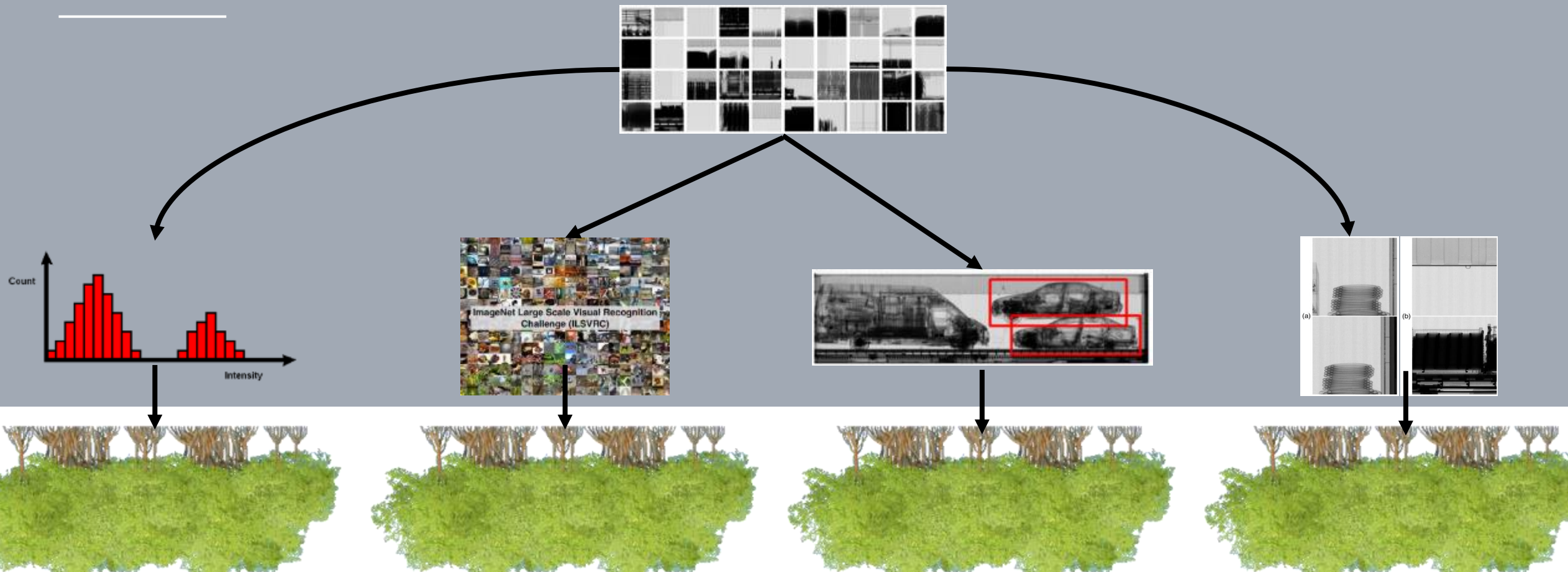


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Forest of random-split trees anomaly detector



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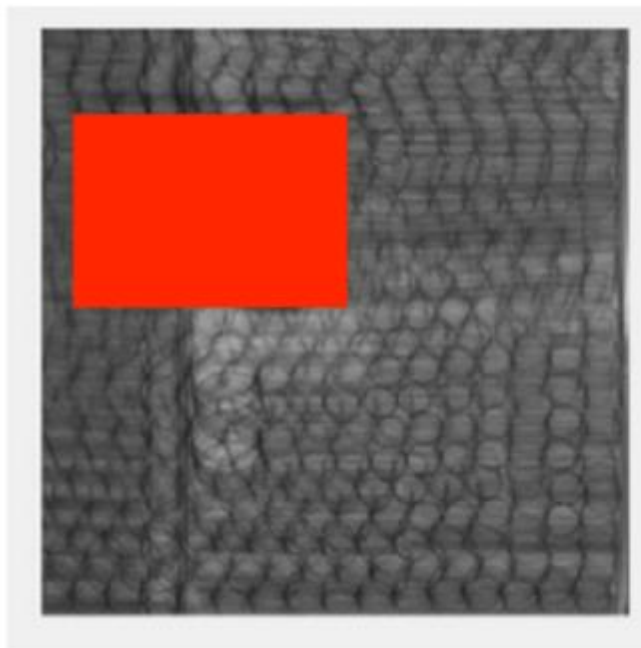


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Generation of anomaly samples for testing

Example: Threats projected into SoC patches

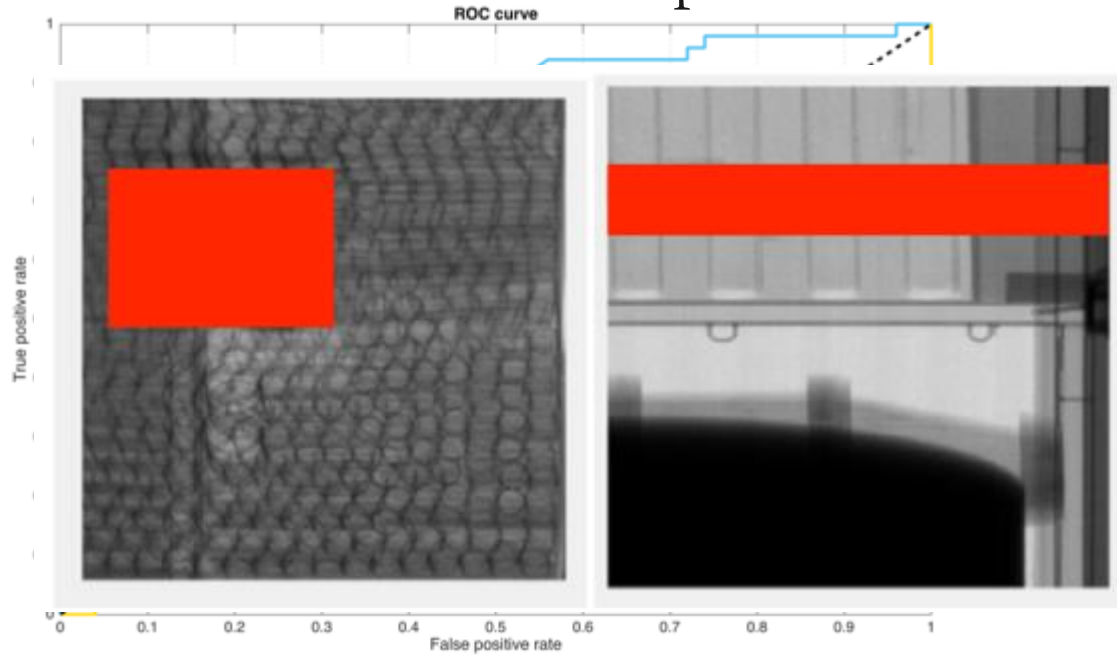
Example: Threats



the bender

DSDS'16: *Anomaly Detection for Security Imaging* Results

12.5k **threat** patches

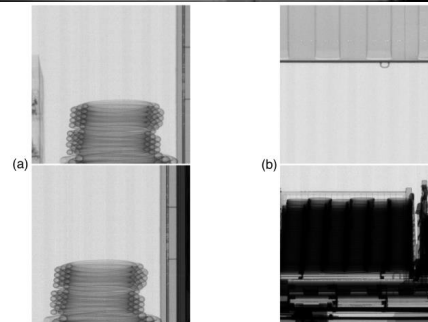
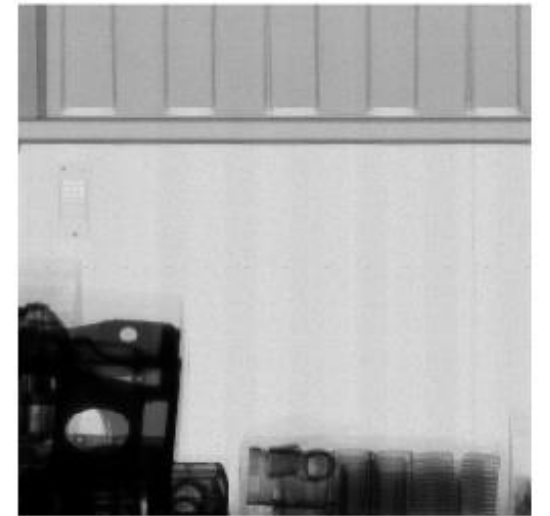
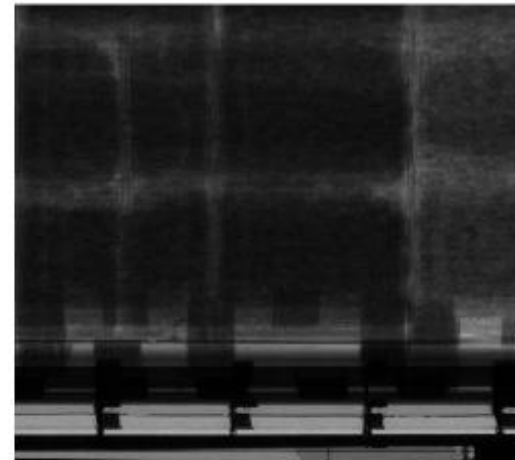


12.5k **normal** patches

Representation

AUC

In
CNN
CNN tr
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Summary:

Our *preliminary* results demonstrate that FRST is a viable approach to AD, when applied to representations learnt for a comparable task.

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Future work:

1. CNN architecture and auxiliary task evaluation
2. Image-level heat maps by fusion of patch-level results

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Thanks for listening!

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