

Robust Object Detection via Kronecker Tensor Decomposition

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Tensor Completion

$$\min ||P_{\Omega}(X) - P_{\Omega}(M)||_F^2$$

s.t. $rank(X) = R$

Reformulated via auxiliary variable C:

$$\min ||X - C||_F^2$$

s.t. $rank(X)=R, P_{\Omega}(C) = P_{\Omega}(M)$

Random KTD FRAMEWORK

Kronecker Tensor Decomposition (KTD):

$$X = \sum_{r=1}^R \sigma_r X_r^{(1)} \otimes X_r^{(2)} \otimes \dots \otimes X_r^{(M)},$$

Random KTD algorithm:

- Compression:** Apply random projections to each mode of the tensor, reducing dimensions from I_n to k_n where $k_n \ll I_n$
- Decomposition:** Perform KTD on the compressed tensor
- Reconstruction:** Recover the full decomposition using the random projection matrices

ROBUST DETECTION PIPELINE



$$X^{(n)} \leftarrow \mathcal{L}(C^{(n)}) \quad \leftarrow \text{KTD Restore}$$

$$C^{(n+1)} \leftarrow \Omega \odot M + (1 - \Omega) \odot X^{(n)} \quad \leftarrow \text{Estimate the masked pixels}$$

Visualization of restoration



Step 0

Step 15

Step 30

Key Insight: Masking 70% of pixels disrupts structured adversarial perturbations. KTD completion restores clean image structure — enabling the detector to succeed despite attacks.

EXPERIMENTAL RESULTS — YOLOv11m on COCO

Attack Type	Image	Precision	Recall	F1	mAP
No Attack	Original	74	61	67	59
TOG-Localization	Attacked	29	29	29	23
	KTD-Restored	71	42	53	44
TOG-Vanishing	Attacked	5	37	9	22
	KTD-Restored	73	45	56	47
TOG-Mislabeled	Attacked	51	20	29	15
	KTD-Restored	76	47	58	46
TOG-Fabrication	Attacked	33	13	19	19
	KTD-Restored	76	45	56	47

+21%

mAP gain vs attacked
(Localization: 23→44)

76%

Precision restored
(Mislabeled & Fabrication)

30

Optimal iteration
count for restoration

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