

#### Introduction

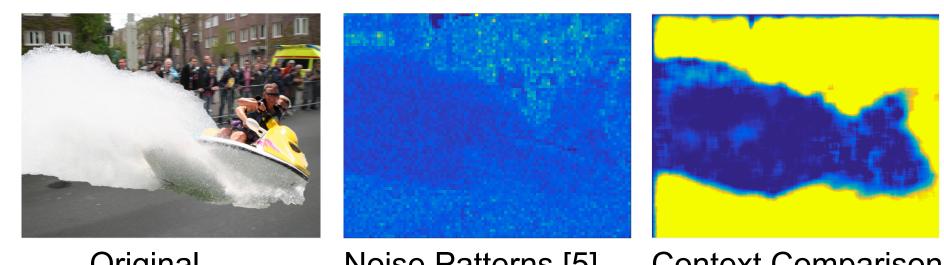
- We leverage the power of wide-scale image search to find contextual clues in related media. These clues allow us to more accurately detect and localize forgeries in tampered images. We test our method using the Nimble Challenge 2016 dataset[1], provided by NIST, with millions of distractor images provided by RankOne[2].
- We develop and analyze the performance of 4 novel image comparison techniques to extract contextual clues in the presence of synthetic noise and perturbation, to analyze robustness to "noisy clues"
- We benchmark our contextual clue-based methods against well-established traditional PDIF methods from [3]



Images from the NIST NC2016 Forgery Dataset

#### **Comparison Algorithms**

- PSNR of Gaussian Image Residual (IRPSNR)
- Pseudo-PRNU Patch Comparison (PRNU)
- Structural Similarity Comparison (SSIM)
- HSV Histogram Patch Comparison (HIST)
- PatchMatch2 [4] Random Comparison





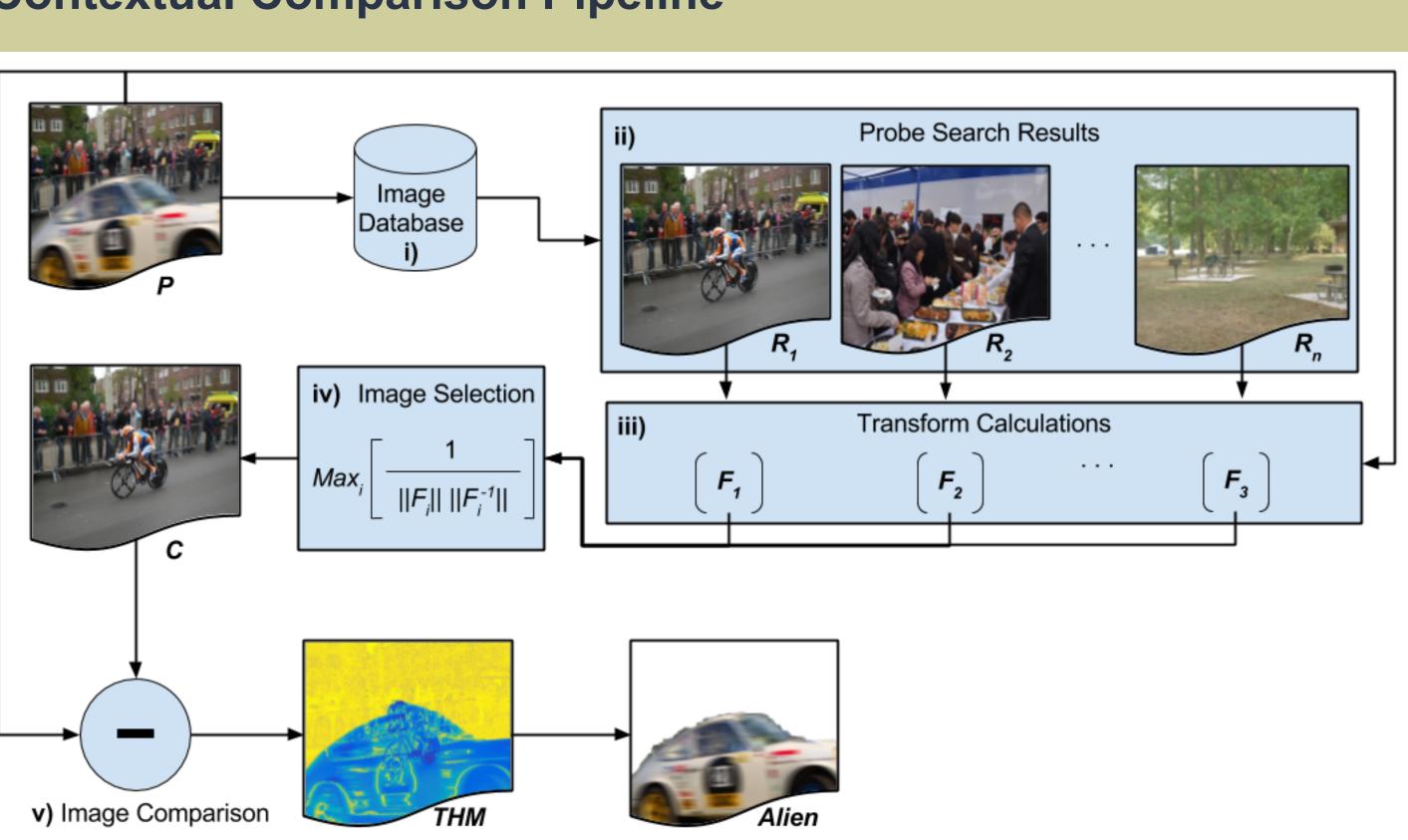
Noise Patterns [5]

**Context Comparison** 

## **Spotting the Difference: Context Retrieval and Analysis for Improved Forgery Detection and Localization**

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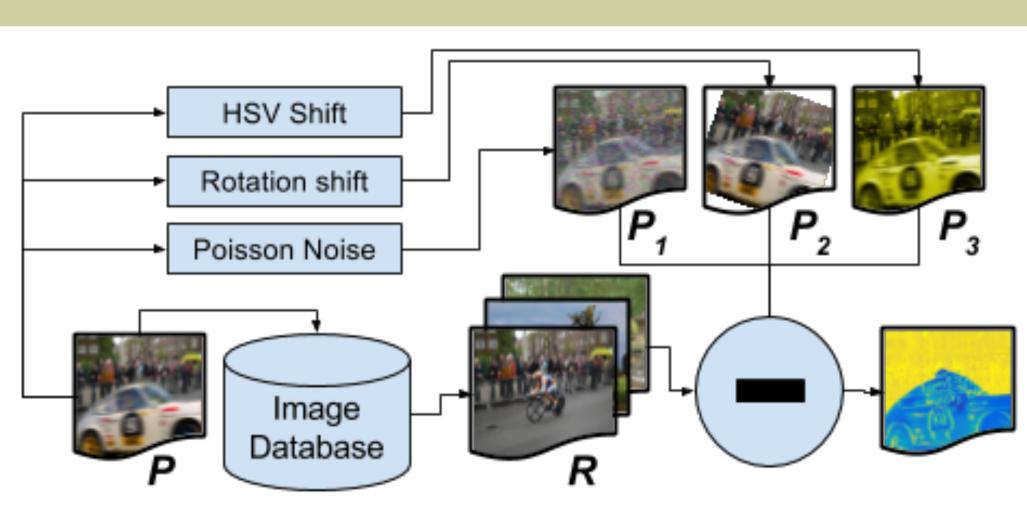
### **Contextual Comparison Pipeline**



- An image *P* is is used to query a database of over 1M images
- Top N Related results  $R_N$  are collected for P İİ.
- The transforms between *P* and *R* are calculated using SURF matching and affine homographies
- iv. Best non-duplicate image C is selected from  $R_N$  using the Reciprocal Condition measure
- Images are compared using 1 of 4 methods to build a Tamper Heat Map (THM)

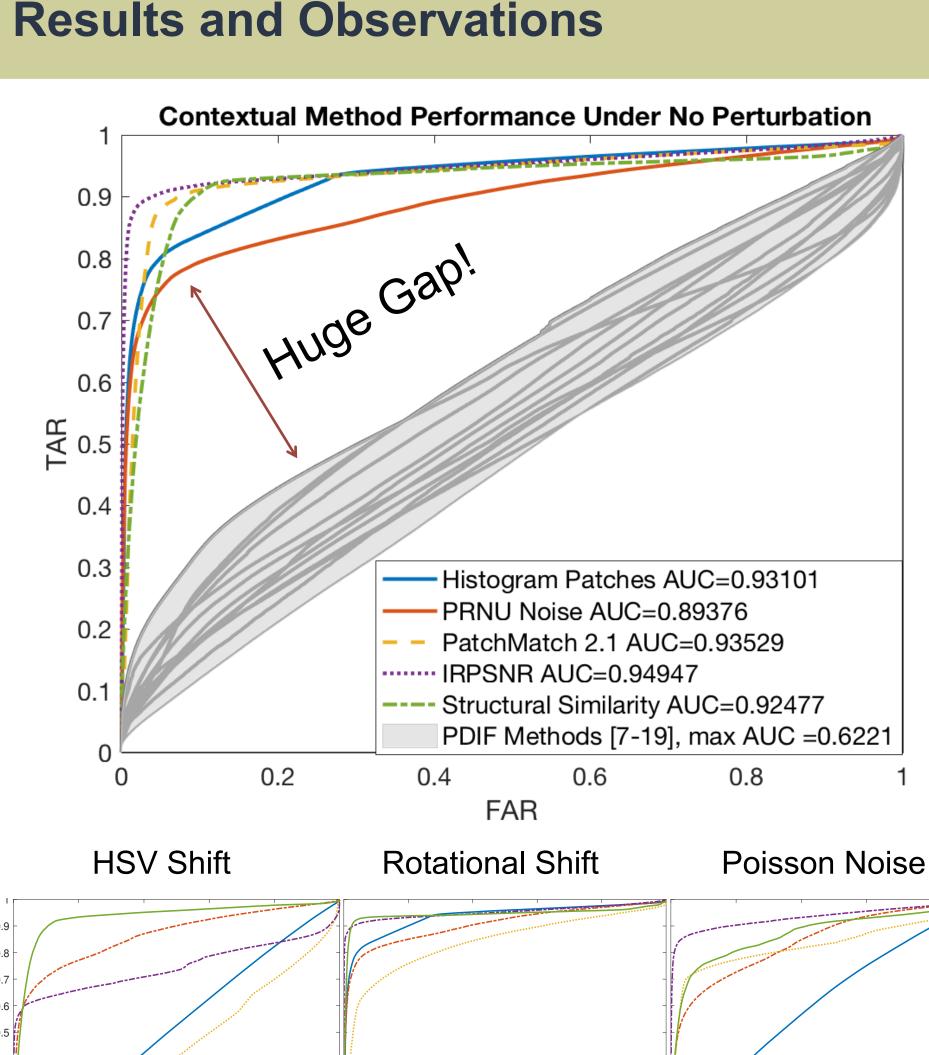
#### **Perturbation Pipeline**

- Each query *P* is perturbed in 3 ways
- Each  $P_i$  is used to generate a new THM
- Image comparison methods are analyzed for robustness under perturbation



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 Context Comparison greatly out-performs Stateof-the-art Passive Image Forensics methods

Histogram Patches AUC=0.9334

- PRNU Noise AUC=0.90544

IRPSNR AUC=0.95118

PatchMatch 2.1 AUC=0.8363

- SSIM comparison is invariant to HSV color space and rotational shifts, while Noisy images are better tackled by IRPSNR
- [1] NIST, "The 2016 Nimble chal-lenge evaluation dataset," https:// www.nist.gov/itl/iad/mig/nimble-challenge, Jan. 2016.
- [2] http://medifor.rankone.io/

listogram Patches AUC=0.5169

PRNU Noise AUC=0.86373

- IBPSNR AUC=0.74549

PatchMatch 2.1 AUC=0.44091

- [3] L. Gaborini, P. Bestagini, S. Milani, M. Tagliasacchi, and S. Tubaro, "Multiclue image tampering localization," in2014 IEEE WIFS, 2014, pp.125–130.
- [4] C. Barnes et. Al, "Patch-match: Randomized correspondence algorithm for structural imageediting,"ACM TOG
- [5] Mahdian, Babak, and Stanislav Saic. "Using noise inconsistencies for blind image forensics." IVC 27, no. 10 (2009): 1497-1503.

