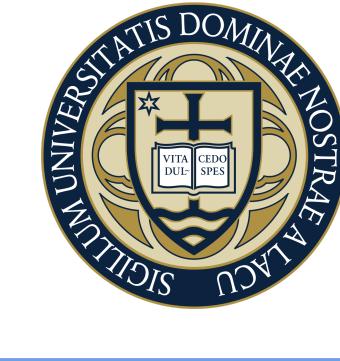


Motif Mining: Finding and Summarizing Remixed Image Content



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We aim to provide a robust, accessible and fast tool for exploring large datasets of unsorted and unlabelled images. By discovering emerging trends in online social media artifacts human rights watchers can more quickly address and refute false and misleading narratives.

- Telegram posts related to the Ukraine-Russo War
- Accounts identified by subject matter experts
- Now over 4.5M posts spanning 2015 - Present
- Collection on-going, a cleaned snapshot is in preparation



Motivation

Data set

Basic Pipeline

- 1.) Extract local and global image features
- 2.) Append single global 'tag' to each local feature
- 3.) Feed images to indexing system (FAISS)
- 4.) Use retrieval function to generate similarity graph
- 5.) Cluster similarity graph to discover visual motifs

• Use Amazon Mechanical Turk to perform 'impostor-host' test to measure motif similarity with a human feedback component. The impostor-host test ask humans to identify the image in a set of five that is from a different cluster from the other four. We use this as a proxy-measure for human-salient cluster inter-relatedness.

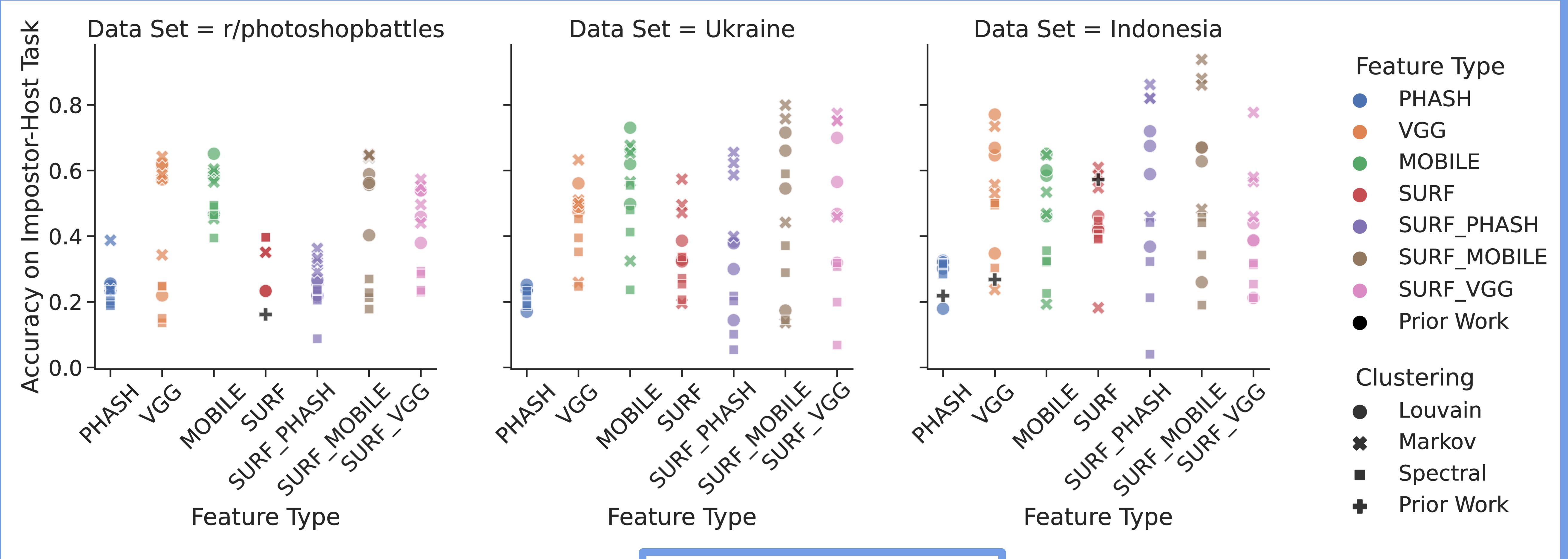
• Pipeline tested on 3 data sets, /r/photoshopbattles (10,588 images), Indonesia (44,612 images), and our Telegram new data set (16,433 images).

• State-of-the-art accuracy results were achieved on all three data sets.

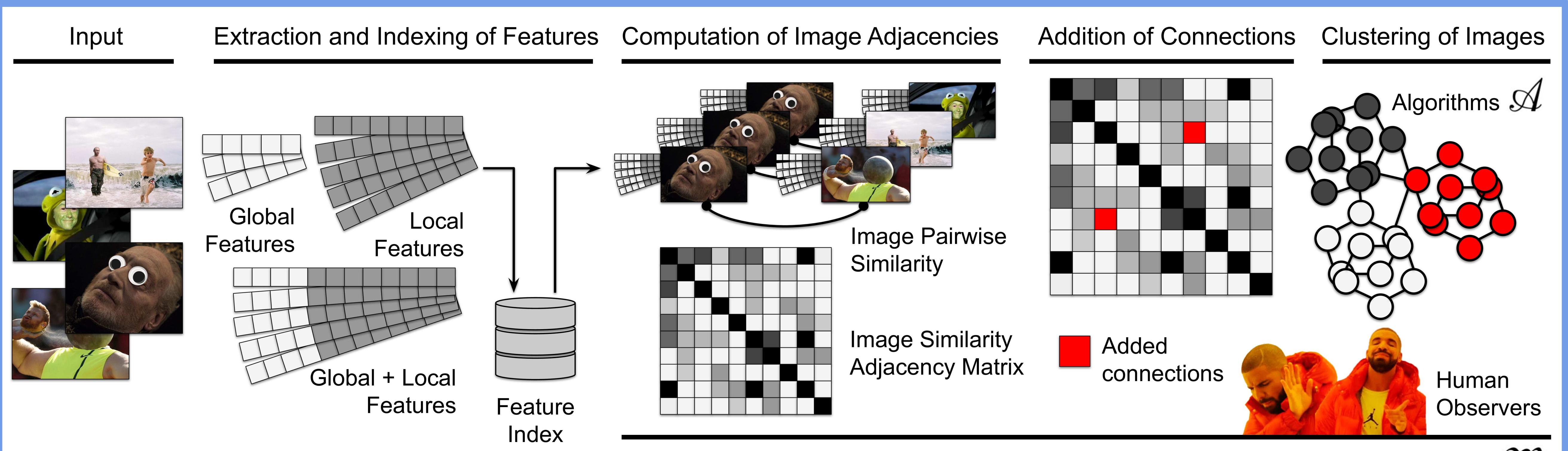
• Globally-tagged surf features (SURF-MOBILE) gave the best results on the Imposter-Host test with Louvain clustering (taking into account image distribution).

• Additionally near 10x speedups were seen.

Methodology and Experiments



Experimental Results



Mapping Process \mathcal{M}

