



NIST Open Media Forensics Challenge (OpenMFC)

Presented by Haiying Guan

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What

Our Mission

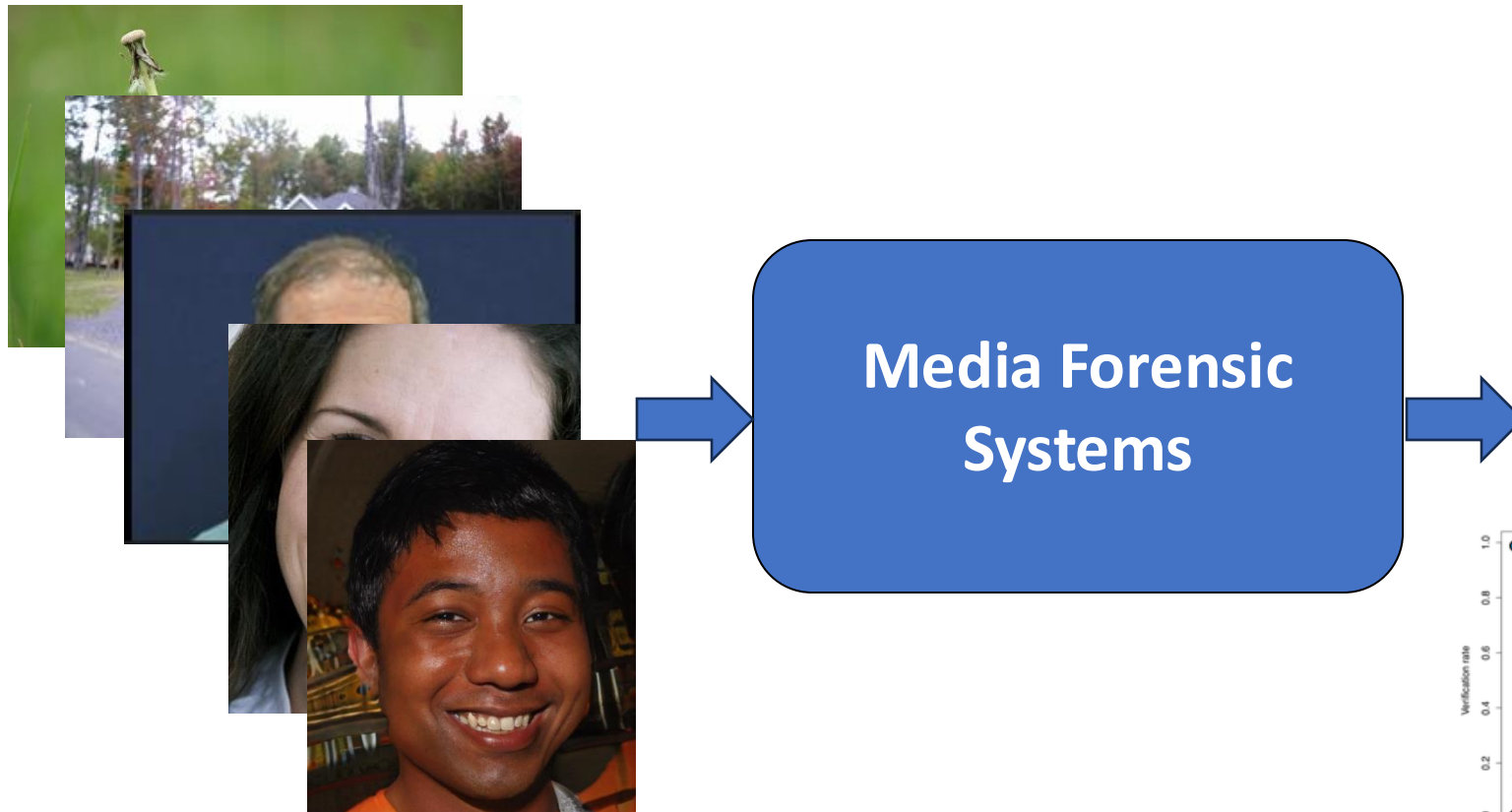
How

Insights

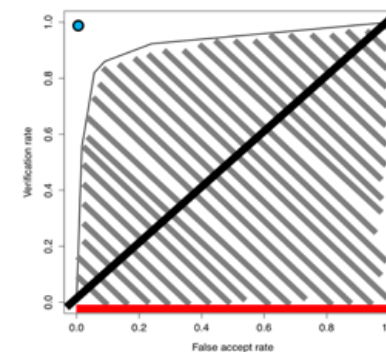
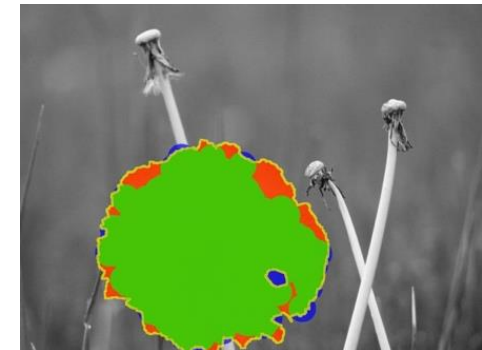
Ongoing Effort

Vision/Plan

What is OpenMFC?



System Performance Reports



RANK	SUBMISSION ID	SUBMISSION DATE	TEAM NAME	SYSTEM NAME	AUC	CDR@0.05FAR
1	90	2021-07-10 09:56:14	TEAM_ID_1	test	0.689716	0.207018
2	93	2021-07-30 18:13:11	TEAM_ID_1	test	0.683956	0.187135
3	75	2021-06-14 04:12:30	TEAM_ID_2	dry-run	0.554261	0.012865
4	52	2021-06-01 15:41:20	TEAM_ID_2	dry-run	0.547125	0.009357
5	82	2021-06-23 09:21:26	TEAM_ID_1	System_ID_1	0.500033	0.051077

OpenMFC is a public evaluation program that invites participants worldwide to contribute to media forensics research and advance state-of-the-art forensic technologies for image and video analysis.

Why Are We Building OpenMFC?

- Generative and manipulated media are ubiquitous
- “Seeing is not believing”
- Impact Areas:
 - Academic misconduct
 - Journalism integrity
 - Criminal and terrorism investigation
 - Geospatial intelligence
 - Cybersecurity
- Benchmark evaluations support forensic researchers in developing advanced technologies

Self-report

vs.

External report

- System performance is highly sensitive to data and task variations.
- Self-evaluation reports may not accurately reflect real-world performance.
- Neutral and unbiased benchmark evaluations are essential for reliable assessment.

One-time eval.

vs.

Recurring eval.

- Media manipulation and generation technologies are continuously evolving.
- Recurring evaluations are more effective than one-time assessments for tracking technological advancement.

OpenMFC Goals and Challenges

Goals



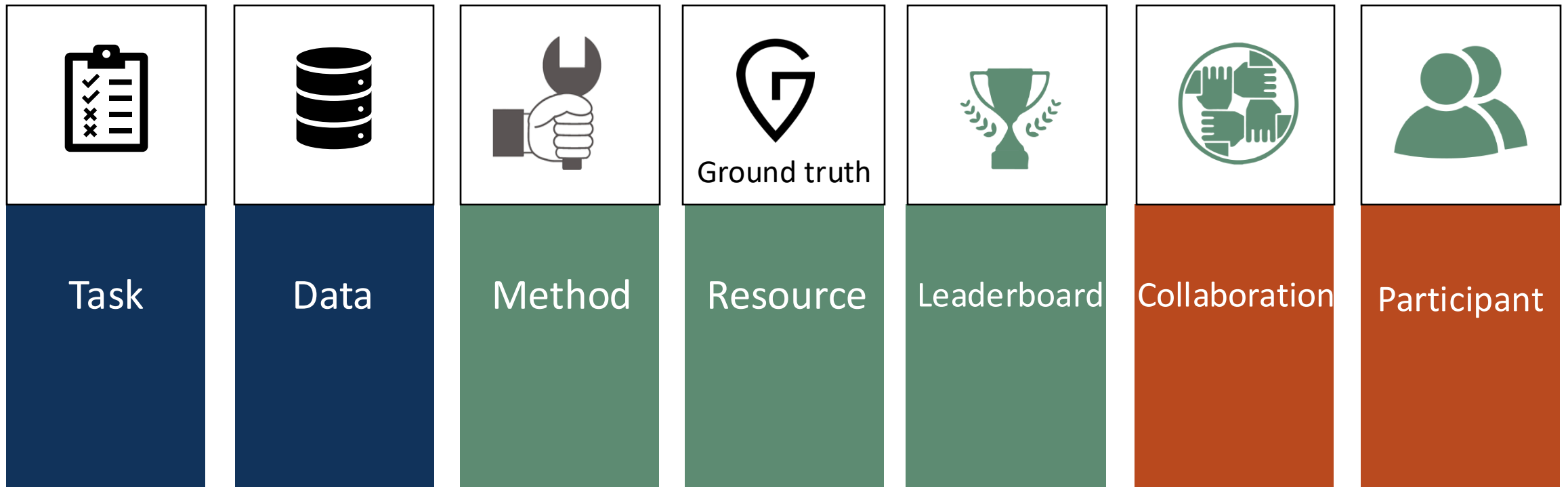
- Foster research in media forensics.
- Gain insights into state-of-the-art techniques through recurring evaluations.
- Provide continuous, year-over-year comparison reports.
- Assist researchers in improving their systems through detailed performance analysis.
- Support the transition from lab-based research to real-world field evaluations.

Challenges



- **Media forensics vs. Anti-forensics:** A non-linear trajectory in technological advancements.
- **Target drift:** Rapid evolution of technologies (e.g., GANs, deepfakes, stable diffusion, DALL·E 2, etc.)
- **Data Challenges:** Issues with availability, scalability, privacy, and high resource demands.

OpenMFC Evaluation Design



Task Design Strategy

Detection

Identifying whether a given piece of media has been manipulated or synthesized

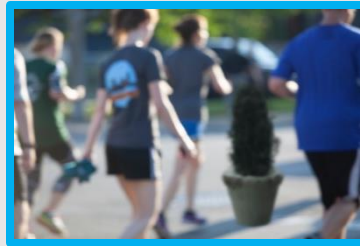


Y
?
N

Deepfake Detection

Verification

Determining whether a given claim is supported by available evidence

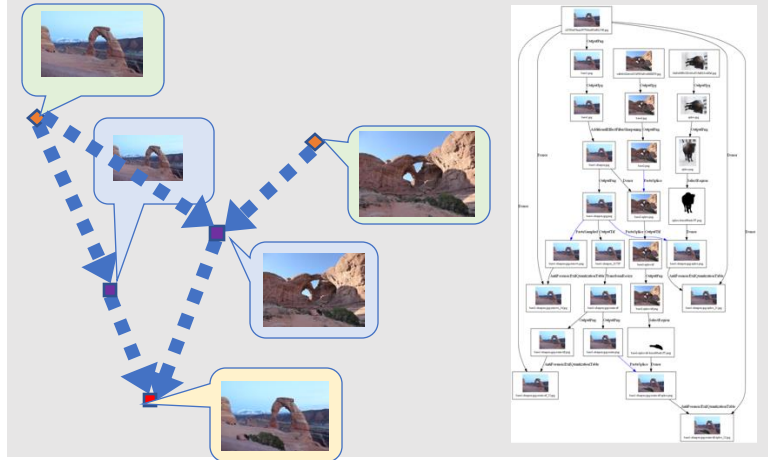


? Boston Marathon

Event Verification

Association

Establishing relationships or connections between different elements



Provenance Graph Building



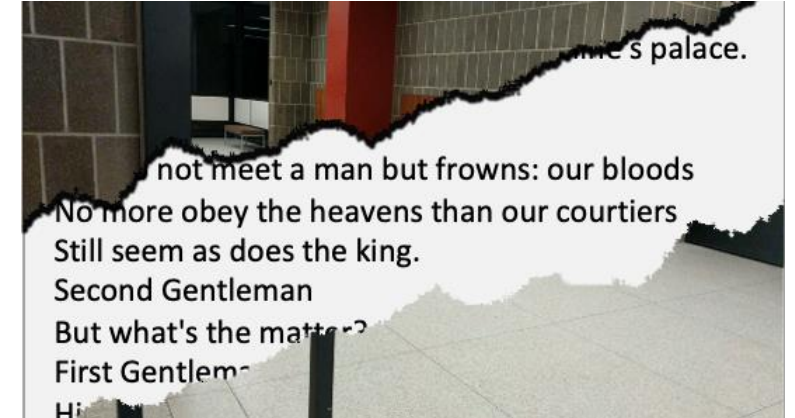
Manipulation Detection (MD)

- Image_MD (IMD)
- ImageSplice_MD (ISMD)
- Video_MD (VMD)



Deepfake Detection (DD)

- Image Deepfake/GAN Detection (IDD)
- Video_DD (VDD)



Steganography Detection (StegD)

- StegD

Example Task: Deepfake Detection

System Input

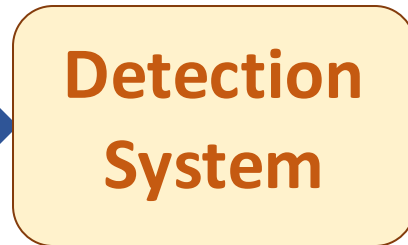
System

System Output

Leaderboard

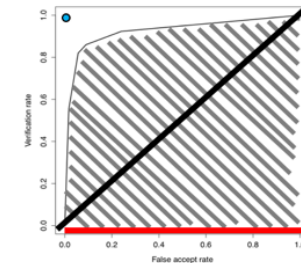


Probe media



Confidence score

Receiver Operating Characteristic (ROC)



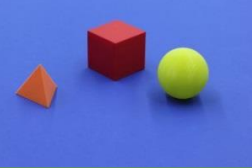
Area Under the Curve (AUC)

Correct Detection (CD) at False Alarm Rate 5%


RANK	SUBMISSION ID	SUBMISSION DATE	TEAM NAME	SYSTEM NAME	AUC	CDR@0.05FAR
1	90	2021-07-10 09:56:14	TEAM_ID_1	test	0.689716	0.207018
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5	82	2021-06-23 09:21:26	TEAM_ID_1	System_ID_1	0.500033	0.051077

Figure: An OpenMFC Evaluation Task Example

Evaluation Data Challenges



vs.




Lab Algorithm Evaluation
vs.
Real-World Field Application


Curse of Dimensionality & Dynamic Shift

Media complexity
Manipulation complexity

- Hundreds of manipulation tools and operations
- Emerging technologies: Stable Diffusion, Deepfake, GAN, Generative AI, etc.
- Anti-forensic technologies



Ground truth labeling



Post interpretation is nearly impossible

- Has it been manipulated? Was it malicious?
- Who performed the manipulation?
- What was the original media source?
- Which editing tool was used?
- What specific operations were applied?
- How was the media manipulated?
- Where did the manipulation occur?

Concerns & Constraints

Availability
Scalability
Privacy, Ethics, Consent
Resource intensity (time, labor, cost)
Distribution

Media Manipulation Journaling

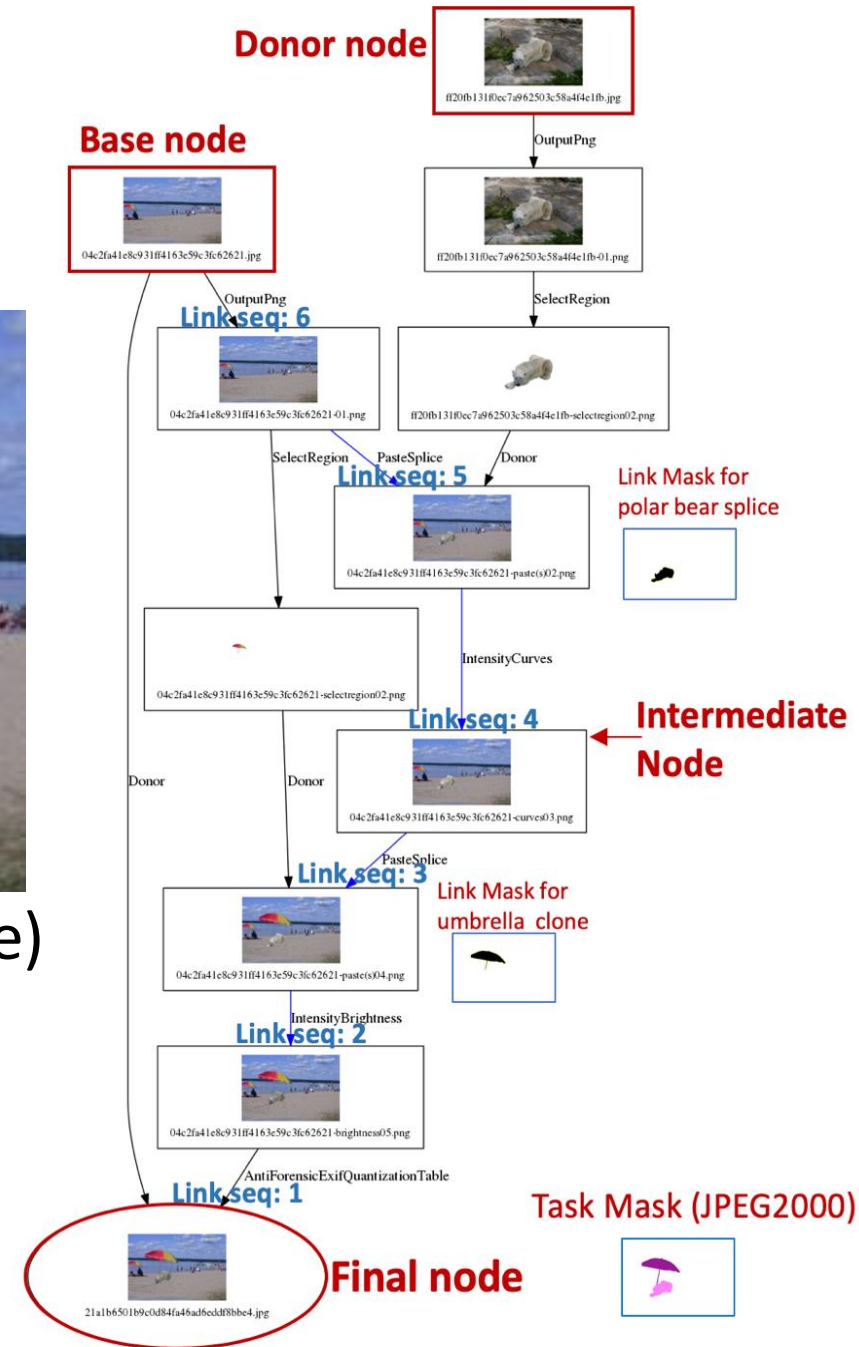


a. Manipulated image



b. Original image (Base image)

Figure: Example of a Manipulation Journal Log



1 Data Generation

- Establish a dedicated data generation team that operates independently from the analytic teams
- Prevent evaluation and data bias
- Encourage analytics teams to focus on developing and evaluating more realistic test scenarios.

3 Annotation: Journaling Tools

- Manipulation Journaling Tool (JT)
- Automatic Journaling Tool (Auto-JT)
- Extended Journaling Tool (Extended-JT)

2 Annotation: Human + Machine

- Human Manipulation (realistic)
- Automatic Manipulation (reduce cost)
- Extended Manipulation (special study)

4 Real-World Simulation

- Post-Processing
- Anti-Forensics
- Recapture

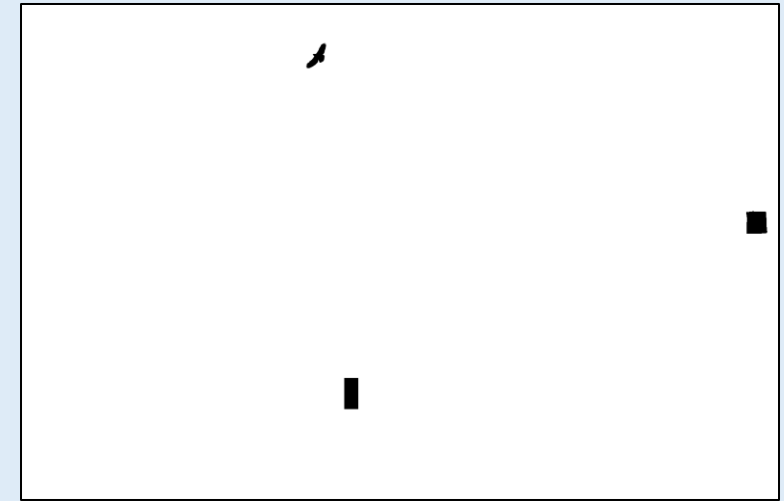
Image Manipulation Examples



1.1 Manipulated image



1.2 Original image



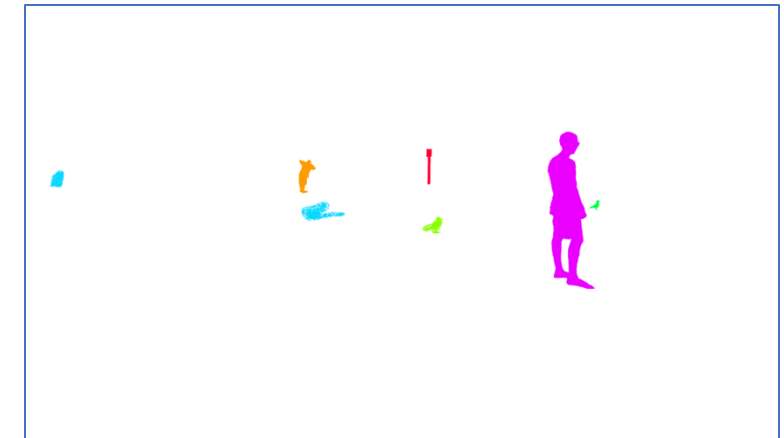
1.3 Manipulation Mask



2.1 Manipulated Image



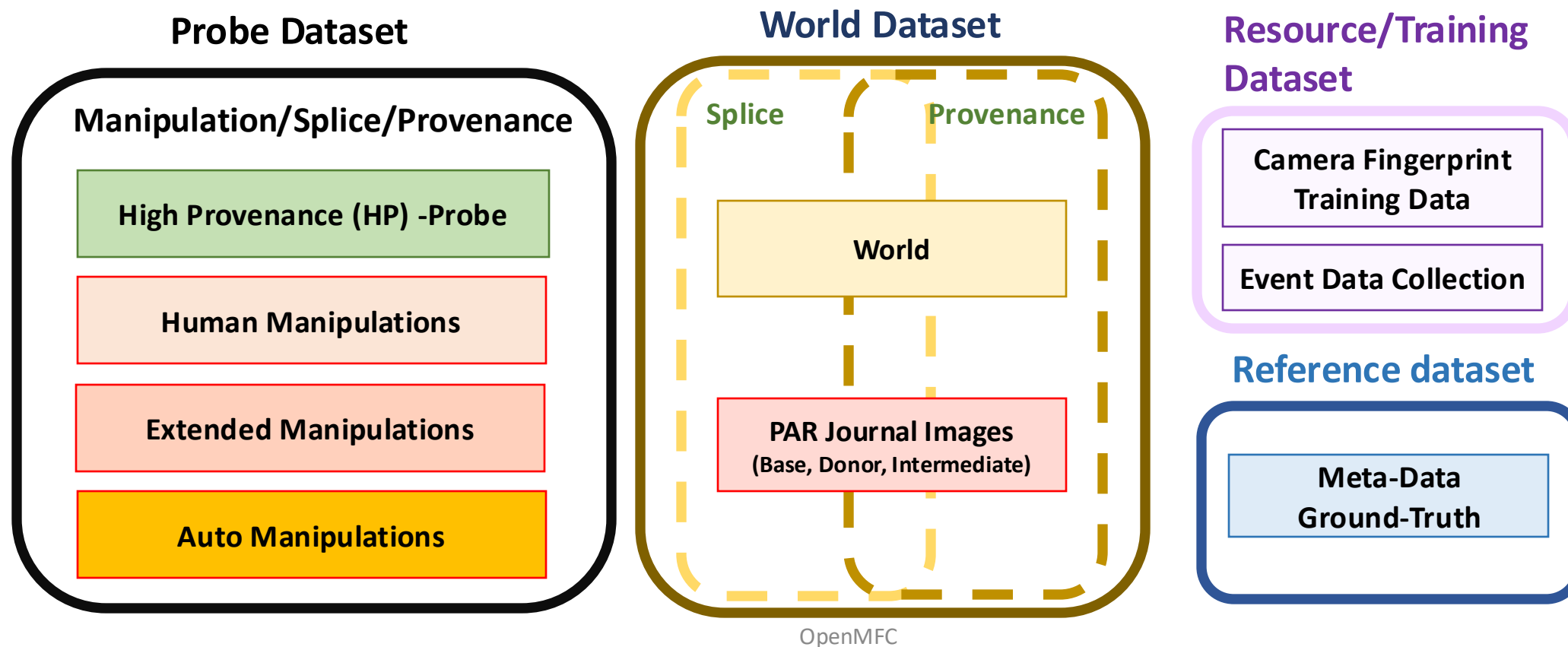
2.2 Original Image



2.3 Manipulation Mask (JPEG 2000)

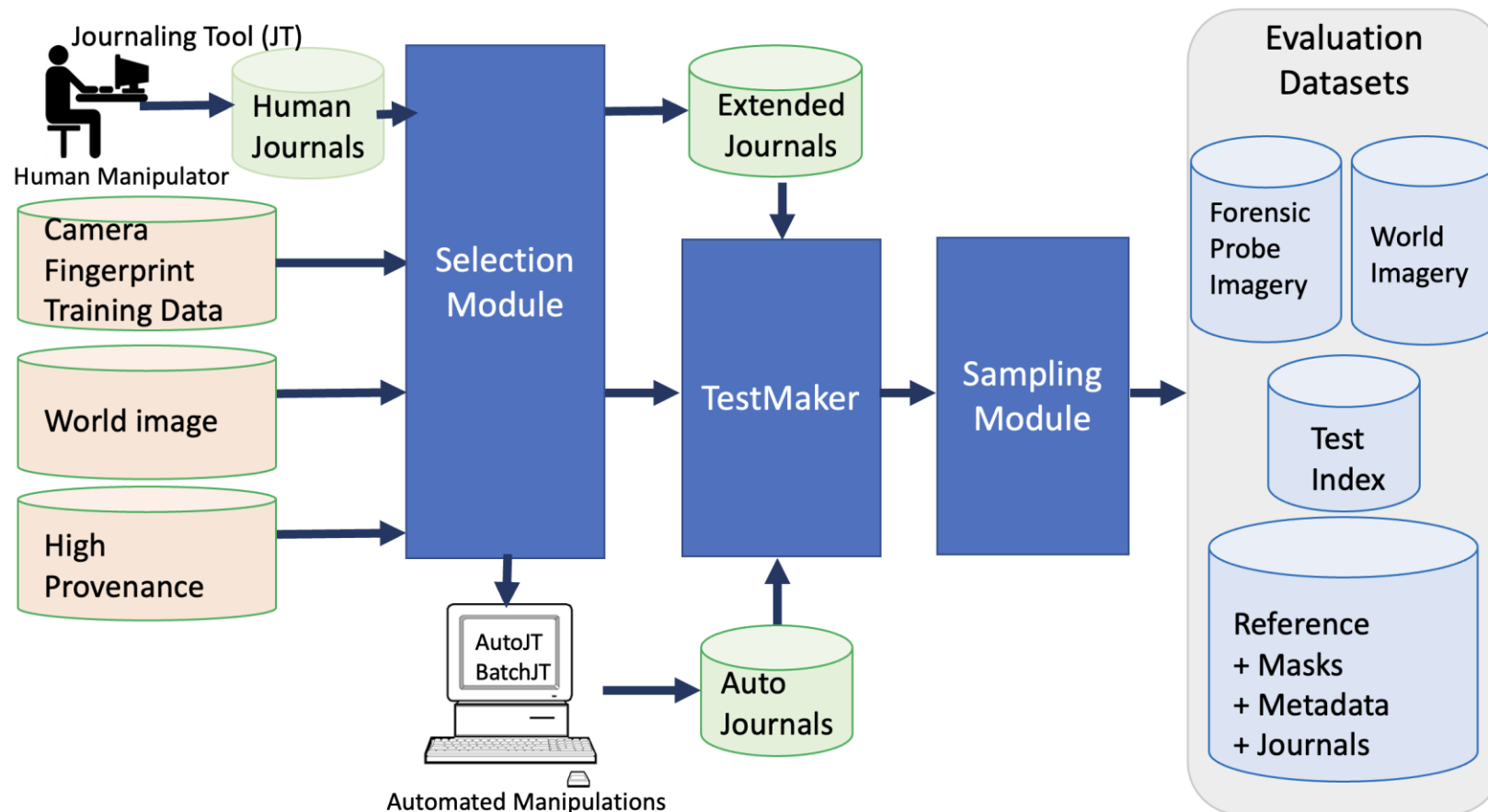
Evaluation Data Overview

We optimize data utilization across all tasks while minimizing overall costs by leveraging the diverse evaluation data collected by various teams. This includes original media, manipulation journal graphs, globally sourced media, and camera training data.



Evaluation Dataset Production Infrastructure

We have developed an infrastructure for generating evaluation datasets that incorporates various types of data, metadata, and ground-truth references, enabling the continuous updating of evaluation data.



Evaluation Data Diversity

How effectively do the algorithms detect different types of manipulations, given the wide range of existing image and video editing methods, techniques, and software?

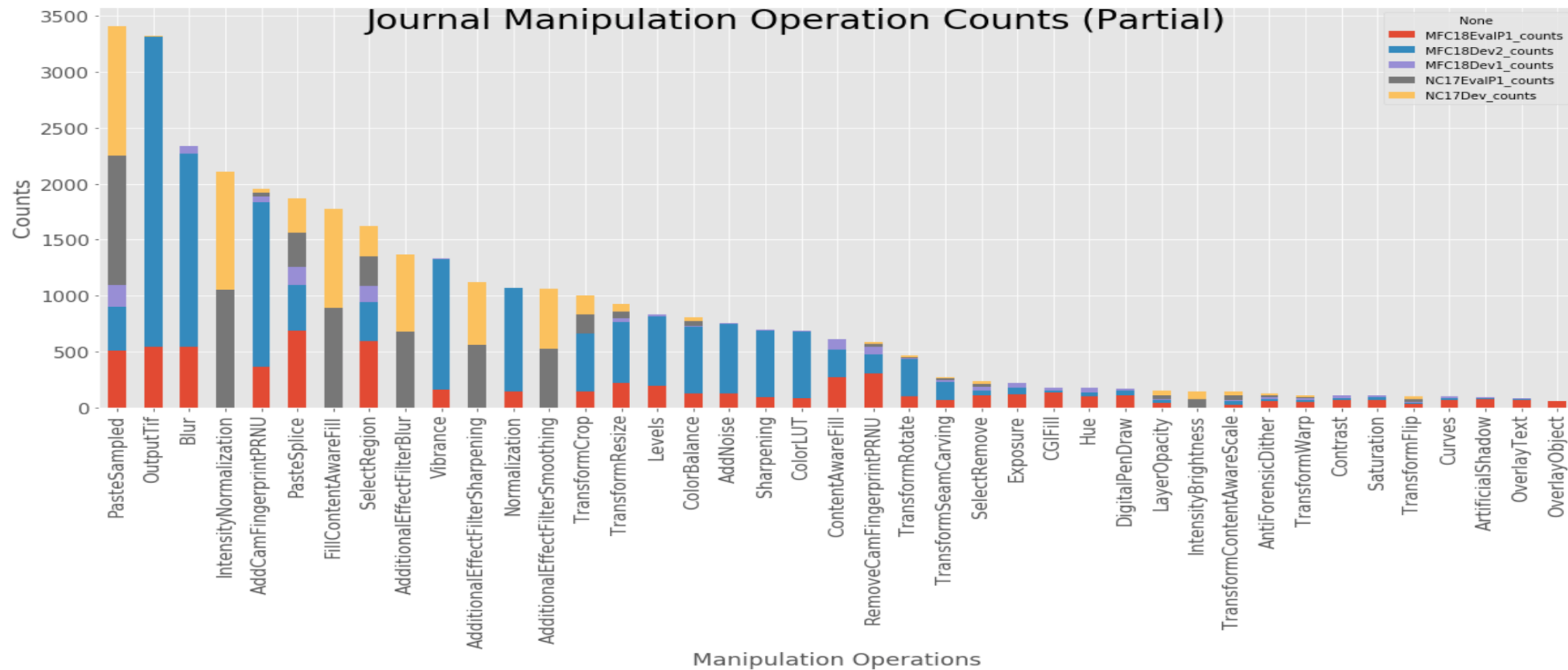
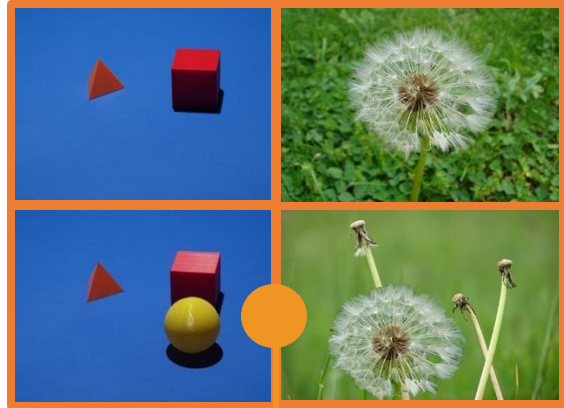
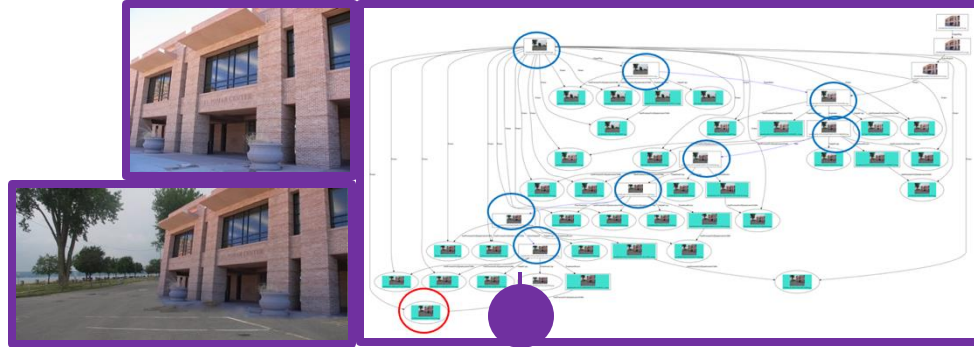


Figure: Histogram of Manipulation Operations Across Datasets

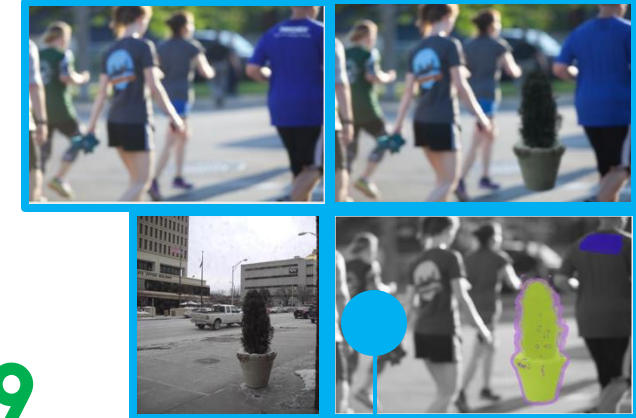
OpenMFC and MediFor Evaluation Datasets (1) NIST



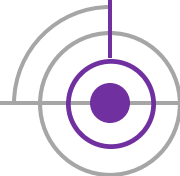
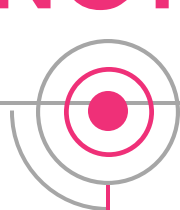
NC16



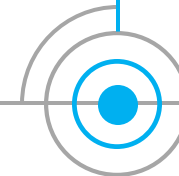
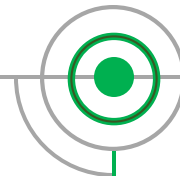
NC17



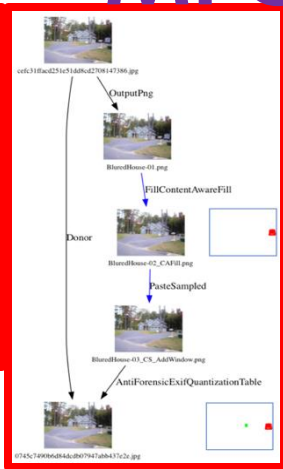
MFC19



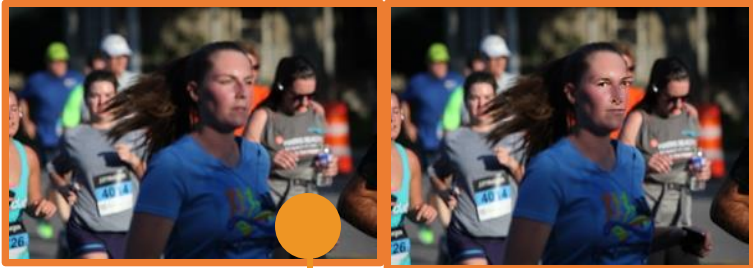
MFC18



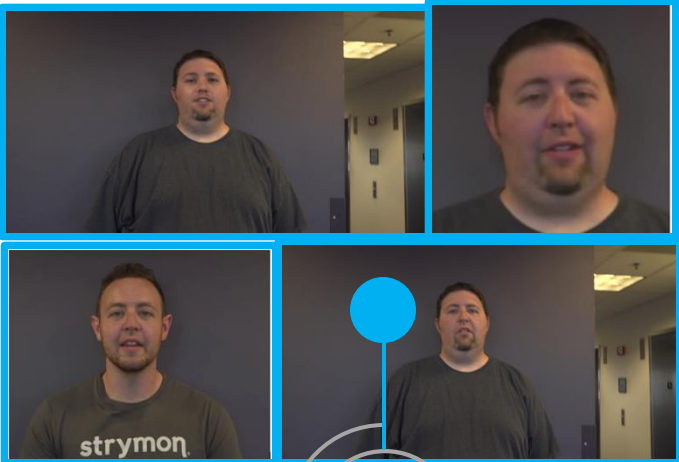
MFC20



OpenMFC and MediFor Evaluation Datasets(2) NLST

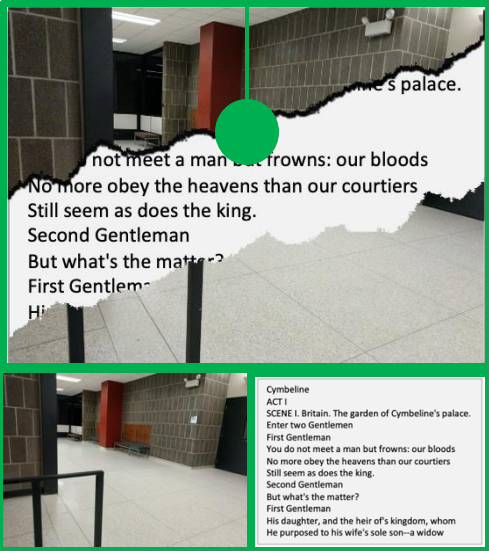


OpenMFC StegD



OpenMFC Image Splice

OpenMFC GAN Image



OpenMFC Deepfake Video

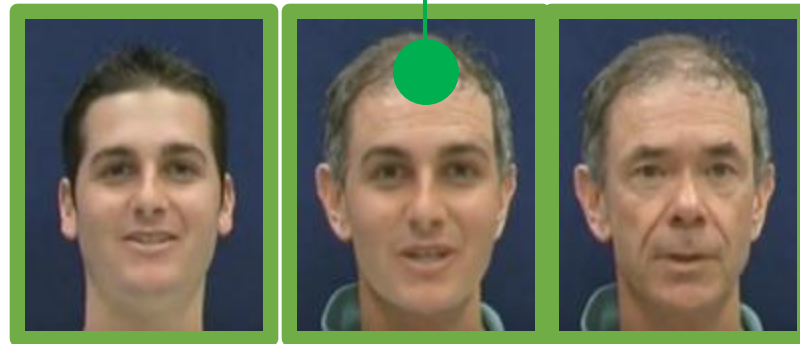
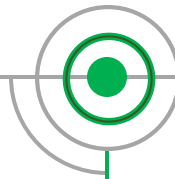


OpenMFC Evaluation Dataset (3) - Coming

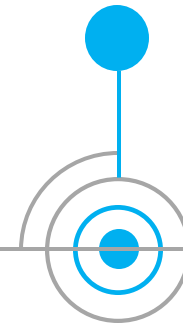


Deepfake Image

Deepfake Video



Deepfake Audio



OpenMFC Evaluation Datasets

Table: OpenMFC2020 Evaluation datasets

OpenMFC 2022 Evaluation Dataset	OpenMFC 2022 Task	Corresponding MFC Dataset Name	Media Type	Media Number	Journal #	Create Date
OpenMFC20_Image_MD	IMD	MFC19 EP1 Image	Image	16K	1383	2022
OpenMFC20_Video_MD	VMD	MFC19 EP1 Video	Video	1.5K	163	2022
OpenMFC20_Image_DD	IDD	MFC18 GAN FULL Image	Image	1.3K	267	2022
OpenMFC20_Video_DD	VDD	MFC18 GAN Video	Video	118	19	2022

New! Table: OpenMFC2022-2023 Evaluation datasets

OpenMFC 2022 Evaluation Dataset	OpenMFC 2022 Task	Media Type	Media Number	Create Date
OpenMFC22_SpliceImage_MD	ISMD	Image	2K	2022
OpenMFC22_Image_StegD	StegD	Image	480	2022
OpenMFC23_Image_DD	IDD	Image	(TBD)	2023
OpenMFC23_Video_DD	VDD	Video	(1.5K)	2023

OpenMFC Evaluation Pipeline

NIST has developed the flexibility to customize the evaluation strategy to meet various assessment requirements, including both take-home and sequestered evaluations, each with its own specific needs and prerequisites.

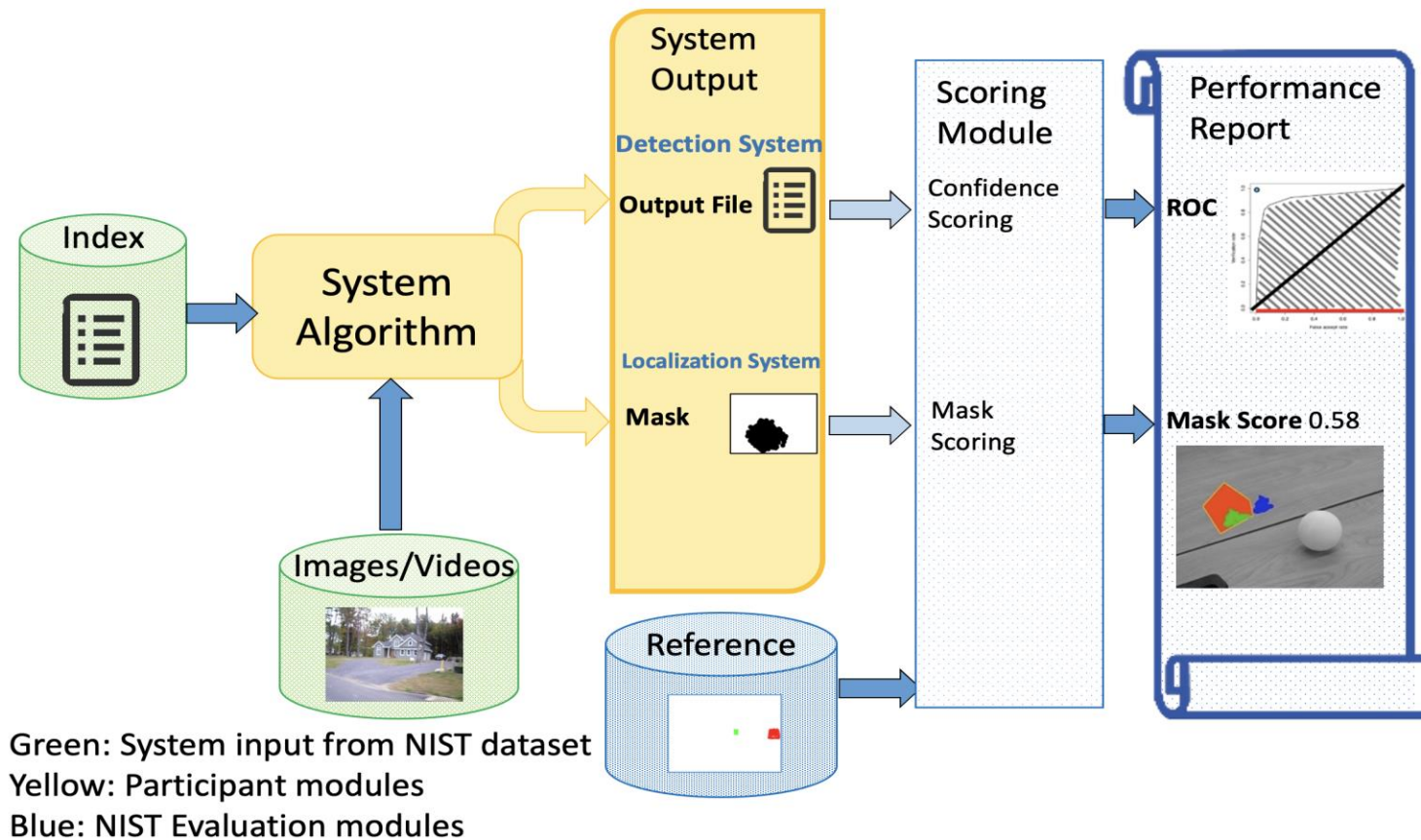


Figure: Evaluation Pipeline

Evaluation Scoring Software: MediScore

<https://github.com/usnistgov/MediScore/>



Unified evaluation framework to support:

- Detection tasks
 - Image/Video Manipulation Detection (Single Input)
 - Splice Detection task (Pair Input)
- Localization tasks
 - Image/Video Manipulation Localization task (Single Input)
 - Splice Localization task (Pair Input)
- Verification tasks
 - Camera ID Verification
 - Event Verification
- Provenance tasks
 - Provenance filter
 - Graph builder

OpenMFC

Major Functions

1. Validator
2. DetectionScorer
3. MaskScorer
4. Provenance
 - ProvenanceValidator
 - ProvenanceGraphBuildingScorer
 - ProvenanceFilteringScorers

Flexible Scoring Options

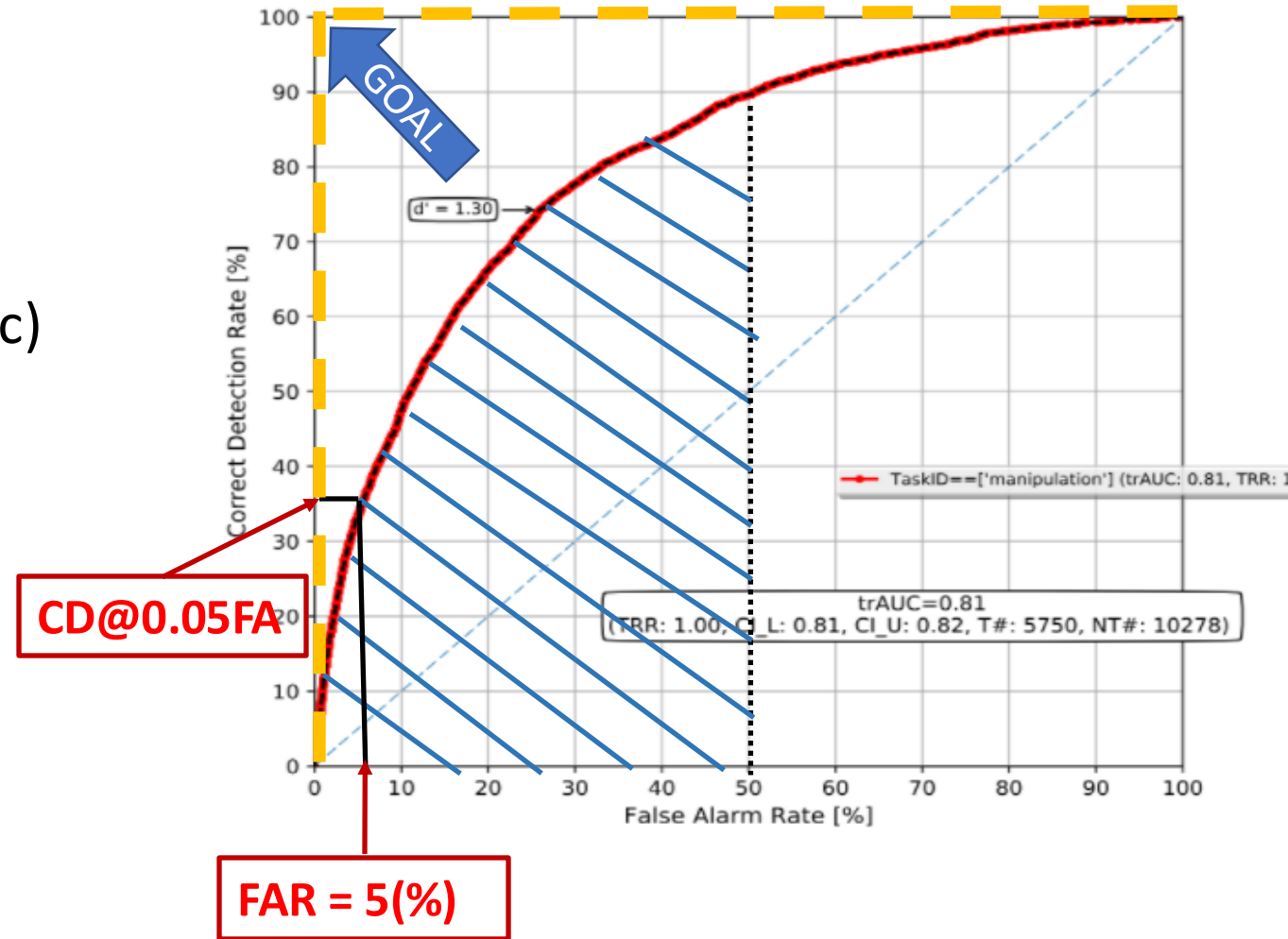
1. OptOut

The testing data is defined by user
2. SelectiveScoring

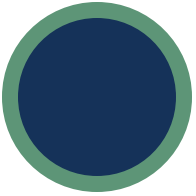
The testing data is defined by selected manipulation operation

Detection System Evaluation Metrics

- Evaluate the accuracy of a system output (e.g., confidence score)
- Evaluation metrics
 - ROC (Receiver Operating Characteristic)
 - AUC (Area Under Curve)
 - CD (Correct Detection) @ FAR = 0.05

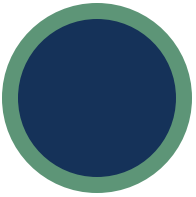


- When reviewing the system performance report, it is important to clearly understand the characteristics of the evaluation data, including its diversity and realism.
- The NIST cross-year evaluation report does not offer a straightforward apples-to-apples comparison due to the increasing complexity of datasets over time.
- The performance of media forensic systems may decline due to the growing diversity and complexity of datasets, including factors such as manipulation complexity, the introduction of new manipulations, and the emergence of anti-forensic technologies.
- The effectiveness of media forensic systems can be influenced by factors such as resizing, compression, post-processing, and the use of anti-forensic tools.



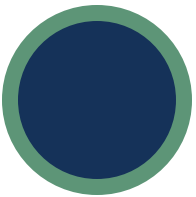
Dynamic Update

The landscape of media forensic research is continuously evolving with the advancement of media editing and generation technologies.



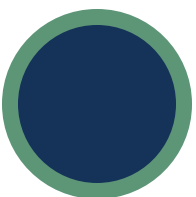
Media Forensics vs. Anti-Forensics

The performance of forensic systems may decline as anti-forensic technologies adapt and improve.



Data Diversity

A wide range of diversity in real applications.



Forensic System Performance

There is still considerable room for improvement in media forensic technologies.

Challenges



Continuous Evaluation

The data generation infrastructure supports regularly update.



Red team vs. Blue team evaluation



Collaboration Platform

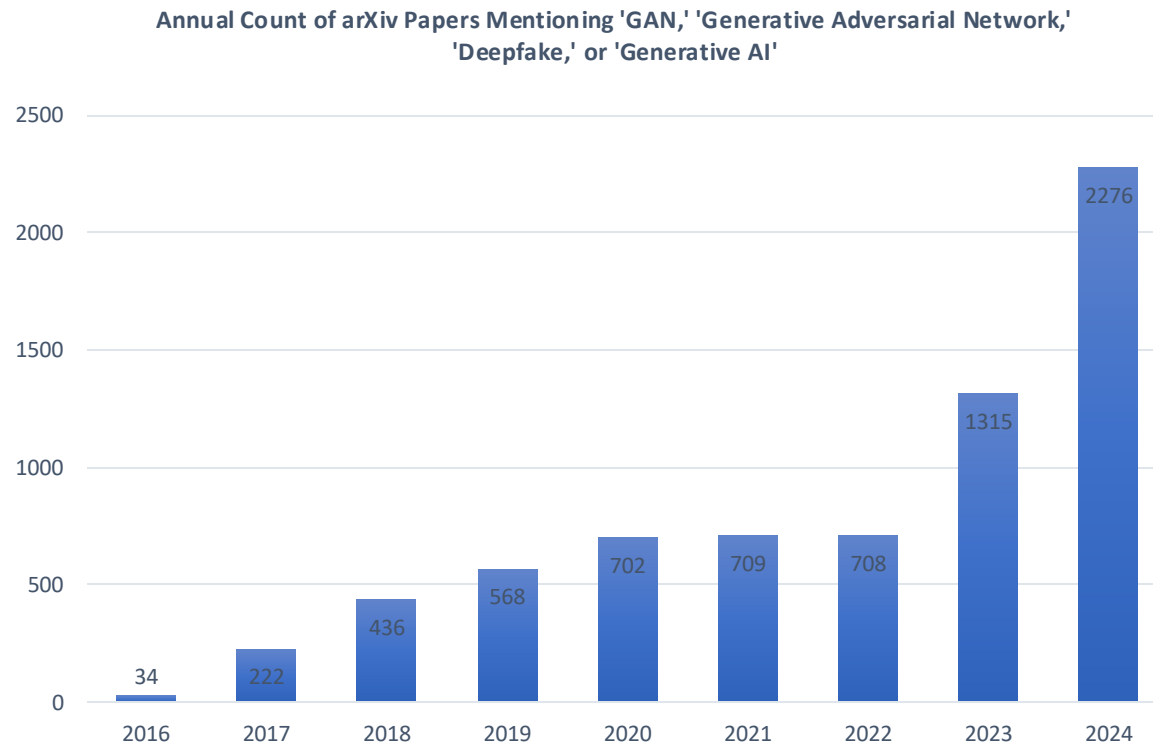
Extensive experience in dataset design, collection, annotation, and generation



Flexible and Customized Evaluation Strategies

NIST Solutions

Ongoing Efforts: Deepfake Challenge



Annual Count of arXiv Papers Mentioning 'GAN,' 'Generative Adversarial Network,' 'Deepfake,' or 'Generative AI' as 12/20/2024

Generative AI models:

- GAN (Generative Adversarial Networks) (2014)
- Text-to-Image Synthesis (2015)
- Image-to-Image Generation (2017)
- Deepfakes (2018)
- Diffusion Models (2021)
- Text-to-Audio and Voice Cloning Models (2023)
- Multimodal Models (2024)

Focus:

- Media Types: Image, Video, and Audio
- Content Types: Face, nonface,

AI Media Generation Tools

- DeepFaceLab, StyleGAN 1/2/3, Celeb-DF
- Post-processing or Anti-forensics

Data Release IRB approval

Collaborators: Prof. Conrad Sanderson (VidTIMIT); DARPA MediFor data teams from U.C. Denver and PAR Government

Capable for Special Studies

- Enhanced with additional meta-data collection

Support for Forensic System Development

Collaborators: Prof. Siwei Lyu, Dr. Shan Jia, and Yan Jun at University at Buffalo

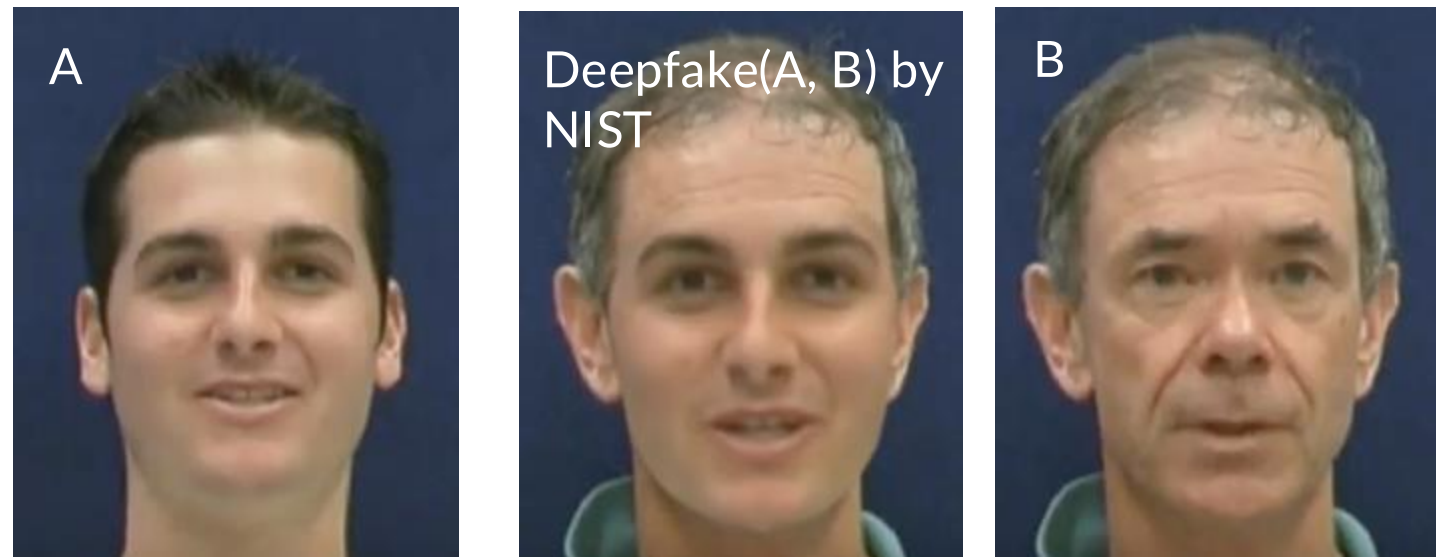


Figure: Example of a Deepfake and Its Two Source Videos



Text-to-Image

DALL·E 2 (OpenAI), Midjourney, Runway, Adobe Firefly, Picsart, Craiyon, Jasper, Canva, PicSo, Bing Image Creator



Image-to-Image

RunwayML, Media.io, AI Avatar, Midjourney (Discord), Stable diffusion

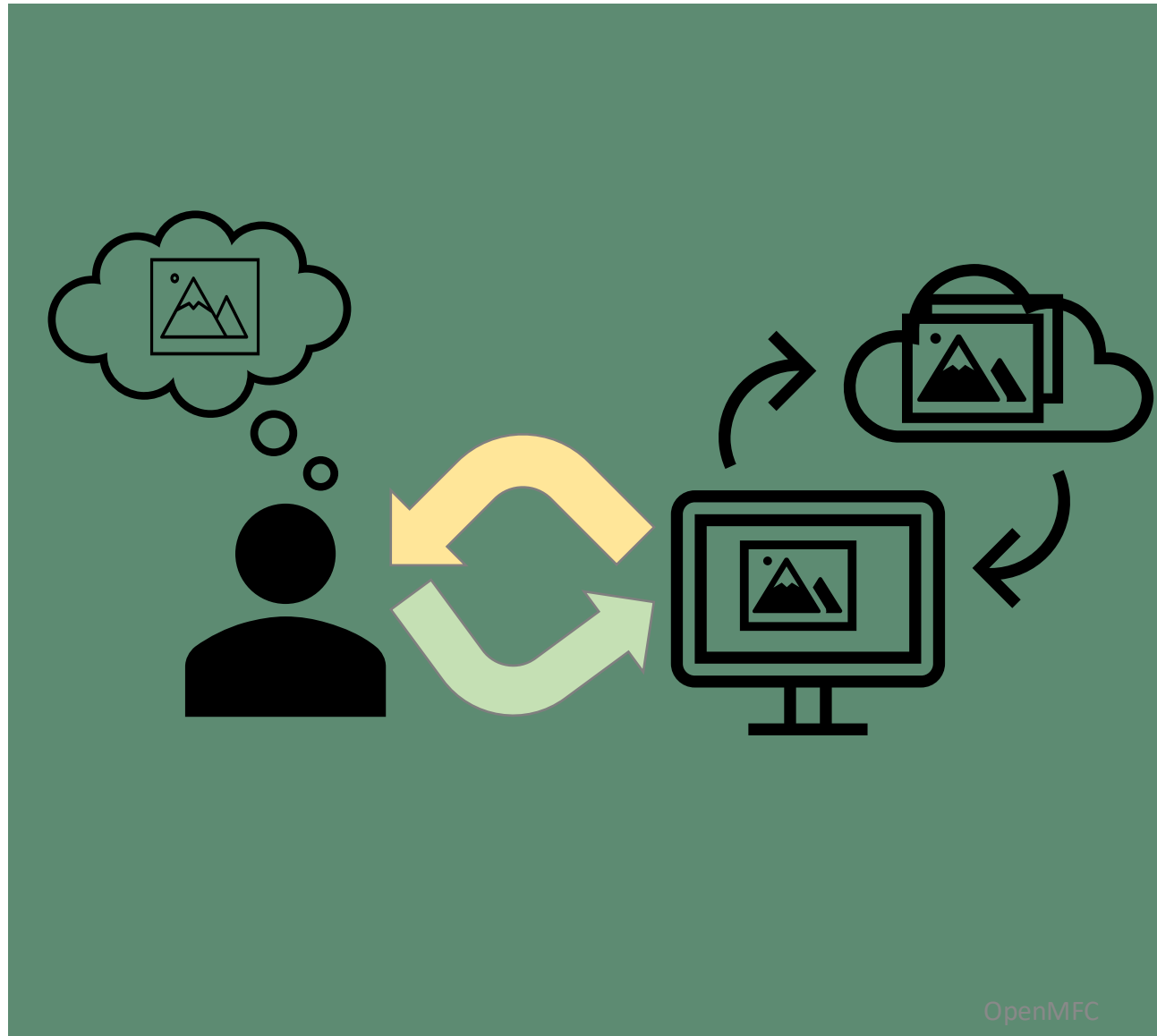


Beyond Forensics

Art and content creation, image inpainting, video synthesis, character animation, super-resolution, etc.

Future Direction: New Evaluation Methodology

Bridging the Gap Between Traditional AI and HMI Assessment Paradigms



- **Objective Evaluation**
Traditional AI evaluation methods for detection, verification, retrieval systems
 - Is it real or generated?
 - What is the provenance?
- **Subjective Evaluation**
Human as an evaluator
 - Creativity
 - User satisfaction
 - User interaction experiences
- **Human-Machine Interaction (HMI)**
Humans as a users
HMI evaluation methods
 - Usability and design
 - Overall performance and effectiveness

How to organize an evaluation program?

1 Artists - Media Creation and Manipulation

Data collection

Data manipulation

Data annotation

Data creation

3 Anti-Forensic Experts

Post-processing algorithms

Recapture

Anti-Forensic systems

2 Forensic Experts

Media forensics specialists

Forensic practitioners

4 AI and HMI Experts

Generator model

Detector model

Forensic model

Human-Computer-Interaction

- Challenges

- Data
 - expert, time, cost, labor, IRB agreements
- Systems
 - forensic baseline systems
 - anti-forensic systems
- Participants
 - high demands on time and efforts

- Collaboration Platform

- Cross-team collaboration: share knowledge, skill, tools, and resources
- Intra-team collaboration: foster cooperation within teams rather than competition

NIST evaluations

- Maintain a neutral and unbiased stance
- Provide ongoing assessments of emerging technologies
- Support a diverse range of stakeholders, including government policymakers, industry product developers, and academic researchers

NIST OpenMFC at a Glance



800+¹
people



200+
organizations



26+
countries



4 + 2
evaluations



15+
publications



5+
collaborative
organizations

¹ Over the years of MFC and OpenMFC, more than 800 individuals have been involved as organizers, program developers, participants, stakeholders, government agency representatives, registered users of the evaluation website, or recipients of our development and evaluation datasets, among other roles.

Publications (1)

Yan Ju, Shan Jia, Jialing Cai, Haiying Guan, Siwei Lyu, "GLFF: Global and Local Feature Fusion for AI-synthesized Image Detection," IEEE Transactions on Multimedia, 2023, <https://ieeexplore.ieee.org/document/10246417>, Digital Object Identifier: 10.1109/TMM.2023.3313503.

Haiying Guan, Yooyoung Lee, Lukas Diduch, Jesse Zhang, Ilia Ghorbanian, Timothee Kheyrkhah, Peter Fontana, Jonathan Fiscus and James J Filliben, "Open Media Forensics Challenge (OpenMFC) 2020-2021: Past, Present, and Future," NIST Interagency/Internal Report (NISTIR) Number 8396, <https://doi.org/10.6028/NIST.IR.8396>.

Haiying Guan, Andrew Delgado, Yooyoung Lee, Amy N. Yates, Daniel Zhou, Timothee Kheyrkhah, and Jonathan Fiscus, "User Guide for NIST Media Forensic Challenge (MFC) Datasets," NIST Interagency/Internal Report (NISTIR) Number 8377, July 2021, <https://doi.org/10.6028/NIST.IR.8377>.

Haiying Guan, Mark Kozak, Eric Robertson, Yooyoung Lee, Amy N. Yates, Andrew Delgado, Daniel Zhou, Timothee Kheyrkhah, Jeff Smith and Jonathan Fiscus, "MFC Datasets: Large-Scale Benchmark Datasets for Media Forensic Challenge Evaluation," 2019 IEEE Winter Applications of Computer Vision Workshops (WACVW), pp. 63-72, 2019, <https://ieeexplore.ieee.org/document/8638296>.

Eric Robertson, Haiying Guan, Mark Kozak, Yooyoung Lee, Amy N. Yates, Andrew Delgado, Daniel Zhou, Timothee Kheyrkhah, Jeff Smith and Jonathan Fiscus, "Manipulation Data Collection and Annotation Tool for Media Forensics," IEEE Computer Vision and Pattern Recognition conference 2019, Workshop on Applications of Computer Vision and Pattern Recognition to Media Forensics, 2019. https://openaccess.thecvf.com/content_CVPRW_2019/html/Media_Forensics/Robertson_Manipulation_Data_Collection_and_Annotation_Tool_for_Media_Forensics_CVPRW_2019_paper.html.

Publications (2)

Haiying Guan, Baptiste Chocot, Iliia G. Bajgiran, Lukas Diduch, Yooyoung Lee, Siwei Lyu, Yan Ju, Shan Jia, Conrad Sanderson
"Deepfake Challenge 2023 Video Dataset ReadMe," NIST Publication ID 936550, ERB Control#: N2023-0762.

Haiying Guan, Yooyoung Lee, and Lukas Diduch, "Open Media Forensics Challenge 2022 Evaluation Plan,"
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NIST OpenMFC Resource

Overview

The Open Media Forensics Challenge (OpenMFC) is a media forensics evaluation to facilitate development of systems that can automatically detect and locate manipulations in imagery (i.e., images and videos).

What

The NIST OpenMFC evaluation is being conducted to examine the performance of system's accuracy and robustness over diverse datasets collected under controlled environments.

Who

The NIST OpenMFC is open worldwide. We invite all organizations including past DARPA MediFor Program participants to submit their results using their technologies to the OpenMFC evaluation server. Participation is free. NIST does not provide funds to participants.

How

To take part in the OpenMFC evaluation you need to register on this website and complete the data license to download the data. Once your system is functional you will be able to upload your outputs

News

- 23 JAN, 23 OpenMFC2022 Workshop talks and slides are uploaded!
- 6-7 DEC, 22 OpenMFC2022 Workshop
- 01 AUG., 22 Stego data available
- 26 JULY, 22 New OpenMFC data download webpage
- 03 MAR, 22 OpenMFC2022 Eval Plan
- 15 FEB, 22 OpenMFC2021 Workshop Talks and Slides Uploaded!
- 7-9 DEC, 21 OpenMFC2021 Workshop

Website

<https://mfc.nist.gov>

Slack

<http://openmfc.slack.com>

Scoring Software

<https://github.com/usnistgov/MediScore/>

Contact

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Acknowledgement

- External collaborators:
 - Prof. Siwei Lyu, Dr. Shan Jia, and Yan Ju at University at Buffalo
 - Prof. Conrad Sanderson
 - DARPA MediFor Program
 - Data team from National Center for Media Forensics, U.C. Denver
 - Data team from PAR Government
 - Prof. Jennifer Newman, Li Lin, Prof. Yong Guan from Iowa State University, and Prof. Roy A. Maxion from Carnegie Mellon University
- NIST contributors
 - James Horan
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Questions?

Stay In Touch!

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