

Decentralized Video Verification System Using Blockchain to Detect and Prevent Deepfake Media

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Abstract

The increasing prevalence of AI-generated deepfake videos has raised significant concerns regarding the trustworthiness of digital media. This project presents a decentralized platform that combines artificial intelligence with blockchain technology to detect, prevent, and trace deepfake media. By analyzing videos and images using perceptual hashing, histogram-based features, and deep learning models, the system compares uploads against previously stored media on the blockchain. Metadata is stored securely using smart contracts, and permission-based reuse is enforced. This hybrid approach ensures real-time verification, accountability, and transparency.

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1 Introduction

Deepfakes are synthetic media in which a person in an existing image or video is replaced with someone else's likeness using artificial intelligence. Their increasing quality and accessibility have caused serious concerns in misinformation, political sabotage, and digital fraud. Current detection methods are largely centralized and reactive. Our system introduces a decentralized, proactive mechanism to detect and block deepfake media during the upload process itself.

2 Problem Statement

There is a lack of effective systems that verify the authenticity of videos in real time at the point of upload. Additionally, centralized databases are vulnerable to manipulation. We aim to solve this using blockchain for immutability and AI for intelligent detection.

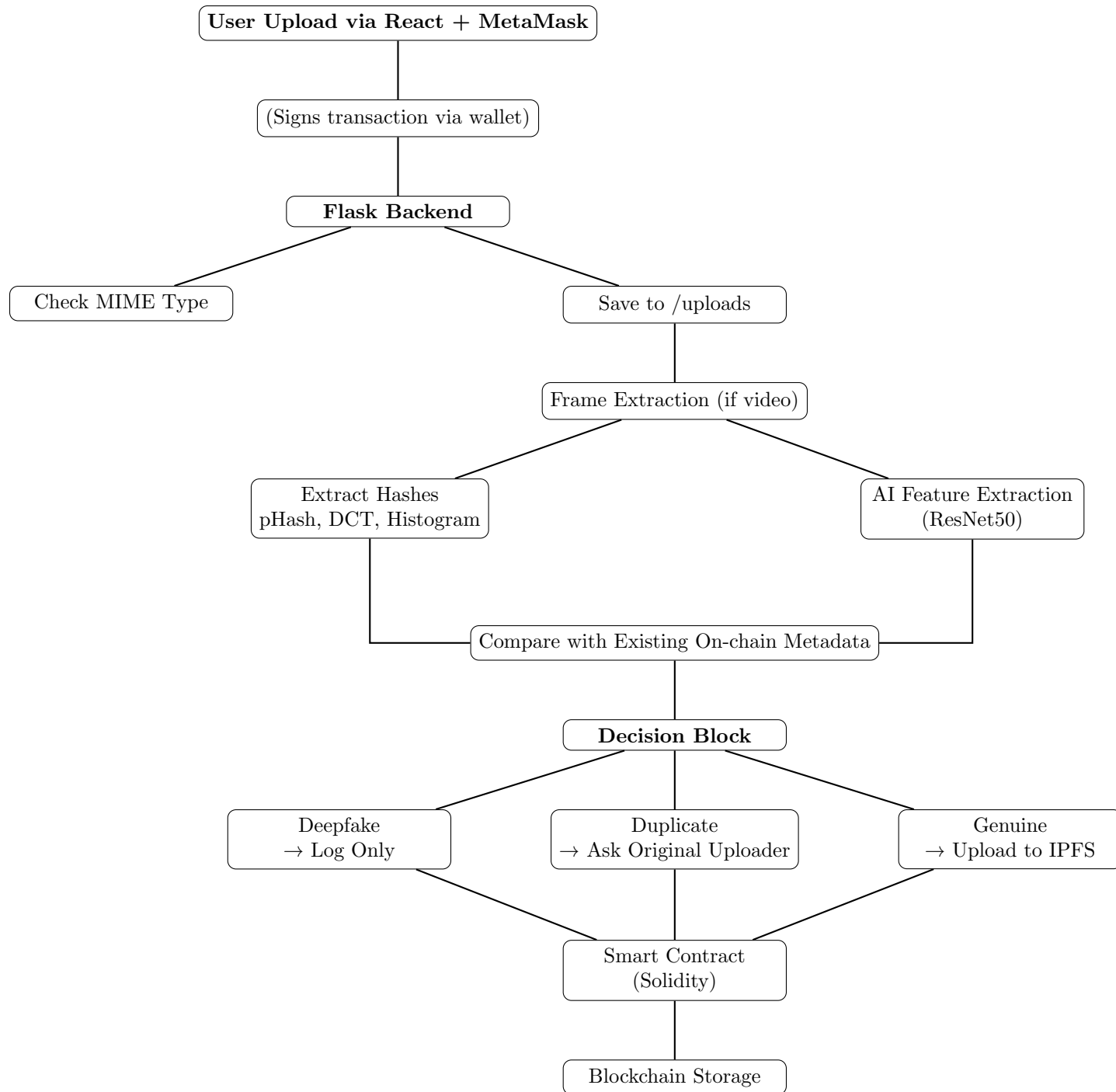
3 Proposed System

Our solution combines:

- React frontend with MetaMask for decentralized authentication
- Python Flask backend for media processing
- Smart contracts on Ethereum for recording metadata and enforcing permission-based reuse
- Perceptual hashing, DCT, Histogram, and AI features for similarity detection

4 System Architecture

4.1 System Design Flow



5 Backend Architecture

5.1 Technologies Used

- Flask
- OpenCV for frame extraction

- ImageHash for perceptual hashing
- PyTorch for AI feature extraction
- SQLite and JSON for local storage and interim data

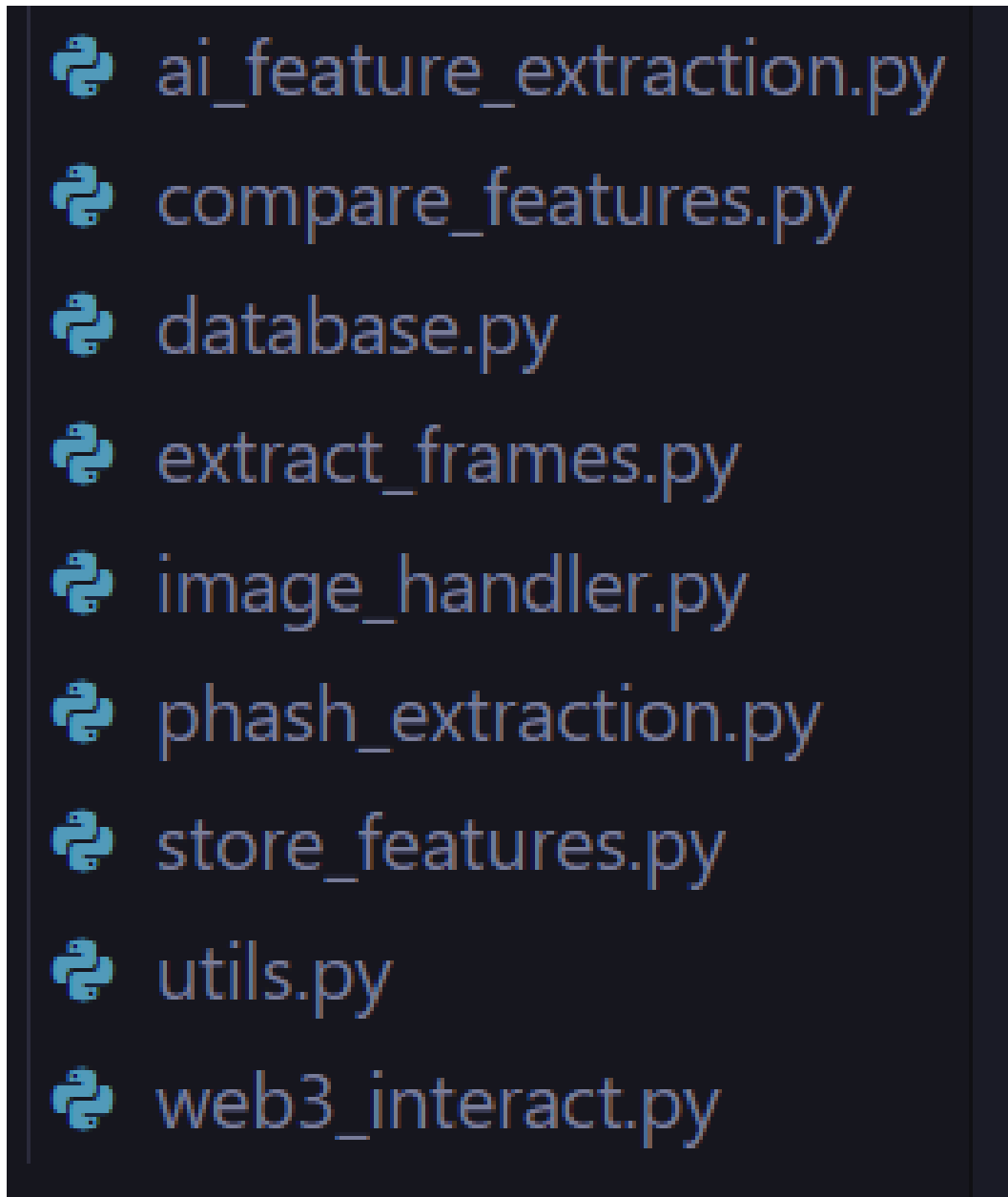


Figure 1: Flask backend logging the analysis pipeline

A screenshot of a code editor with a dark background. It displays a JSON object representing the contents of a file named 'latest_upload.json'. The JSON is formatted with syntax highlighting: strings are in green, numbers in orange, and null in red. The structure includes fields for filename, three lists of hashes (phash_list, dct_list, hist_list), an AI feature path, a duplicate flag, and a best_match object containing various percentage values and a matched video status.

```
[{"filename": "icons8-medium-50.png", "phash_list": ["0000000000000000"], "dct_list": ["0000000000000000"], "hist_list": ["0000000000000000"], "ai_feature_path": "static/features/icons8-medium-50.png.pkl", "duplicate_flag": 1, "best_match": {"matched_video": null, "phash_percent": 0, "dct_percent": 0, "hist_percent": 0, "ai_match_percent": 0}}
```

Figure 2: Sample contents of latest_upload.json

5.2 Flow

1. File uploaded via frontend
2. Flask identifies MIME type
3. If video → extract frames
4. Run pHash, DCT hash, Histogram
5. Run ResNet to extract AI features
6. Compare hashes and features with existing records
7. Save summary to latest_upload.json

6 Smart Contract Design

6.1 Responsibilities

- Upload metadata for genuine videos
- Log deepfake attempts (with penalty)

- Allow permission request workflow
- Track user's genuine score (+1/-10)

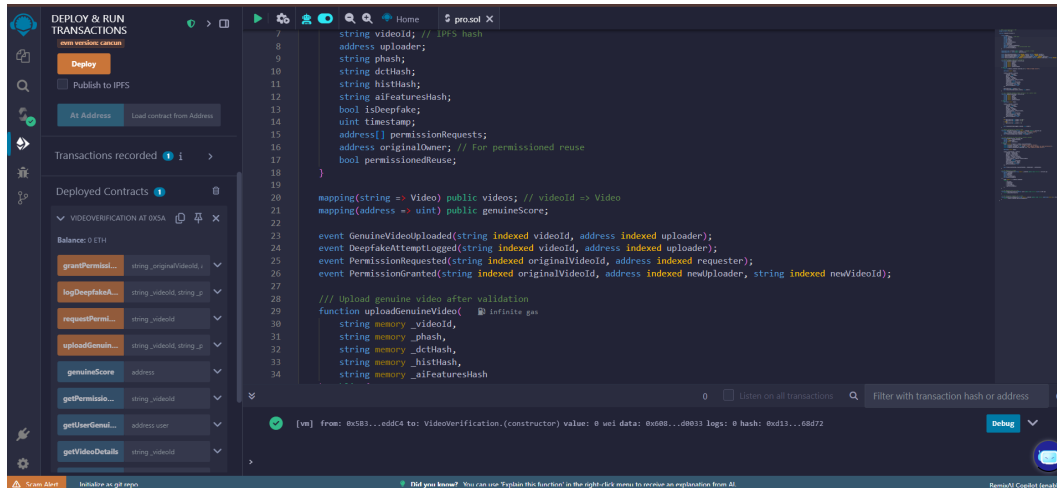


Figure 3: Deploying the smart contract using Remix

6.2 Core Methods

- `uploadGenuineVideo()`
- `logDeepfakeAttempt()`
- `requestPermission()`
- `grantPermission()`
- `getUserGenuineScore()`

7 Frontend Interface

- Built in ReactJS
- Wallet connect via MetaMask
- Upload panel (image/video)
- Output feedback
- Feed of previously uploaded content

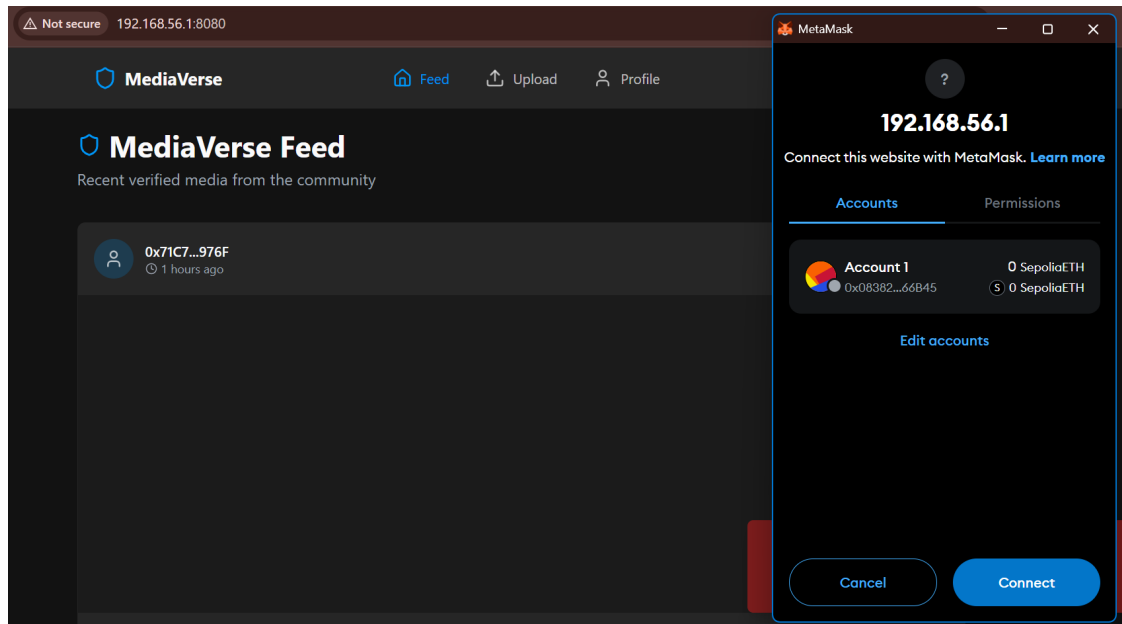


Figure 4: React frontend with MetaMask-connected video upload panel

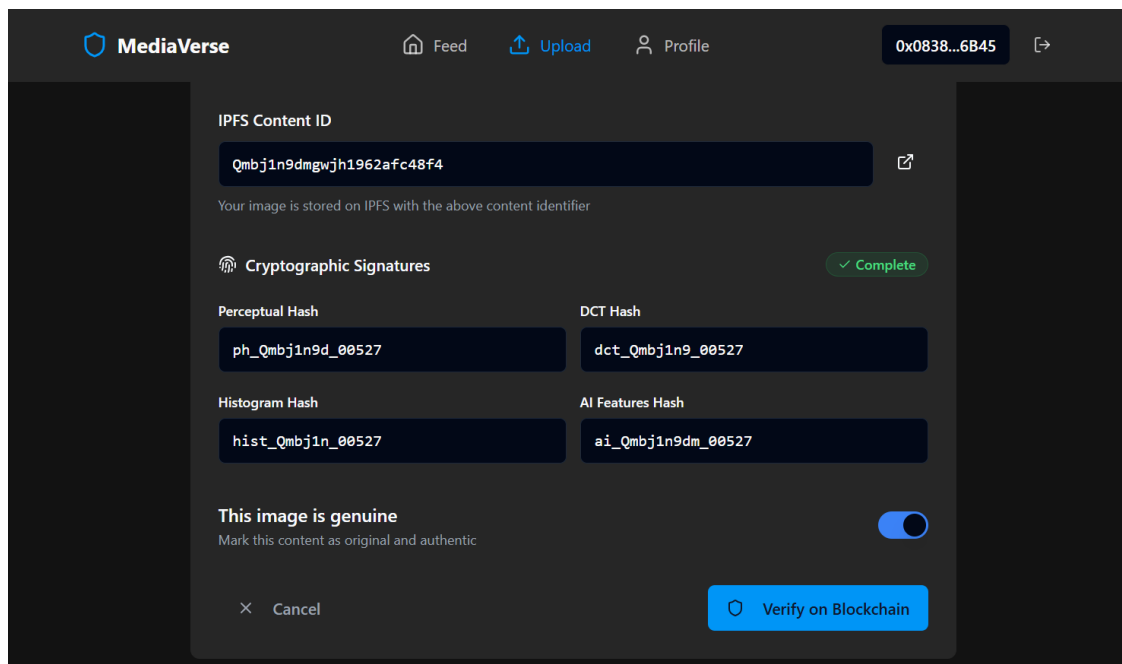


Figure 5: React frontend when uploading the media

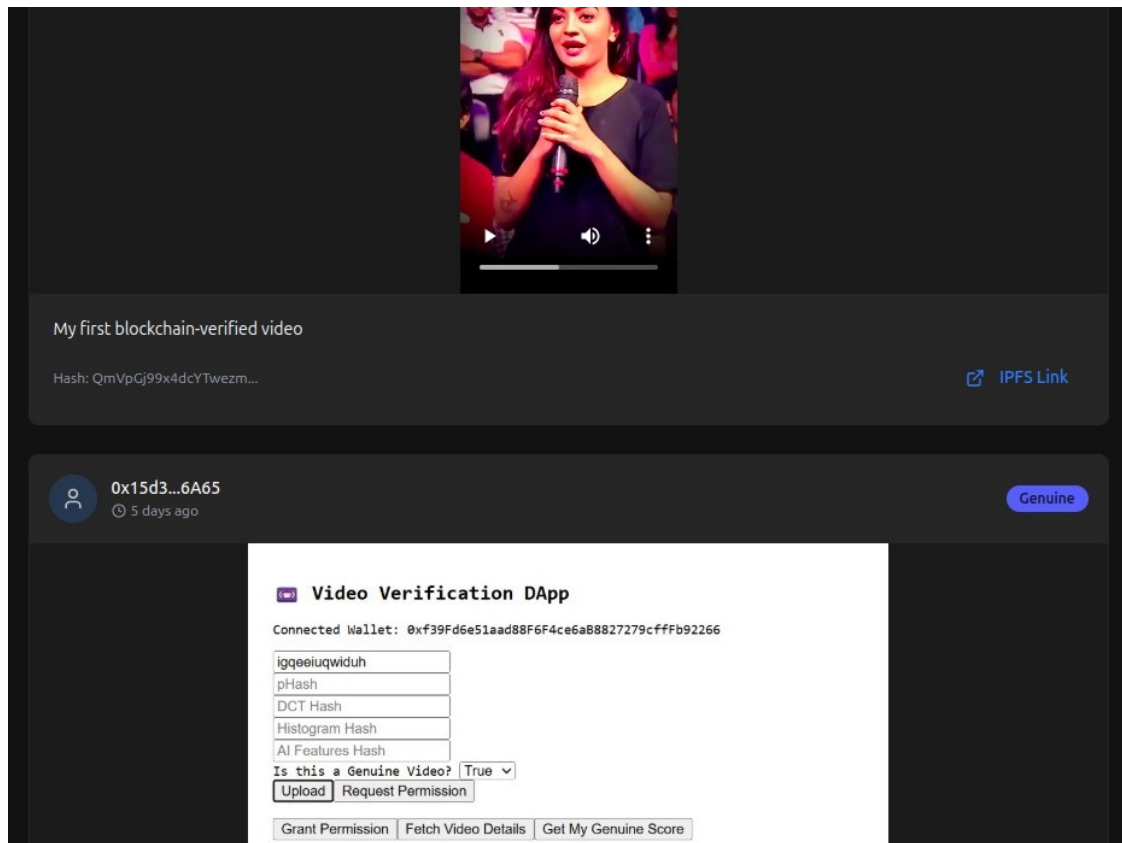


Figure 6: Feed showing previously uploaded verified videos

8 JSON Data Format

latest_upload.json example:

```
{
  "filename": "cat.mp4",
  "phash_list": [...],
  "dct_list": [...],
  "hist_list": [...],
  "ai_feature_path": "static/features/cat.pkl",
  "duplicate_flag": 1,
  "best_match": {
    "matched_video": "lion.mp4",
    "phash_percent": 91.3,
    "dct_percent": 89.4,
    "hist_percent": 87.2,
    "ai_match_percent": 93.1
  }
}
```

9 Edge Cases Handled Effectively

Our system is not only capable of identifying exact matches or fully synthetic videos but is also robust against many real-world edge cases. Here's how:

1. Exact Reuploads and Minor Modifications

Scenario: A user uploads the same video with slight edits (cropping, re-encoding, noise).

Detection:

- **pHash:** Matches almost exactly.
- **DCT Hash:** Captures compression artifacts.
- **ResNet50 features:** Remain similar.

Action: Video is flagged as duplicate and blocked.

2. Partial Deepfakes in Long Videos

Scenario: Only a few seconds of deepfake in an otherwise real video.

Detection:

- **Frame-wise hashing** highlights altered frames.
- **Deep features** flag semantic inconsistencies.

Action: Flag only deepfake frames, alert uploader.

3. Color-Based Manipulation (e.g., Skin Tone Changes)

Scenario: Altering color spectrum to bypass visual detection.

Detection:

- **Histogram Hash:** Sensitive to color tone changes.
- **DCT Hash:** Detects low-frequency distortions.

Action: Flag the video for review if histogram anomaly is high.

4. Same Scene Filmed by Different Users

Scenario: Multiple users record the same event from different angles.

Avoiding False Positives:

- **pHash:** May match partially.
- **DCT + AI features:** Differ clearly.

Action: Not flagged as duplicate unless at least two hash types match.

5. Deepfake Tools That Modify Only Faces

Scenario: Faces are manipulated, but backgrounds remain identical.

Detection:

- **DCT Hash:** Detects facial distortions.
- **ResNet50:** Highlights semantic facial changes.

Action: Flagged as potential deepfake.

6. Video Compression as a Bypass Method

Scenario: A genuine video is compressed and re-uploaded.

Detection:

- **pHash + DCT Hash:** Minor differences, still highly similar.

Action: Allowed if variation is minor, not flagged as deepfake.

7. Video Splicing (Combining Multiple Sources)

Scenario: Real and deepfake segments combined in one file.

Detection:

- **Frame-wise Hashing:** Detects inconsistent segments.
- **Histogram:** Highlights visual mismatches.

Action: Only fake segments are flagged; user notified.

10 Key Innovations

- Multi-modal verification before publishing media
- Local + blockchain storage combo for speed and trust
- Reusability and ownership governance using smart contract

11 Scope for Expansion

- Integrate IPFS for decentralized media storage
- Replace SQLite with decentralized DB
- Add live AI-based deepfake classifier (GAN detection)
- Token-based reward/penalty system

12 Conclusion

This project bridges the gap between blockchain and AI for responsible media verification. It lays a foundation for scalable, tamper-proof systems capable of halting deepfake spread at the source.

References

- FaceForensics++ Dataset (Rössler et al.)
- Solidity Documentation – Ethereum.org
- PyTorch ResNet Models
- MetaMask Web3 Integration Guide