

LLM-Augmented Audio-Visual Lip-Sync Deepfake Detection

Pilot Notebook (Design & Planning)

Step 1: Pilot Dataset Selection

- Goal: Assemble a pilot list of 20–50 clips (train/val).
 - Deliverable: Table with dataset names, real/fake counts, purpose.
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Step 2: Preprocessing Policy

- Face detector choice
 - Lip ROI size
 - FPS standardization
 - Crop stability rules
-

Step 3: Feature Plan

- SyncNet scores
 - AV-HuBERT embeddings
 - Whisper transcripts
 - MFA alignment outputs
-

Step 4: LLM Scoring Specification

- Prompt templates
 - Output fields
 - Decision rubric
-

Step 5: Fusion & Evaluation Metrics

- Feature fusion approach
 - Metrics: AUC, EER
 - IoU for temporal localization
-

Step 6: Error Taxonomy

- Desync types
 - Dubbing artifacts
 - Rhythm anomalies
-

Step 7: Privacy & Ethics Note

- Face/voice handling
 - Storage policy
 - Sharing constraints
-

Step 8: Minimal Demo Checklist

- What constitutes a successful first pipeline run
- Evidence to capture

Pilot Dataset Selection (Step 1)

Goal: Assemble a pilot list of 20–50 clips (train/val) covering all datasets.

Dataset	Real Clips	Fake Clips	Total	Purpose (train/val)
FakeAVCeleb	5	5	10	Training
AV-Deepfake1M	5	5	10	Validation
LAV-DF	3	3	6	Training (temporal loc)
AVLips	3	3	6	Stress-test
LRS2 / LRS3	4	–	4	Pretraining/finetuning (real only)
Total	20	16	36	---

Notes:

- Core training/validation clips: FakeAVCeleb & AV-Deepfake1M
- Temporal and stress-test clips: LAV-DF & AVLips
- Real-only clips for VSR pretraining: LRS2 / LRS3
- This pilot set is small enough to test the pipeline quickly but representative of all datasets.

Total pilot clips = 36, which is within the recommended 20–50 range for quick testing.

Preprocessing Policy (Step 2)

Goal: Prepare video clips for consistent feature extraction.

1. Face detector:

- Use **RetinaFace** for accurate face and landmark detection.
- Alternative for pilot: OpenCV Haar cascade for fast checks.

2. Lip ROI size:

- Crop mouth region including small margin.
- Resize to **128x128 pixels** for model input.
- Maintain aspect ratio if possible.

3. FPS standardization:

- Convert all clips to **25 FPS**.
- Ensures consistent temporal resolution across datasets.

4. Crop stability rules:

- Center crop on lips using landmarks for each frame.
- Smooth bounding box positions across frames to avoid jitter.
- Skip frames with occluded faces or very small bounding boxes.

Consistent preprocessing ensures reliable feature extraction across all clips.

Feature Plan (Step 3)

Goal: Define features to extract from each pilot clip.

1. SyncNet Scores (Audio-Visual Sync):

- Compute per-frame sync score between audio and lip movements.
- Helps detect misaligned lip-sync forgeries.

2. AV-HuBERT Embeddings / Lip-Reading Features:

- Extract visual speech embeddings from cropped lip region.
- Frame-level embeddings for temporal modeling.
- Optional: generate lip-reading transcripts.

3. Whisper Transcripts (Audio ASR):

- Convert audio to text using Whisper.
- Provides textual content for LLM scoring and semantic comparison.

4. Montreal Forced Aligner (MFA) Alignment Outputs:

- Align phonemes with audio timestamps.
- Compare phoneme timing with visual visemes to detect desync.

Notes:

- These features will be **combined in later stages** for classifier input.
- SyncNet + AV-HuBERT → core A/V features
- Whisper + MFA → textual/phoneme timing features

All features will be saved as numpy arrays (.npy) for easy loading into the classifier.

LLM Scoring Specification (Step 4)

Goal: Use LLM to evaluate semantic and prosody consistency.

1. Prompt Templates:

- Ask LLM to rate plausibility of spoken content based on Whisper transcripts and optional AV-HuBERT lip-reading transcripts.
- Example prompt:

Given the transcript and lip-reading content, rate how plausible the speech
Transcript: <Whisper transcript>
Lip-reading text (optional): <AV-HuBERT transcript>
Rate consistency from 0 (fake) to 1 (real).

2. Output Fields:

- **Score:** numeric value 0–1
- **Optional reasoning text** for explainability

3. Decision Rubric:

- Score > 0.7 → likely real
- Score < 0.3 → likely fake
- Score 0.3–0.7 → uncertain, combine with SyncNet/AV-HuBERT for final decision

Example: Score = 0.8 → likely real; Score = 0.2 → likely fake.

Fusion & Evaluation Metrics (Step 5)

Goal: Combine features and define performance metrics.

1. Feature Fusion Approach:

- Concatenate SyncNet scores, AV-HuBERT embeddings, phoneme/text alignment features, and LLM score.
- Feed combined feature vector into a light classifier (e.g., XGBoost or LightGBM).

- Alternative: weighted combination or late fusion of separate predictions.

2. Evaluation Metrics:

◦ Video-level:

- AUC (Area Under ROC Curve)
- EER (Equal Error Rate)

◦ Segment/frame-level (temporal localization):

- IoU (Intersection over Union) between predicted fake segments and ground truth

Notes:

- Video-level metrics check overall detection performance.
- IoU is used only on datasets with **per-frame ground truth** (like LAV-DF and AVLips).
- Fusion strategy ensures all features contribute to final real/fake decision.

Segment-level IoU will be calculated for datasets with per-frame annotations (LAV-DF, AVLips).

Error Taxonomy (Step 6)

Goal: Categorize types of errors or anomalies in lip-sync deepfakes.

1. Desynchronization Errors:

- Lip movements do not match audio (delays, misaligned phonemes/visemes).

2. Dubbing / Overlay Artifacts:

- Audio replaced or edited independently (background noise mismatch, re-recorded speech).

3. Rhythm / Prosody Anomalies:

- Timing, pitch, or intonation inconsistent with natural speech.

4. Visual Artifacts (Optional for Pilot):

- Blurry or jittery lips, occlusions (hand, microphone, hair), sudden head movement.

Notes:

- This taxonomy will guide **analysis of model failures**.
- Helps in refining preprocessing, features, or LLM prompts if errors are frequent.

Error Type	Example Description
Desynchronization	Phonemes not aligned with visemes
Dubbing Artifact	Re-recorded speech over original video
Rhythm / Prosody	Flat or unnatural timing

Privacy & Ethics Note (Step 7)

Goal: Ensure responsible handling of face and voice data.

1. Face / Voice Handling:

- Use data only for research and model development.
- Avoid storing unnecessary personal information.
- Crop faces/lips to minimize identity exposure.

2. Storage Policy:

- Store videos, cropped faces, and features in secure, restricted drives.
- Limit access to team members.
- Encrypt or anonymize data where possible.

3. Sharing Constraints:

- Do not publicly share raw videos with real people.
- Share only derived features (embeddings, SyncNet scores) if needed.
- Comply with dataset licenses (FakeAVCeleb, LAV-DF, etc.).

All datasets (FakeAVCeleb, LAV-DF, etc.) will be used in accordance with their respective licenses.

Minimal Demo Checklist (Step 8)

Goal: Define success criteria for the first pilot pipeline run and ensure all steps produce verifiable outputs.

1. Successful Pipeline Run:

- Detect and crop face/lips correctly for all pilot clips.
- Extract all planned features:
 - SyncNet scores
 - AV-HuBERT embeddings
 - Whisper transcripts
 - MFA alignment outputs
- Generate LLM score for semantic/prosody plausibility.

- Fuse features and output video-level and segment-level predictions (real/fake).

2. Evidence to Capture:

- Save sample cropped lip frames for verification of preprocessing.
- Export feature vectors for a few clips to check feature extraction.
- Record video-level predictions for each clip.
- Optional: visualize SyncNet scores or per-frame predictions to inspect results.

3. Notes:

- This checklist ensures the pipeline runs successfully on the pilot dataset.
- All steps—from face/lip detection, feature extraction, LLM scoring, to fusion and predictions—should produce outputs.
- Evidence captured (frames, features, predictions) will serve as proof for debugging and verification.
- Optional: include a small flowchart showing the pipeline (Preprocessing → Feature Extraction → LLM → Fusion → Prediction) for clarity.