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Video and Image Manipulation Detection Based on ResNet and EfficientNet

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Introduction

- Deepfake technology utilizes machine learning (ML) algorithms to produce fake, highly realistic content.
- The technology is increasingly accessible and affordable, leading to a rise in misuse and cybercrimes.
- Examples of misuse include the spread of manipulated photos or videos intended to damage reputations.



Polda Jatim Ungkap Kasus Penipuan Deepfake Al Kepala Daerah, Pelaku Kantongi Keuntungan Hingga Rp87 Juta

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Reference to Previous Research



Original software publication

FDT: A python toolkit for fake image and video detection

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ABSTRACT

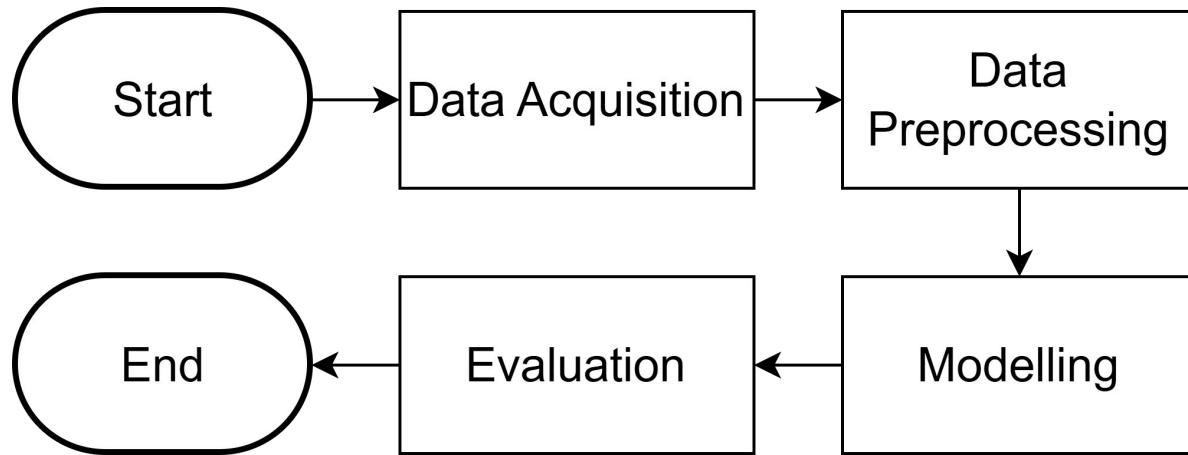
With the advent of readily and widely available applications based on deepfake technology, several cybersecurity threats are on the rise. It is challenging to curtail these threats as deepfakes are realistic and very difficult to detect. The present work proposes a Fake Detection Tool (FDT) that streamlines the procedure of fake detection by incorporating various manipulation techniques and aids users in detecting and visualizing the same. The tool is also integrated with Twitter for streaming facial image posts based on hashtags. It provides an output dataframe and presents statistics of virality, sentiments, etc, using pie charts for better visualization. The proposed tool uses a wide variety of large-scale datasets for training to deal with the fakes in the wild and deploys models that are at par with the cutting-edge models. It is an efficient, user-friendly, and freely available software for fake detection. The source code of the FDT package toolkit is available at https://github.com/surbhiraj786/GUI_Fake-Detection.

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Research Proposal:
This study proposes a focused experimental framework to benchmark two complementary deep architectures:

- ResNet-based models
- EfficientNet-based models

Research Method



Research Method (Data Acquisition)



(FaceForensics++ Sample Data)

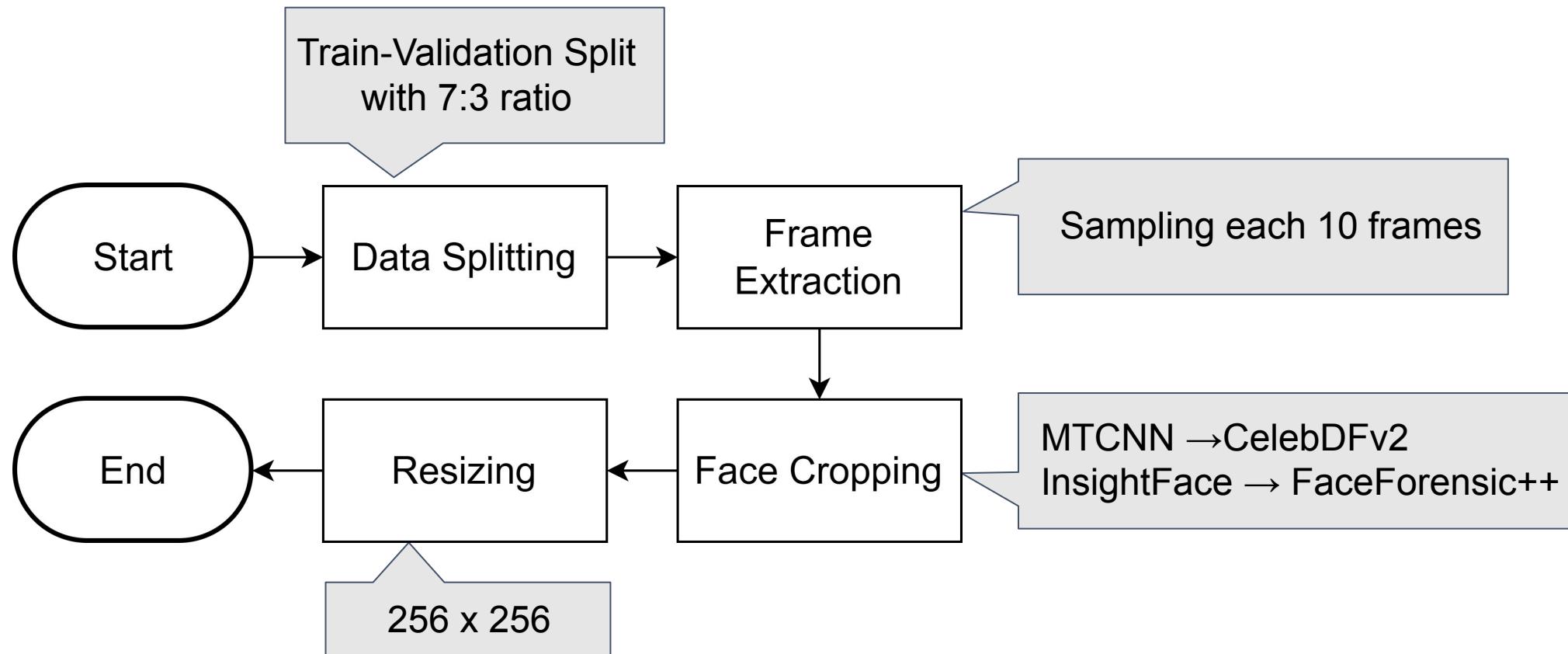


(CelebDFv2 Sample Data)

FaceForensics++: contains 1,000 real videos, and their respective Deepfakes, Face2Face, FaceSwap, and NeuralTextures version.

CelebDFv2: contains 590 real videos and 5,639 fake videos generated from those real videos.

Research Method (Data Preprocessing)



Research Method (Modelling)

Dataset	Model	Training Parameter
CelebDFv2	EfficientNetB7	Epoch: 10(20); Batch Size: 1; Optimizer: Adam; Learning Rate: 1e-4
	ResNet50	Epoch: 10(20); Batch Size: 1; Optimizer: Adam; Learning Rate: 1e-4
FaceForensic++	EfficientNetB7	Epoch: 20; Batch Size: 8; Optimizer: Adam; Learning Rate: 4e-4
	ResNet50	Epoch: 20; Batch Size: 8; Optimizer: Adam; Learning Rate: 1e-5

Research Method (Evaluation Metrics)

$$Accuracy = \frac{TP + TN}{TP + FP + TN + FN}$$

$$Precision = \frac{TP}{TP + FP}$$

$$Recall = \frac{TP}{TP + FN}$$

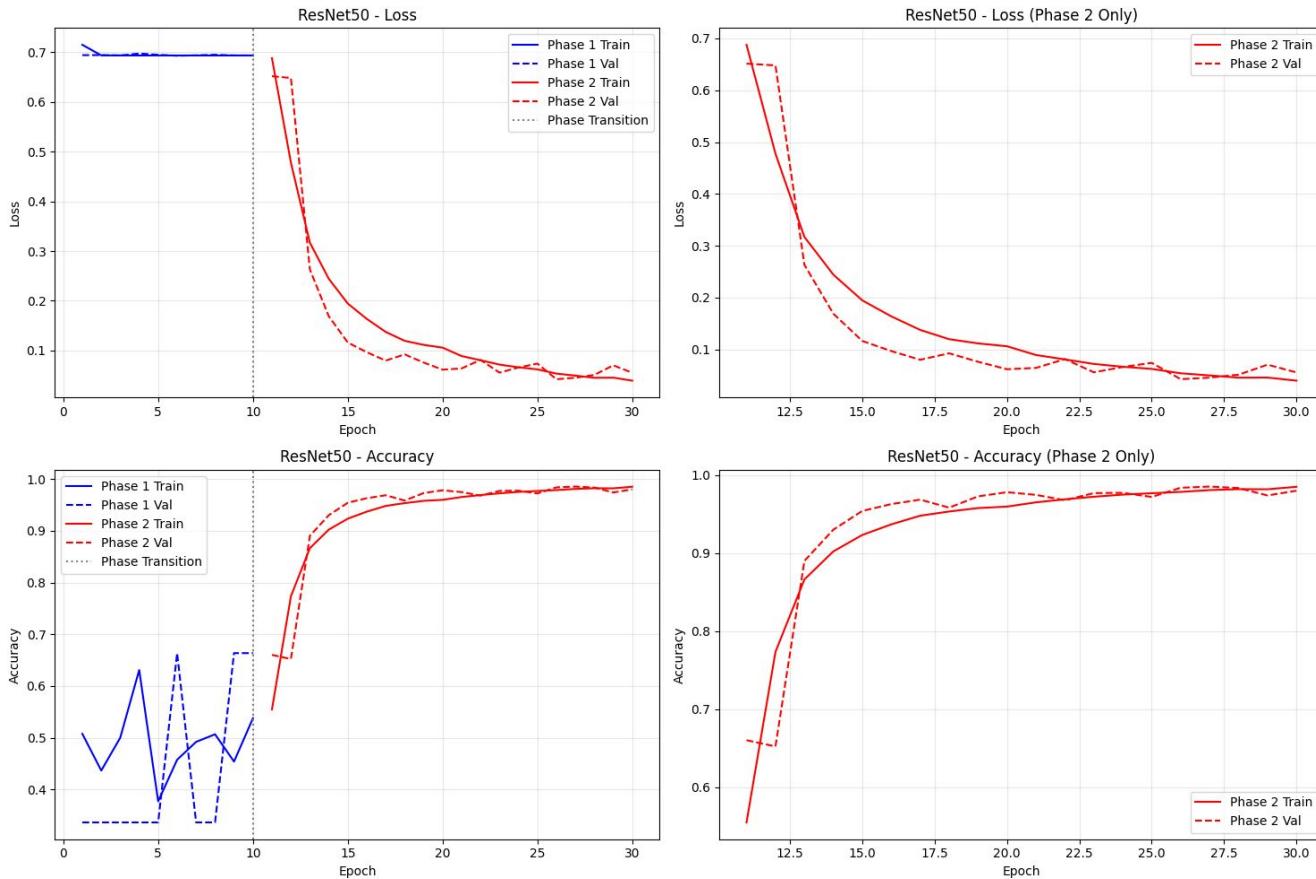
$$F1\ score = \frac{2 \times Precision \times Recall}{Precision + Recall}$$

With:

- TP = True Positive
- TN = True Negative
- FP = False Positive
- FN = False Negative

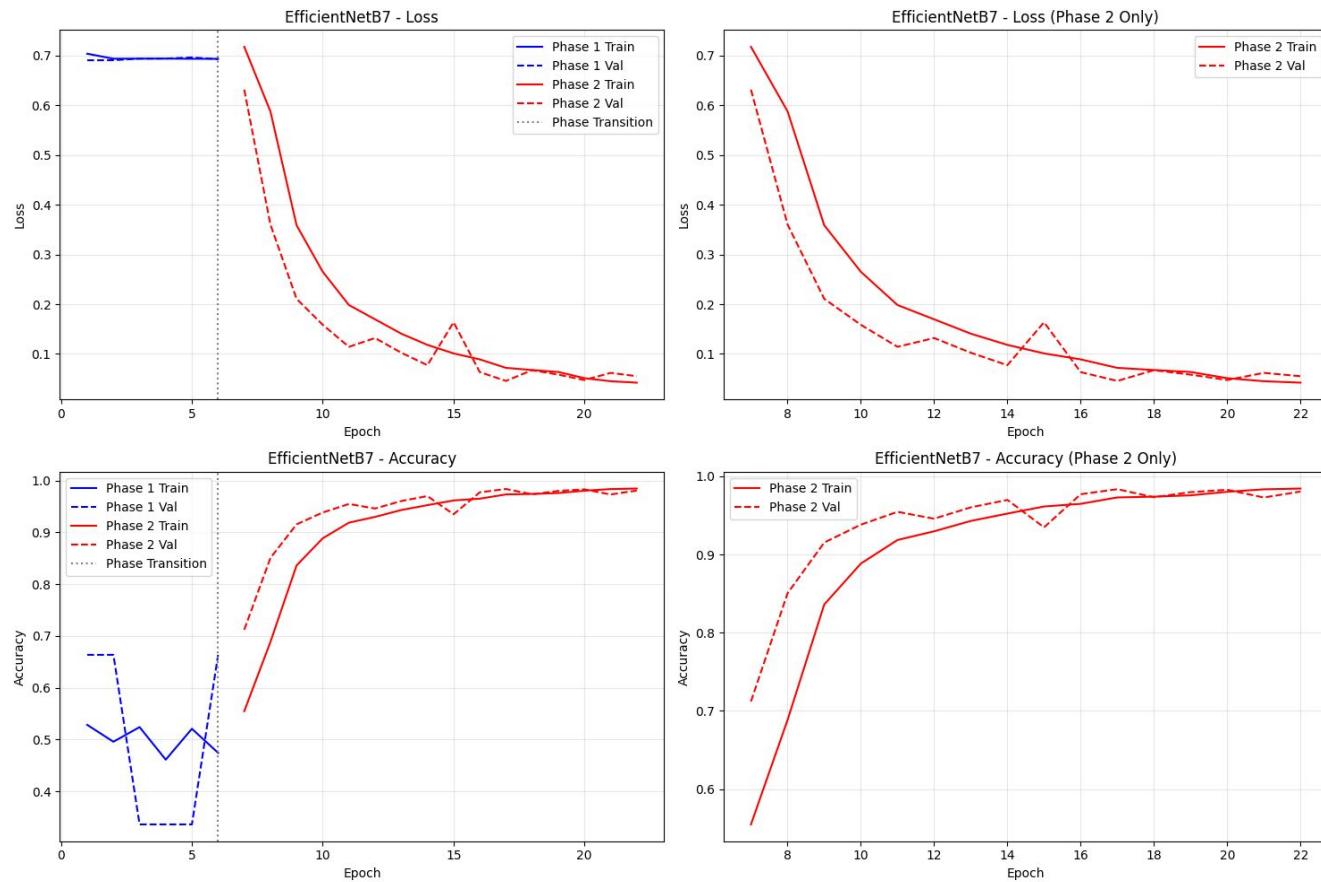
Result and Discussion

ResNet50 model across two-phase training on Celeb-DF-v2



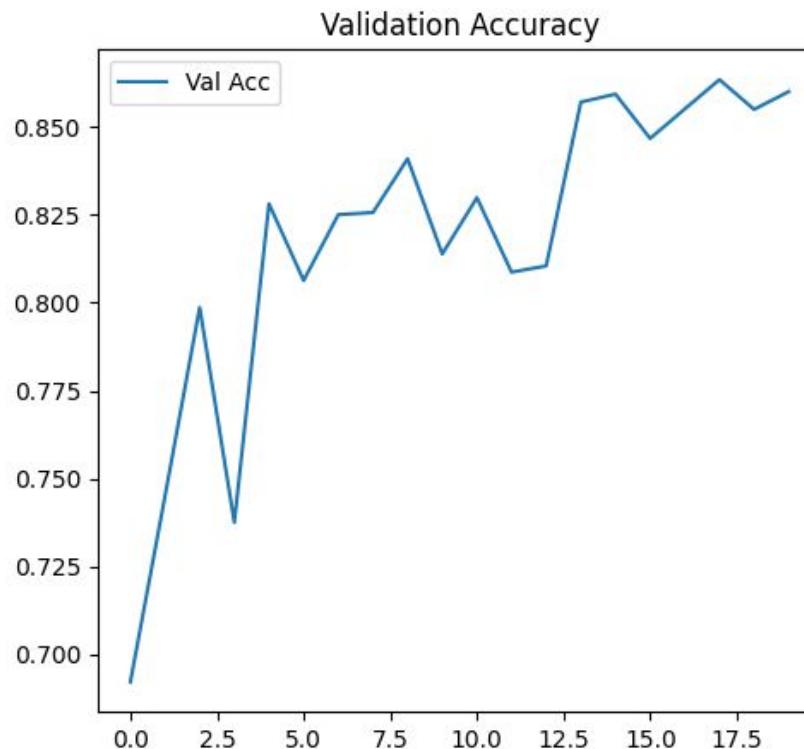
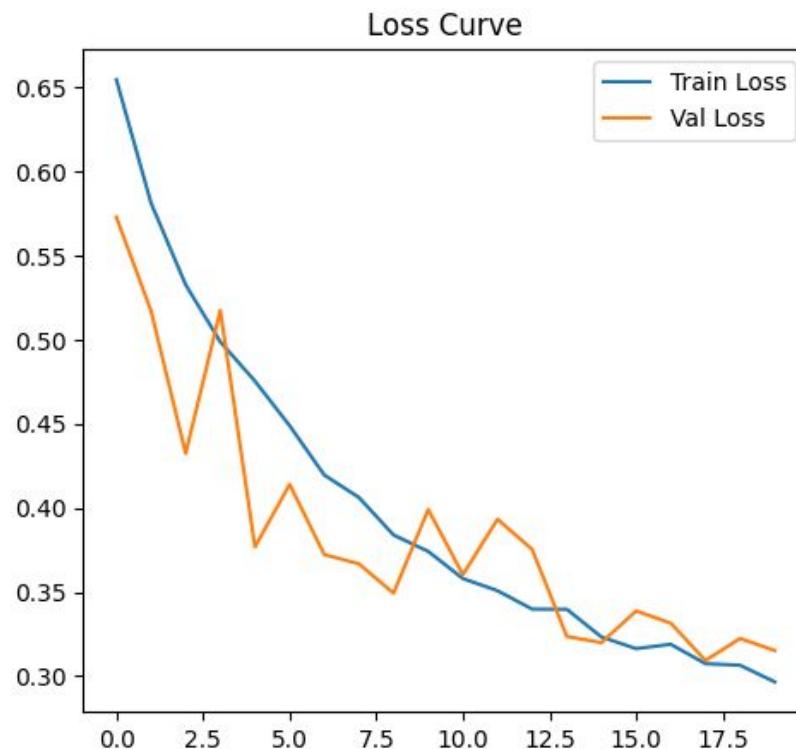
Result and Discussion

EfficientNetB7 model across two-phase training on Celeb-DF-v2



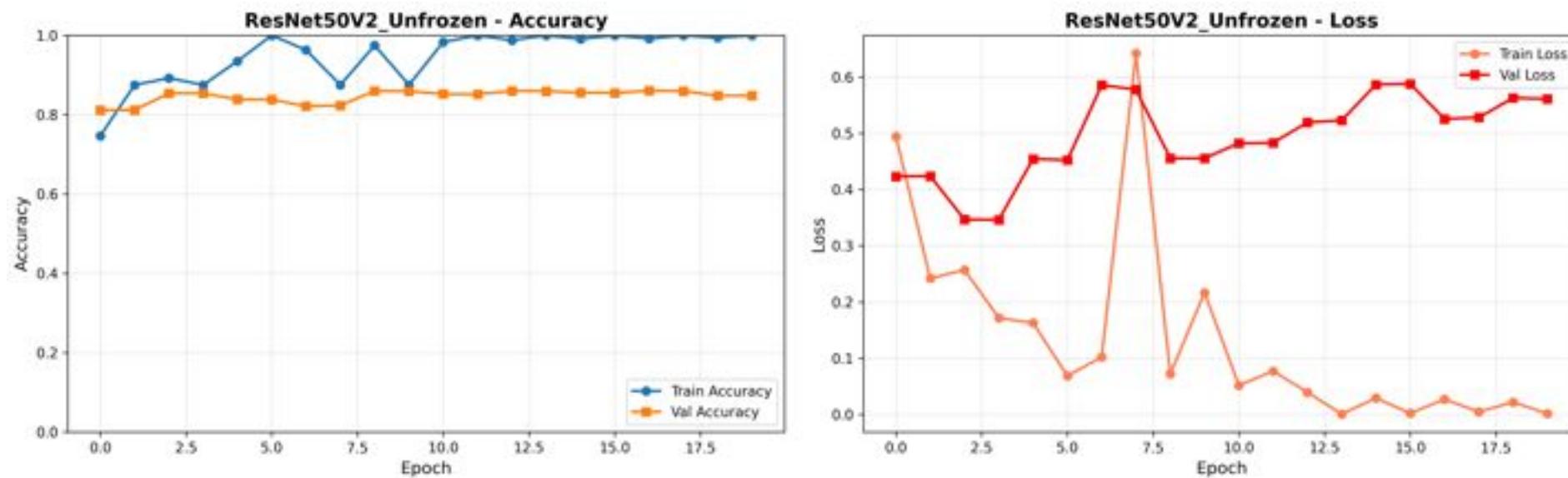
Result and Discussion

EfficientNetB7 model training on FaceForensics++



Result and Discussion

ResNet50 model training on FaceForensics++



Result and Discussion (Evaluation)

Evaluation on Validation Set

Dataset	Model	Accuracy	Precision	Recall	F1-Score
CelebDFv2	EfficientNetB7	97.5%	99.6%	93.4%	96.4%
	ResNet50	93.5%	95.3%	85.9%	90.3%
FaceForensic++	EfficientNetB7	90.1%	83.5%	89.1%	86.2%
	ResNet50	81.98%	90.17%	54.1%	67.63%

Result and Discussion

Evaluation on Unseen Data

Model	Actual	Predicted	Confidence
EfficientNet-FF++	Real	Real	59.16%
	Fake	Real	85.11%
ResNet-FF++	Real	Fake	65.40%
	Fake	Fake	55.35%
EfficientNet-Celeb-DF-v2	Real	Fake	98.63%
	Fake	Real	66.44%
ResNet-Celeb-DF-v2	Real	Real	50.57%
	Fake	Real	54.22%



Conclusion

1. EfficientNetB7 achieved the **highest accuracy on both dataset**, with 90.1% accuracy on FaceForensics++ and 97.5% accuracy on CelebDFv2.
2. **Generalization problem:** Models develop narrow, dataset-specific notions of authenticity, leading to unacceptable false positive rates on unseen authentic content.

Future Works

1. Multi-dataset or progressive training.
2. Data augmentation.
3. Incorporating temporal information.
4. Fusion of additional modalities.
5. Consideration to evaluate model with recently developed deepfake methods.

Thank You



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