

## Multi-Feature Graph Convolution Network for Hindi OCR Verification

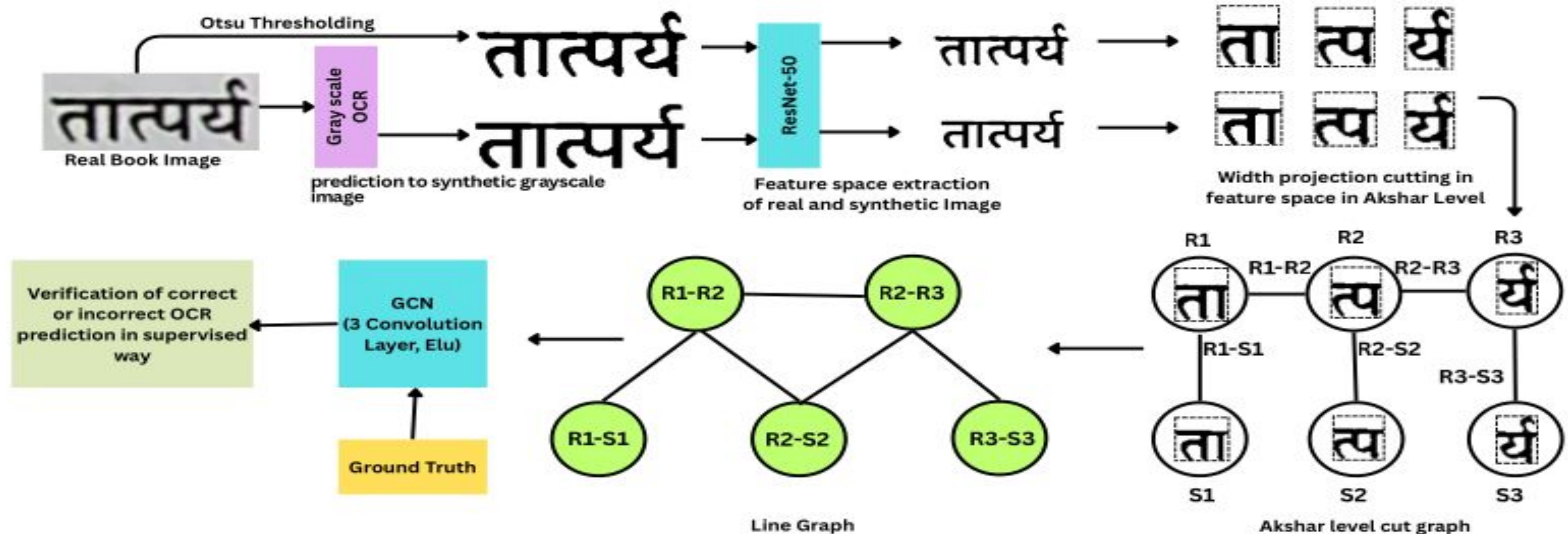
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### Abstract

**Problem:** Traditional OCR struggles with complex Hindi script features like conjuncts and matras, especially in historical or degraded documents.

**Solution:** We propose a novel **Graph Convolutional Network (GCN)** framework to verify OCR predictions by comparing real book images against synthetic images generated from OCR output.

**Method:** The model extracts features using **ResNet-50** and constructs a line graph where nodes represent character pairs.

**Features:** Verification uses semantic (Cross-Entropy) and geometric features (Hu moments, Pixel count) to detect errors.

### Comparison of GNN Architectures

Model Architecture	Precision	Recall	F1-Score
Our Model (3-Layer GCN)	0.7357	0.4285	0.5413
APPNPNet	0.6727	0.6826	0.6825
TAGNet	0.6726	0.6603	0.6485
GATConv	0.6571	0.6309	0.6067

Our 3-Layer GCN achieves the highest precision, minimizing false alarms.

### Performance of Patching Strategies

Patching Strategy	Precision	Recall	F1-Score
Akshar-level	0.6727	0.6926	0.6825
Character-level	0.5363	0.726	0.6169
Random Patch	0.624	0.6602	0.6416
Uniform Patch	0.5327	0.7051	0.6069

Linguistically informed "Akshar-level" patching performs best.

### Conclusion

**Effectiveness:** The framework effectively verifies OCR in diverse Hindi documents, specifically targeting challenging conjunct characters.

**Linguistic Alignment:** Aligning feature segmentation with Hindi linguistic structure (**Akshara-level**) significantly outperforms random or uniform segmentation.

**Future Scope:** We plan to extend this approach to other Indic scripts and integrate attention mechanisms to capture broader contextual dependencies.

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