

| Model                  | Architecture Type                | Dataset Used        | Metric Used           | Reported Accuracy / Score   | Reference                               |
|------------------------|----------------------------------|---------------------|-----------------------|-----------------------------|---|
| U-Net                  | CNN                              | ISIC 2018           | Dice Coefficient      | 85.2%                       | <a href="#">Ali et al., 2022</a>        |
| Attention U-Net        | CNN + Attention                  | ISIC 2018           | Dice Score            | 86.5%                       | <a href="#">Jabbar &amp; Khan, 2022</a> |
| TransUNet              | CNN + ViT Hybrid                 | ISIC 2018, 2017     | Dice Score            | 88.1%                       | <a href="#">Chen et al., 2021</a>       |
| Swin-Unet              | Pure Transformer                 | ISIC 2018, Synapse  | Dice Score            | 88.9%                       | <a href="#">Cao et al., 2021</a>        |
| MedT                   | Pure Transformer                 | ISIC 2018           | Dice Score            | 87.8%                       | <a href="#">Valanarasu et al., 2021</a> |
| GS-TransUNet           | ViT + Gaussian Splatting         | ISIC 2018           | Dice Score            | 89.3%                       | <a href="#">Zhao et al., 2024</a>       |
| ScaleFusionNet         | Swin Transformer + DeformConv    | ISIC 2018           | Dice Score            | <b>89.5%</b>                | <a href="#">Shao et al., 2024</a>       |
| SUTrans-NET            | Dual Encoder (CNN + Transformer) | ISIC 2018           | Accuracy              | 90.1%                       | <a href="#">Wang et al., 2023</a>       |
| Attention Swin U-Net   | Swin Transformer + Cross-Attn    | ISIC 2018           | Dice Score            | <b>90.6%</b>                | <a href="#">Kim et al., 2024</a>        |
| ViT-UNet (Our Project) | CNN + ViT Hybrid                 | ISIC 2018 (assumed) | Dice Score / Accuracy | <b>~91.2%</b> (as reported) | <i>Our Project</i>                      |

| Model / Study   | Architecture Type                              | Dataset(s)                    | Key Features   | Performance Metrics  | Reference                                   |
|---|--|-------------------------------|--|--|---|
| <b>Hybrid Deep Learning Framework</b>                 | U-Net + Inception-ResNet-v2 + ViT              | ISIC 2020, HAM10000           | Combines U-Net for segmentation, Inception-ResNet-v2 for feature extraction, and Vision Transformer for feature refinement | Accuracy: 98.65%;<br>Sensitivity: 99.20%;<br>Specificity: 98.03% | <a href="#">PubMed, 2024</a>                |
| <b>Advanced Deep Learning Models</b>                  | Context Aggregation-based DNN                  | ISIC 2020                     | Utilizes morphological operations and context aggregation for preprocessing and segmentation                               | Classification Accuracy: 93.40%                                  | <a href="#">MDPI Sensors, 2025</a>          |
| <b>Skin Lesion Classification via Ensemble Method</b> | Modified Inception ResNet v2 + EfficientNet-B4 | ISIC 2018                     | Incorporates Soft-Attention mechanism to enhance feature extraction  | Specific performance metrics not provided                        | <a href="#">Springer, 2024</a>              |
| <b>Boundary-Aware Segmentation Network (BASNet)</b>   | CNN + Transformer                              | ISIC 2016/2017/2018, HAM10000 | Applies hybrid loss and residual refinements to handle occlusion and poor contrast images                                  | Specific performance metrics not provided                        | <a href="#">Frontiers in Medicine, 2025</a> |

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|--|------------------------------------|---------------|---|---|--|
| <b>Skin Lesion Segmentation Model Based on Improved U2Net</b>          | Improved U2Net                     | ISIC 2018     | Enhances U2Net architecture for better segmentation performance             | Specific performance metrics not provided                 | <a href="#">ACM ISAIMS, 2023</a>                               |
| <b>Skin Cancer Segmentation and Classification Using ViT</b>           | Vision Transformer                 | HAM10000      | Utilizes pre-trained ViT for segmentation and classification                | Classification Accuracy: 96.15%                           | <a href="#">ResearchGate, 2024</a>                             |
| <b>Deep Learning-based Skin Lesion Segmentation and Classification</b> | U-Net + CNN + ViT                  | HAM10000      | Employs U-Net for segmentation followed by CNN and ViT for classification   | Specific performance metrics not provided                 | <a href="#">ResearchGate, 2024</a>                             |
| <b>Health of Things Melanoma Detection System</b>                      | Deep Learning + Fine-Tuning Models | Not specified | Applies deep learning and fine-tuning models in edge computing environments | Detection Accuracy: 96.39%; Segmentation Accuracy: 96.50% | <a href="#">Frontiers in Communications and Networks, 2024</a> |