

# Investigating and Exploiting Image Resolution for Transfer Learning-based Skin Lesion Classification

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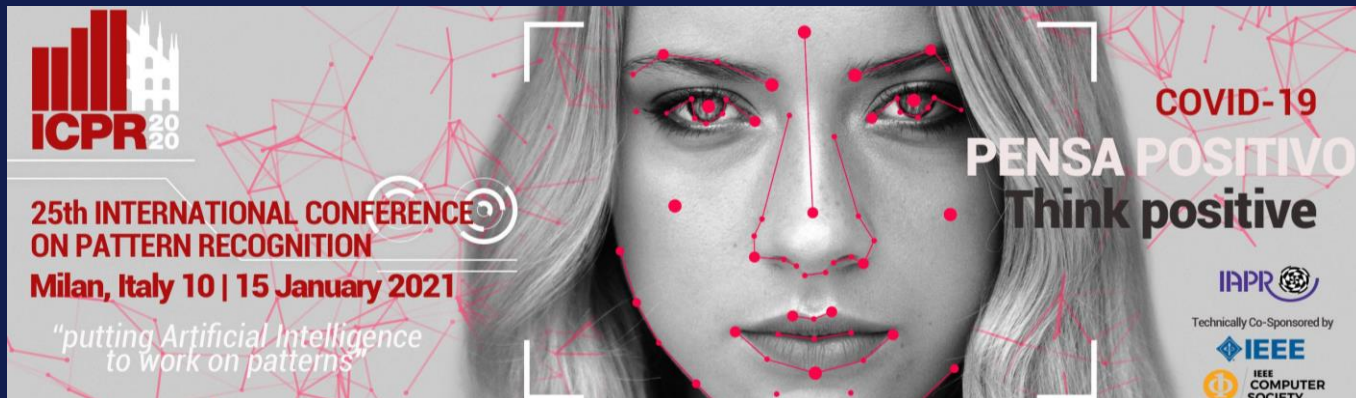
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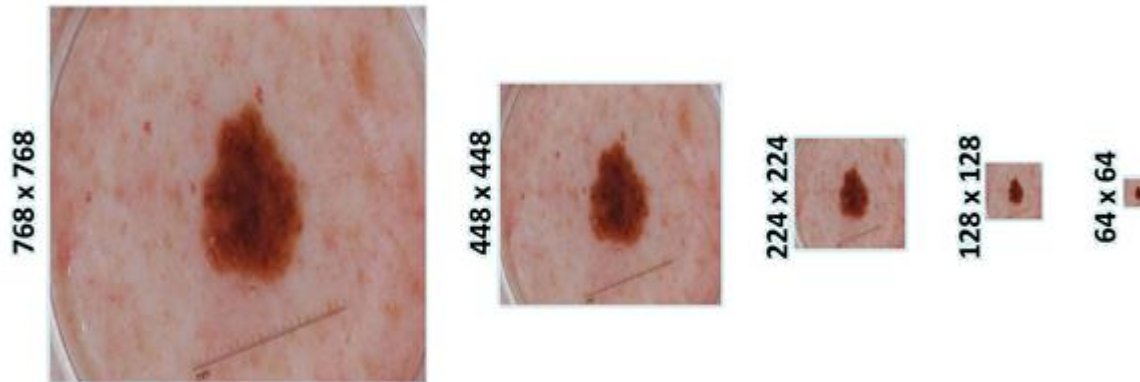


# Introduction

- Skin cancer is among the most common cancer types
- Automated skin lesion classification is typically addressed using fine-tuned convolutional neural networks (CNNs)
- To fine-tune a CNN, image down-sampling is a common pre-processing step, but:
  - What is the proper down-sampling factor?
  - Is it useful to exploit images with different sizes for fine-tuning?

# Aims

- Analyze the **effect of image size** on **skin lesion classification performance**

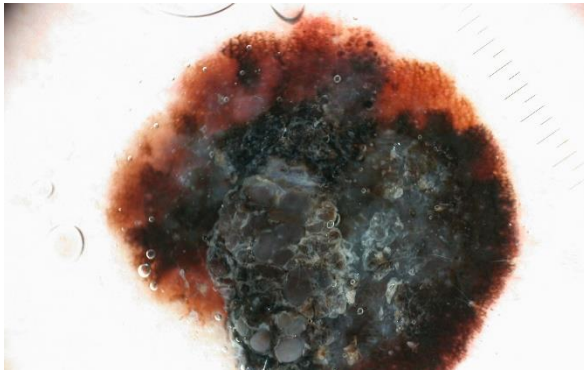


- Develop a **multi-scale and multi-network fusion approach** for skin lesion classification

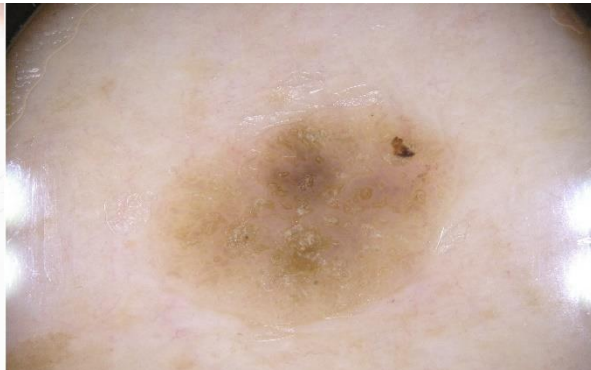
# Method

- **Dataset**

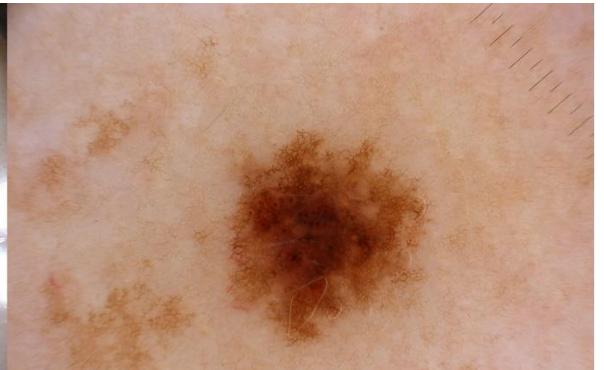
- **Training set:** 2,187 images from ISIC 2016 and ISIC 2017 challenge datasets [1]
- **Test set:** 600 images from ISIC 2017 test set
- **Three skin lesion classes:** malignant melanoma (MM), Seborrheic Keratosis (SK), and Benign nevi (BN)



MM



SK



BN

## Reference

[1] <https://isic-archive.com/>

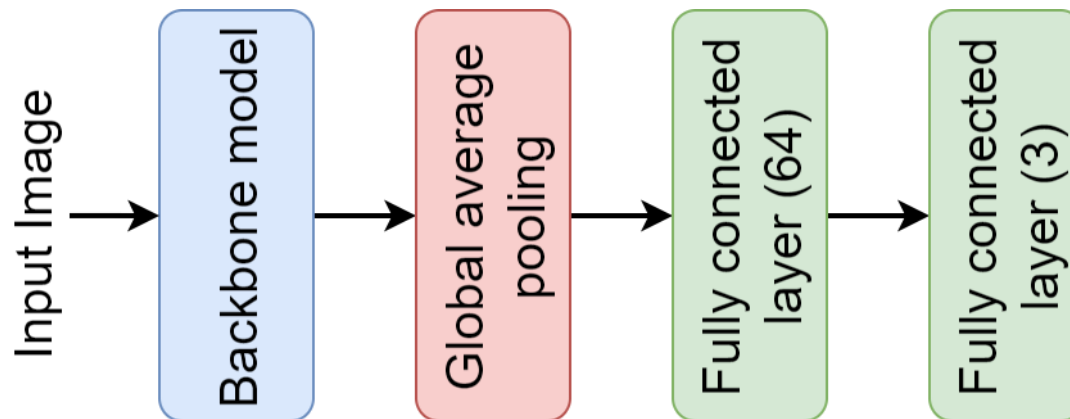
# Method

- **Pre-processing**

- Gray world color constancy normalization
- ImageNet mean intensity subtraction
- Offline data augmentation by rotation and flipping
- Resizing:
  - ✓ 64 x 64 pixels
  - ✓ 128 x 128 pixels
  - ✓ 224 x 224 pixels
  - ✓ 448 x 448 pixels
  - ✓ 768 x 768 pixels

# Method

- **Fine-tuning pre-trained CNNs**
  - ResNet-18 [2]
  - ResNet-50 [2]
  - DenseNet-121 [3]



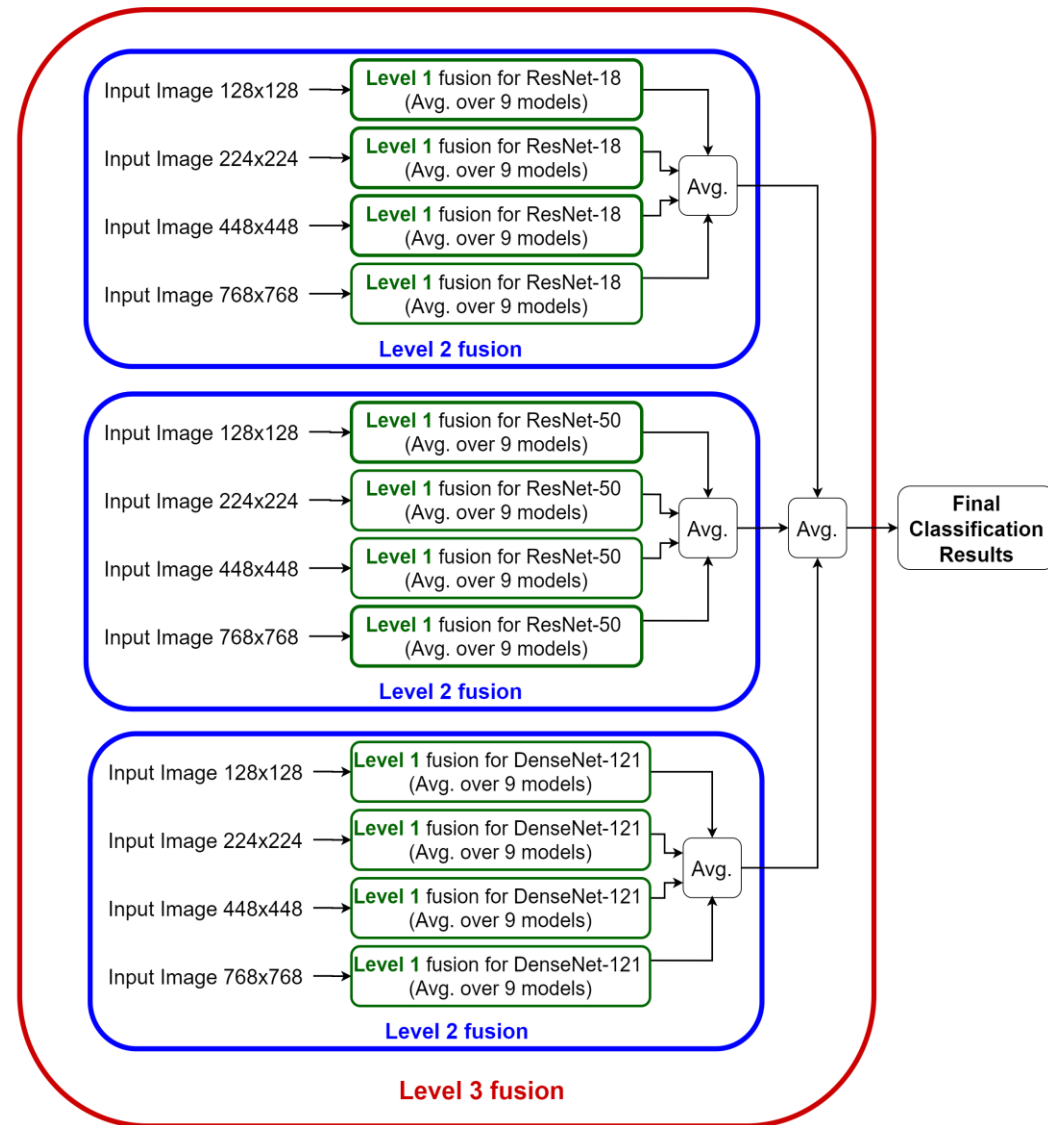
## References

- [2] He et al: Deep Residual Learning for Image Recognition, CVPR, 2016  
[3] Simonyan et al: Densely Connected Convolutional Networks, CVPR, 2017

# Method

- **3-Level fusion**

- **Level 1**
- **Level 2**
- **Level 3**



# Evaluation

- **Area under the curve (AUC)** for two classification problems namely MM vs. all task and SK vs. all classification task (identical to the ISIC 2017 challenge evaluation index)



# Results

**Classification performance** (area under the curve, AUC in %) of ResNet-18 for different image sizes based on **level 1** fusion

Size (pixels)	MM AUC (%)	SK AUC (%)	Avg. AUC (%)
64 x 64	78.86	89.55	84.21
128 x 128	85.46	93.39	89.42
224 x 224	85.37	93.81	89.59
448 x 448	89.20	95.54	92.37
768 x 768	88.89	95.85	92.37

- Using resized images with a size of 64 x 64 → significant drop in performance
- Using resized images with a size of larger than 128 x 128 → comparable results
- Increasing size → slight improvement in performance

# Results

**Classification performance** (AUC in %) based on **level 2** and **level 3** fusion

Network	Sizes	MM AUC (%)	SK AUC (%)	Avg. AUC (%)
ResNet-18 <b>level 2</b> fusion	all	89.12	96.26	92.69
ResNet-50 <b>level 2</b> fusion	all	88.50	96.03	92.27
DenseNet-121 <b>level 2</b> fusion	all	87.69	95.77	91.73
<b>Level 3</b> fusion	all	<b>89.16</b>	<b>96.57</b>	<b>92.86</b>

# Results

## Comparison to the state-of-the-art methods

Network	Sizes	MM AUC (%)	SK AUC (%)	Avg. AUC (%)
Matsunaga <i>et al.</i> [4]	NA	86.8	95.3	91.1
Mahbod <i>et al.</i> [5]	224	87.3	95.5	91.4
Zhang <i>et al.</i> [6]	224	87.5	95.8	91.7
Yan <i>et al.</i> [7]	256	88.3	NA	NA
<b>Three-level fusion</b>	<b>all</b>	<b>89.16</b>	<b>96.57</b>	<b>92.86</b>

### References

- [4] Matsunaga et al: Image classification of melanoma, nevus and seborrheic keratosis by deep neural network ensemble, arXiv, 2017
- [5] Mahbod et al: Fusing fine-tuned deep features for skin lesion classification, CMIG, 2019
- [6] Zhang et al: Attention residual learning for skin lesion classification, IEEE TMI, 2019
- [7] Yan et al: Melanoma recognition via visual attention, IPMI, 2019

# Thank you for your attention!

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