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*An ENAS Based Approach for Constructing Deep Learning Models
for Breast Cancer Recognition from Ultrasound Images*

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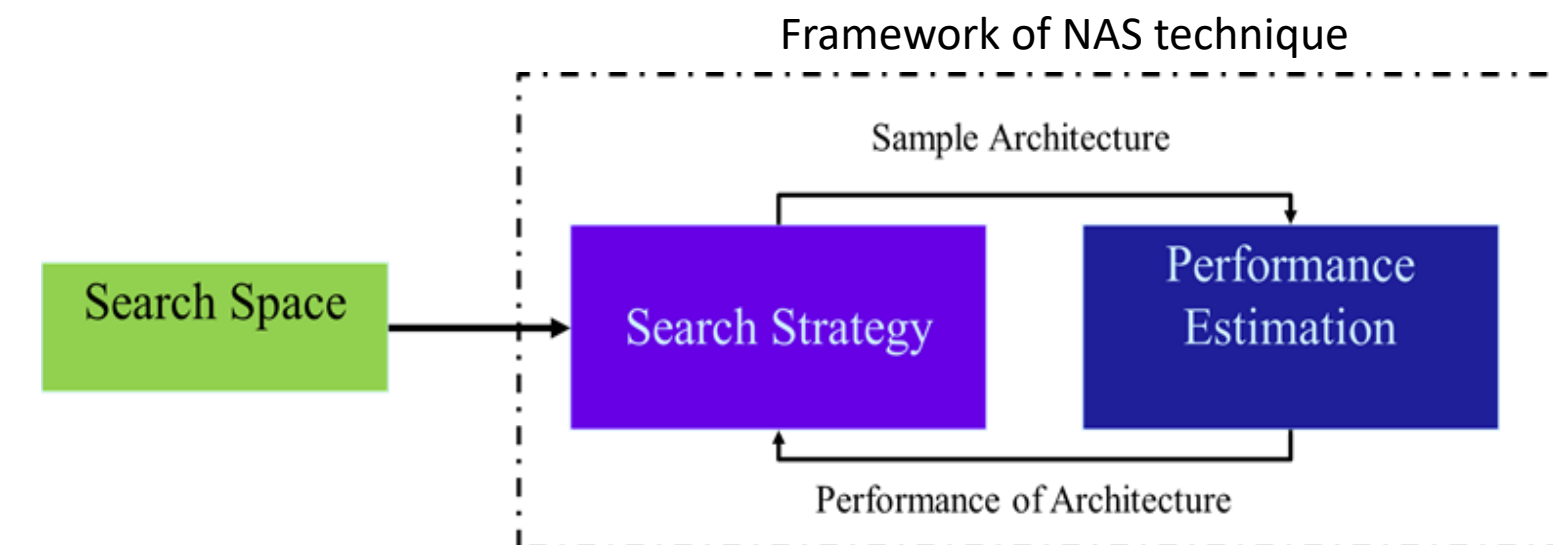
Introduction

Background

- Widespread use of handcrafted CNN architectures (AlexNet, VGGNet, GoogLeNet ,ResNet ...etc.)
- CNN architectures: hard to design
- Recent Development: automatic architecture search
 - Neural Architecture Search (NAS)
 - Efficient Neural Architecture Search (ENAS)

Aim of the Study

To Investigate effectiveness of ENAS for breast cancer recognition from US images



Methods

Hyperparameter setting for Searching:

- Batch size = 8
- Image size = 100×100
- Other hyperparameter settings: the default values of ENAS

Searching for Optimal Cells

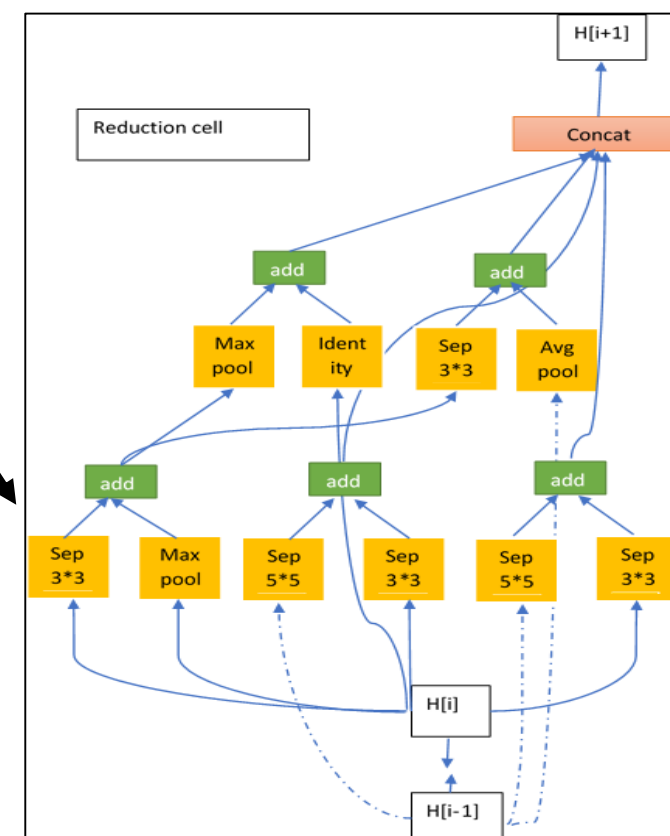
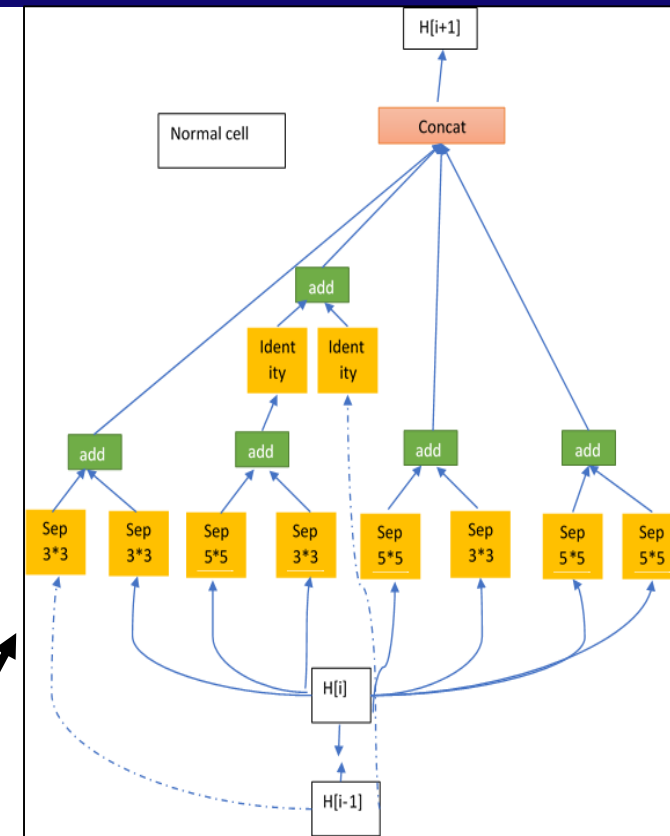
Search Space (Micro)

The controller (RNN)

Performance Estimation

Performance of Architecture

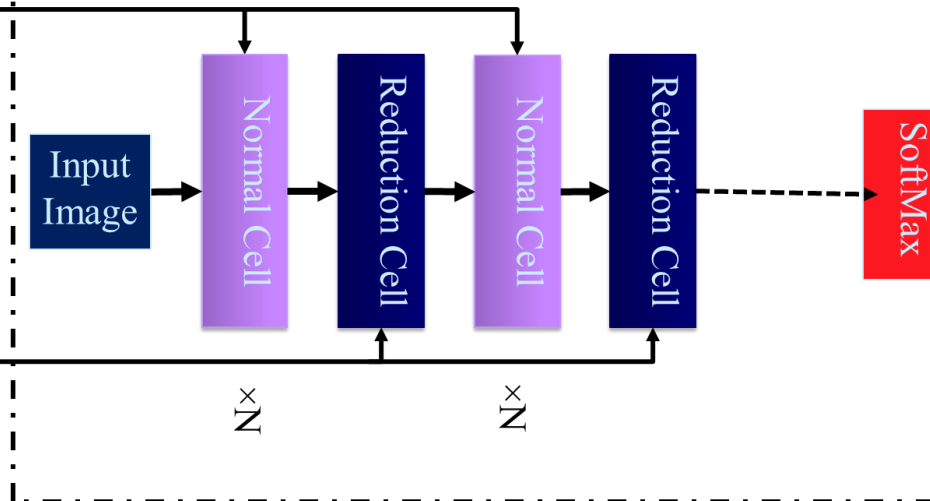
Generated Architecture



Hyperparameter setting for final CNN Model:

- Batch size = 8
- No. of epochs = 100
- Image size = 100×100
- Other hyperparameter settings: the default values of ENAS

Design CNN model based on Optimal Cells



Layers in final architectures:

- ENAS 17 ($5*N, R, 5*N, R, 5*N$)
- ENAS 7 ($N, R, N, R, 3*N$)

Optimal cells(Norman and Reduction) generated by ENAS for Breast Cancer classification

MIDL 2020

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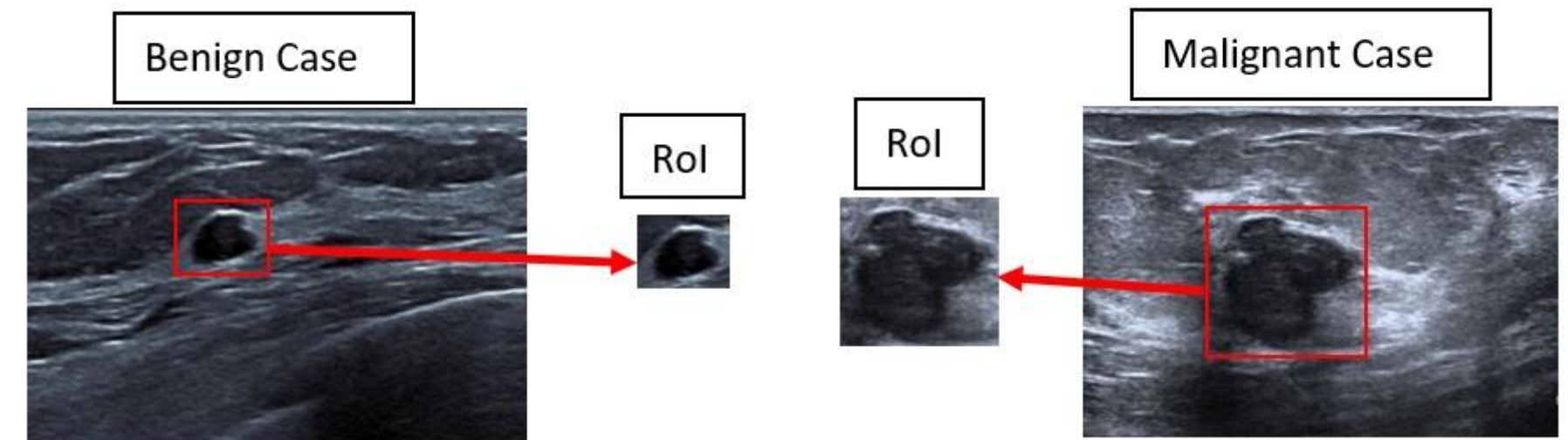
Data and Results

Dataset

- Ultrasound images for breast lesions (262 Benign and 262 Malignant images)
- Different US machine makes

Data Preparation

- Manual cropping of RoI by radiologist
- Training data augmentation:
 - Geometric Methods: Rotation (90,180 and 270), and Mirroring
 - Singular Value Decomposition (SVD) (25, 35 and 45)
- Bicubic Resizing (100×100)



Examples of ultrasound images with labeled region of interest

ENAS Models Performance and Comparison with Models of Other Architectures

Models	TNR	TPR	PR	Accuracy	#Parameters
ENAS 17	86.7%	92.0%	87.5%	89.3%	4,251,780
ENAS 7	90.9%	86.7%	91.0%	88.8%	2,342,484
AlexNet	51.6%	48.5%	50.0%	50.0%	56,858,656
CNN3 [1]	80.5%	75.6%	84.0%	78.1%	619,202

Concluding Remarks

- Conclusion:
 - Investigated the efficacy of the ENAS approach for designing CNN architectures for breast lesion classification from US.
 - Demonstrated that the ENAS technique reduces human interventions in CNN architecture design.
 - The optimized architectures lead to more accurate classification yet simpler models than hand-crafted alternatives for breast lesion classification.
- Future work:
 - Evaluating ENAS models on external datasets
 - Exploiting ENAS architectures for other types of tumors/lesions from ultrasound images