

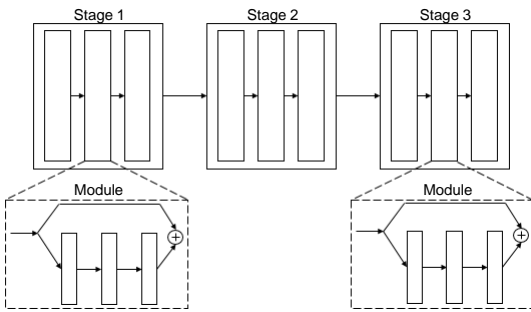
## Introduction

- Neural Architecture Search methods are capable of learning to design high-performance convolutional networks automatically
- Current NAS are computationally expensive
  - They evaluate a large number of candidate architectures
- In this work, we propose a simple, effective, and efficient approach to discover convolutional architectures

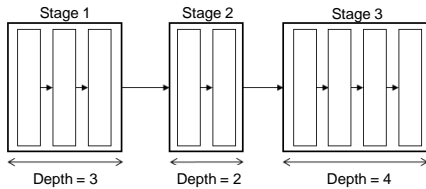
## Proposed Neural Architecture Search

### Problem Definition

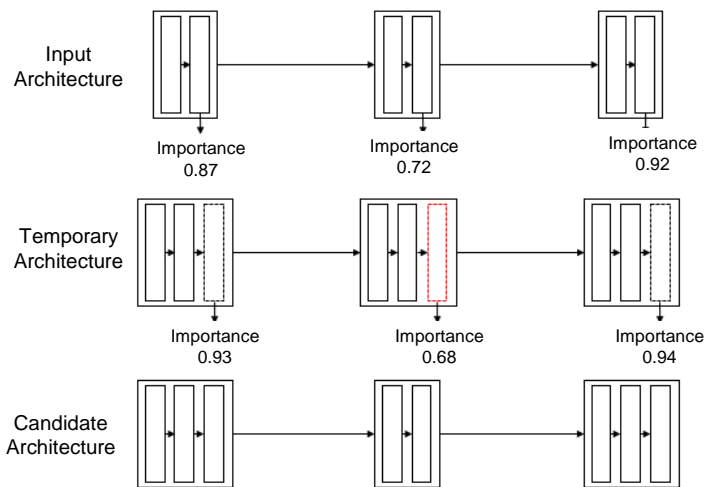
- Structure of modern architectures
  - Depth (number of modules) is the same for all stages



- We propose to learn the depth of each stage based on its importance
  - Stages with low importance are kept shallow
  - Stages with high importance become deeper



### Proposed Method



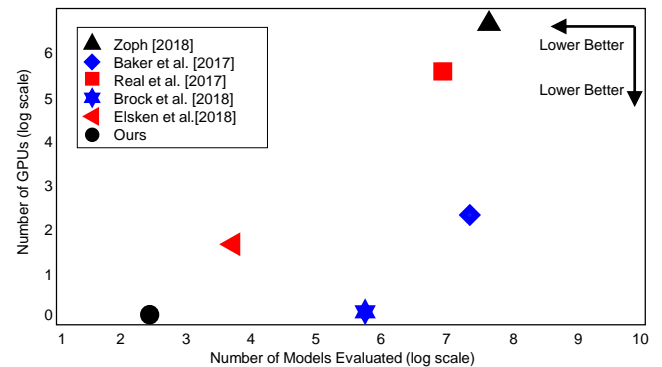
## Experiments

### Comparison with Human-Designed Architectures

Architecture	Depth	Param. ↓ (Million)	FLOP ↓ (Million)	Accuracy ↑
ResNet44	44	0.66	97	92.83
Ours (it=1)	<b>43</b>	<b>0.60</b>	<b>92</b>	<b>93.38</b>
ResNet56	56	0.86	125	93.03
Ours (it=3)	59	<b>0.69</b>	130	<b>93.36</b>
ResNet110	110	1.7	253	93.57
Ours (i=5)	<b>67</b>	<b>0.88</b>	<b>149</b>	<b>94.27</b>

### Comparison with State-of-the-Art NAS

Model	Evaluated ↓ Models	GPUs ↓	Param. ↓ (Million)	Accuracy ↑
Zoph et al. [2018]	20, 000	800	2.5	94.51
Real et al. [2017]	1, 000	250	5.4	94.60
Dong and Yang [2019]	240	<b>1</b>	2.6	<b>96.25</b>
Jin et al. [2019]	60	<b>1</b>	---	88.56
Ours (it=5)	<b>11</b>	<b>1</b>	<b>2.3</b>	94.74



## Conclusions

- We demonstrate that it is possible to design high-performance convolutional architectures by inserting layers based on their importance
- Compared to NAS strategies, our method is more efficient, as it evaluates one order of magnitude fewer models and discovers architectures on par with the state of the art
- Code is available at <https://github.com/arturjordao/StageWiseArchitectureSearch>



## ACKNOWLEDGMENTS

The authors would like to thank the National Council for Scientific and Technological Development – CNPq (Grants 140082/2017-4, 438629/2018-3 and 309953/2019-7) and the Minas Gerais Research Foundation – FAPEMIG (Grants APQ-00567-14 and PPM-00540-17)

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