



# Weight Agnostic Neural Networks

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# Innate abilities in animals





# Innate abilities in machines

**Super-resolution**



**Corrupted**



**Deep image prior**

**Denosing**



**Corrupted**



**Deep image prior**

**Inpainting**



**Corrupted**



**Deep image prior**

**Inpainting**



**Corrupted**



**Deep image prior**

Ulyanov, D., Vedaldi, A., & Lempitsky, V. (2018). Deep image prior. In *Proceedings of the IEEE Conference on Computer Vision and Pattern Recognition (CVPR)*

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To what extent can neural net architectures alone encode solutions to tasks?

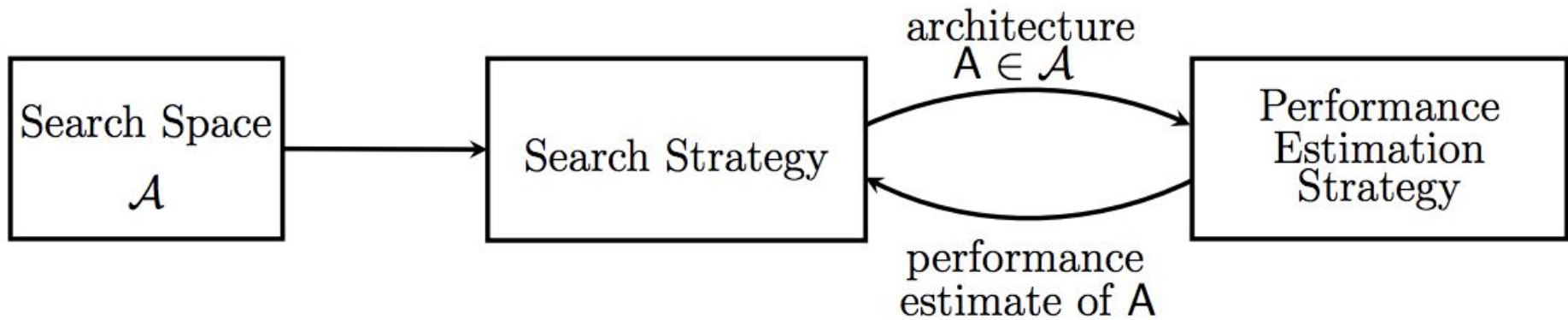


# Neural Architecture Search

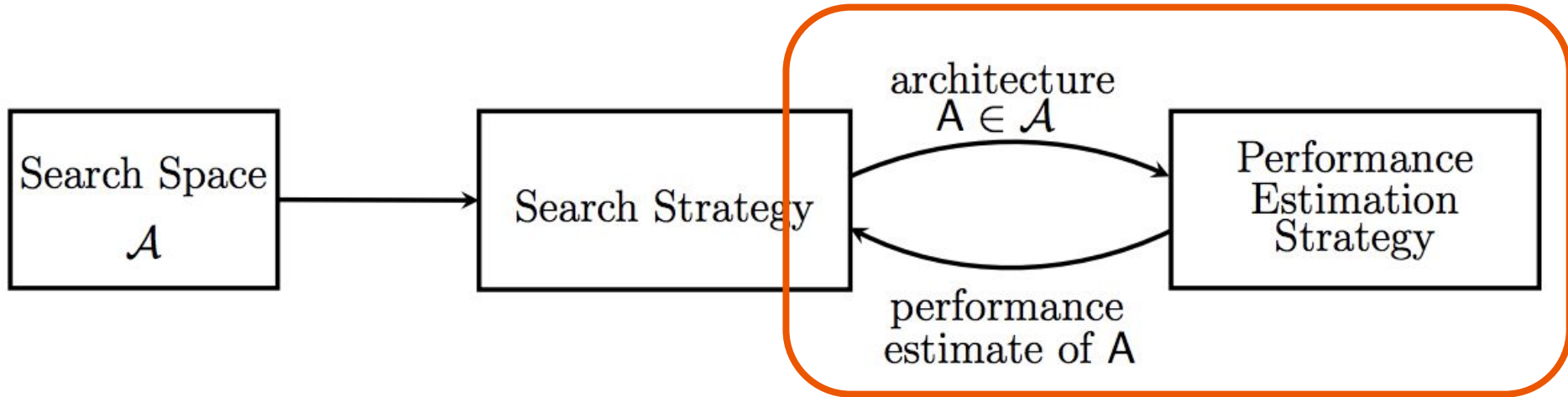
## Searching for trainable networks

- Architectures, once trained, outperform hand designed networks
- Expensive -- training of network required to judge performance
- Solution is still encoded in weights of network, not in architecture

# Searching for Architectures



# Searching for Architectures



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How can we search for  
architectures...not weights?





# Search without Training

## Assume weights are drawn from a particular distribution

- Search for architecture to perform given weights from this distribution

## Replace inner loop training with sampling

- Draw new weights from distribution at each rollout
- Judge network on zero-shot performance



# Weight Sharing

## Single shared weight value used for all connections

- Weight value selected from distribution at each rollout
- Reduces number of parameters of network to 1
  - Reliable expected reward of topology

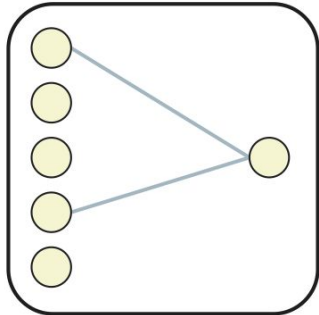
## Architecture search

- Explore space of network topologies
- Judge network architecture based on performance over a series of rollouts

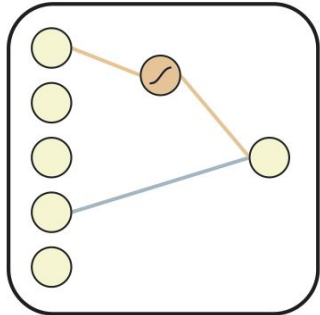


# Topology Search

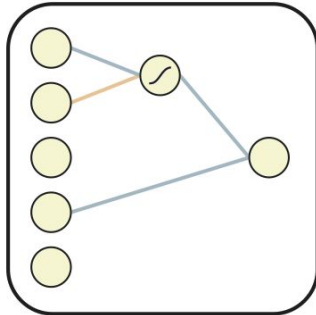
Minimal Network



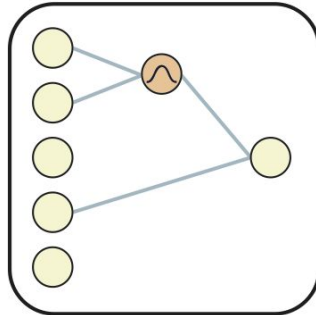
Insert Node



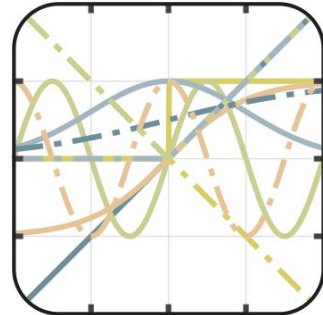
Add Connection



Change Activation

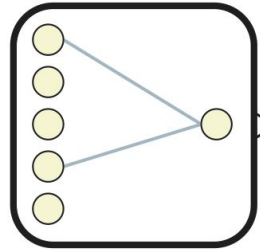


Node Activations



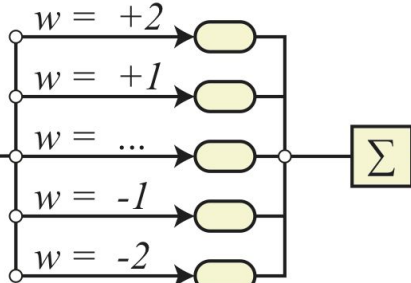
### 1.) Initialize

Create population of minimal networks.



### 2.) Evaluate

Test with range of shared weight values.



### 3.) Rank

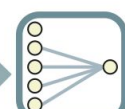
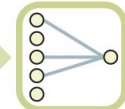
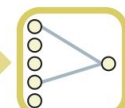
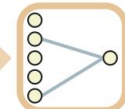
Rank by performance and complexity

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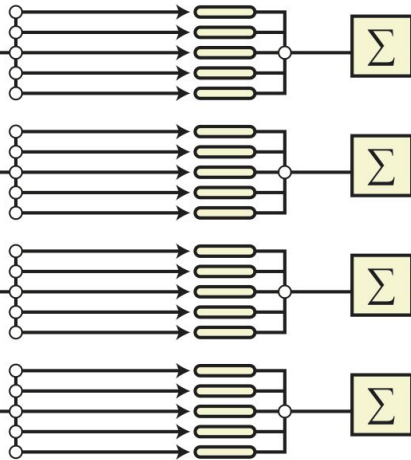
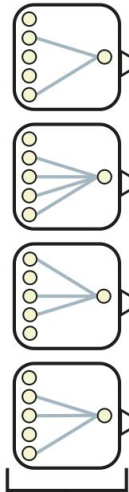
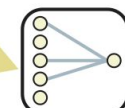
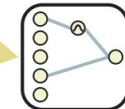
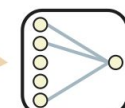
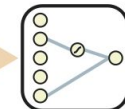
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### 4.) Vary

Create new population by varying best networks.

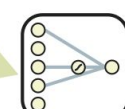
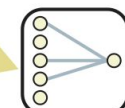
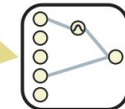
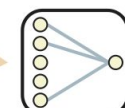
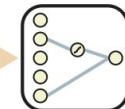
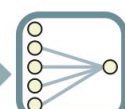
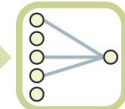
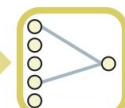
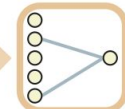


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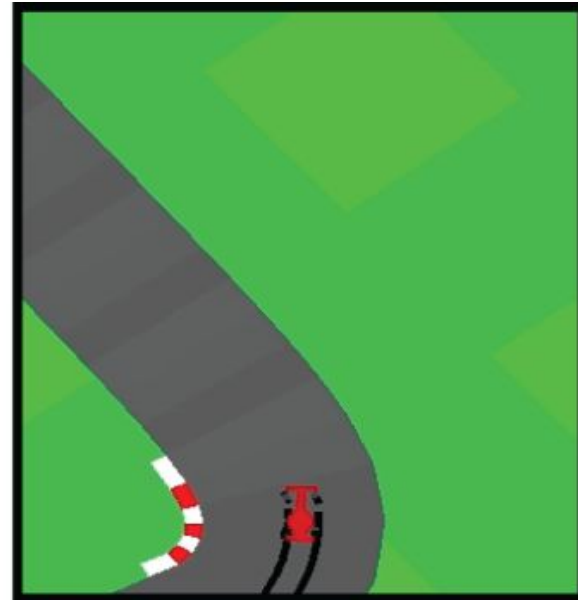
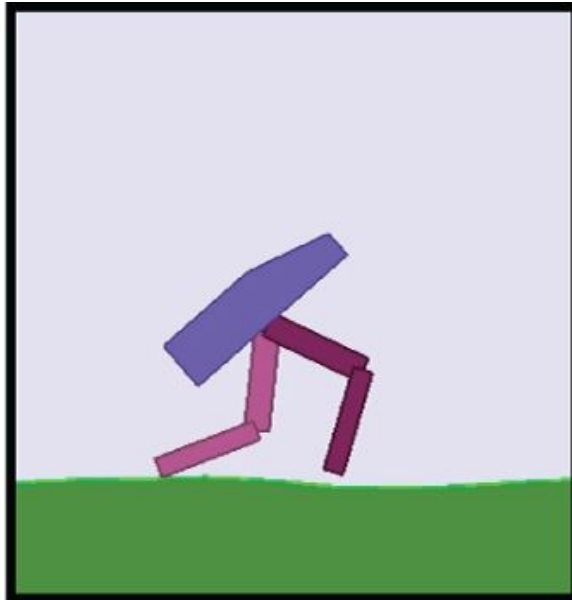
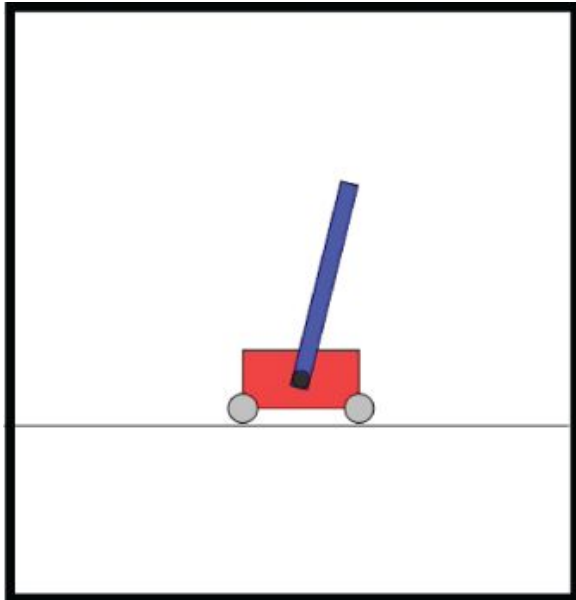
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## WANNs find solutions in variety of RL tasks







## WANNs perform with and without training

| <b>Swing Up</b> | Random Weights  | Random Shared Weight | Tuned Shared Weight | Tuned Weights  |
|-----------------|-----------------|----------------------|---------------------|----------------|
| WANN            | <b>57 ± 121</b> | <b>515 ± 58</b>      | <b>723 ± 16</b>     | <b>932 ± 6</b> |
| Fixed Topology  | 21 ± 43         | 7 ± 2                | 8 ± 1               | 918 ± 7        |

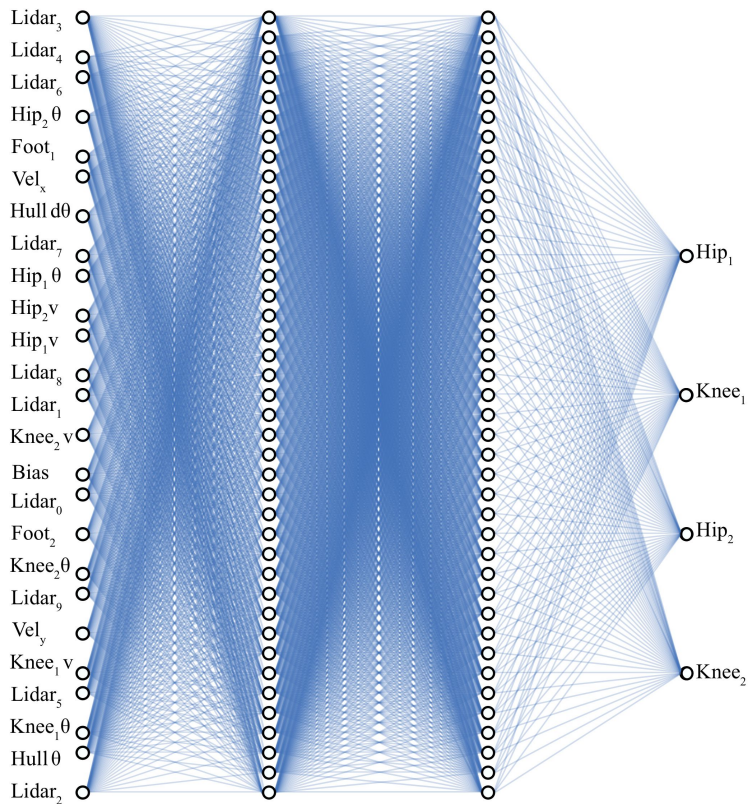
  

| <b>Biped</b>   | Random Weights  | Random Shared Weight | Tuned Shared Weight | Tuned Weights  |
|----------------|-----------------|----------------------|---------------------|----------------|
| WANN           | <b>-46 ± 54</b> | <b>51 ± 108</b>      | <b>261 ± 58</b>     | 332 ± 1        |
| Fixed Topology | -129 ± 28       | -107 ± 12            | -35 ± 23            | <b>347 ± 1</b> |

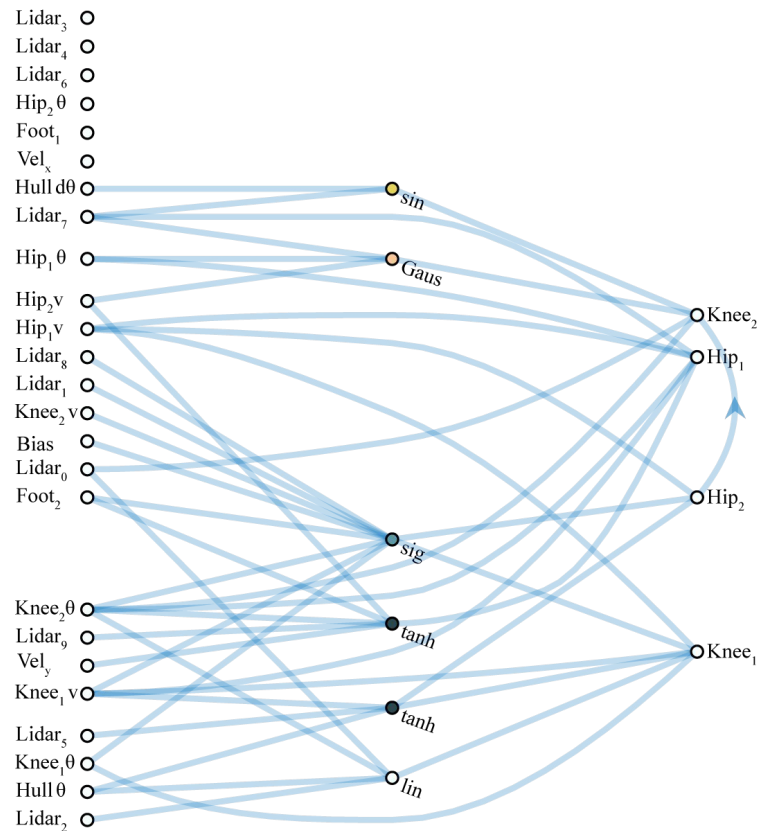
  

| <b>CarRacing</b> | Random Weights  | Random Shared Weight | Tuned Shared Weight | Tuned Weights   |
|------------------|-----------------|----------------------|---------------------|-----------------|
| WANN             | <b>-69 ± 31</b> | <b>375 ± 177</b>     | <b>608 ± 161</b>    | 893 ± 74        |
| Fixed Topology   | -82 ± 13        | -85 ± 27             | -37 ± 36            | <b>906 ± 21</b> |

# *ANN Bipedal Walker* *(2760 connections, weights)*



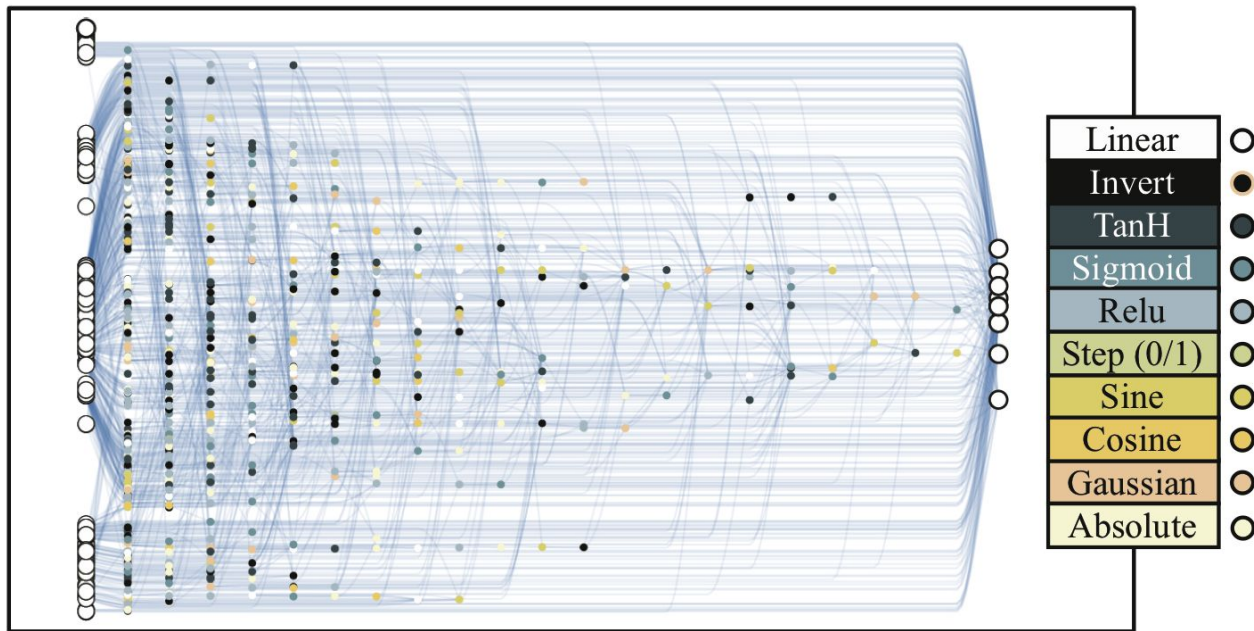
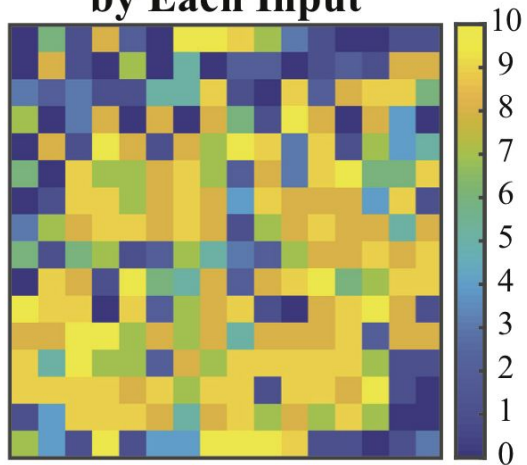
# *WANN Bipedal Walker* *(44 connections, 1 weight)*



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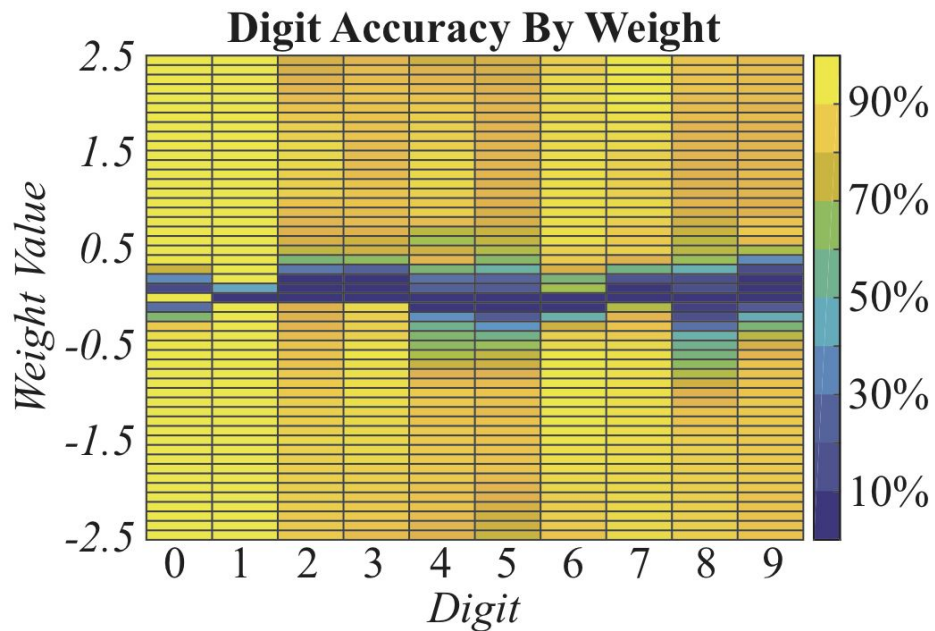
**Can we find WANNs outside of  
reinforcement learning domains?**

**# of Classes Used  
by Each Input**



| <b>WANN</b>      | <b>Test Accuracy</b> |
|------------------|----------------------|
| Random Weight    | 82.0% $\pm$ 18.7%    |
| Ensemble Weights | 91.6%                |
| Tuned Weight     | 91.9%                |
| Trained Weights  | 94.2%                |

| <b>ANN</b>        | <b>Test Accuracy</b> |
|-------------------|----------------------|
| Linear Regression | 91.6% [50]           |
| Two-Layer CNN     | 99.3% [12]           |







# Searching for Building Blocks

## First steps toward a different kind of architecture search

- Network architectures with innate biases can perform a variety of tasks
- ...and these biases can be found through search

## Weight tolerance as a heuristic for new building blocks

- ConvNets and LSTMs can work even untrained
- Finding novel building blocks at least as important as new arrangements of those which already exist

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interactive article @:

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poster @:

**wednesday 10:45**