

Goal Driven PCB Synthesis Using Machine Learning and Cloud Scale Compute

Taylor Hogan Distinguished Engineer

March 27, 2023



Themes

- PCB Design is a Manual Process
 - The solution space is large
- Large Training Sets Available
 - Can be autonomously created
- Designs Mostly Share Common DNA
 - Same parts, Same Manufacturing Facilities
- Competitive Disadvantage if Not using ML
 - It works, let me prove it
- Cloud Scale Compute

Design Spaces are Huge

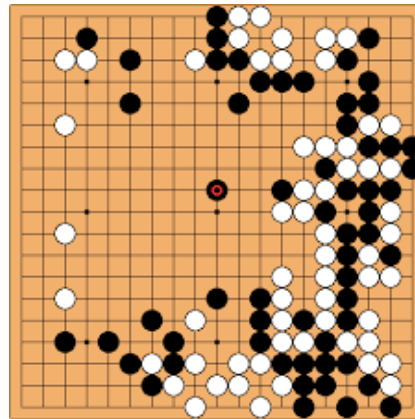
- Complexity of search space is much larger than those of recent successes
 - No hope of using pure trial and error or machine learning
- Need algorithms that automatically learn heuristics to prune search space based off a set of designs or styles

Chess



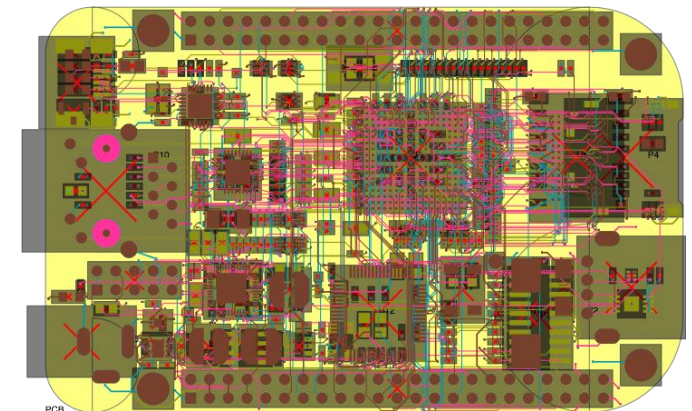
10^{47}

Go



10^{170}

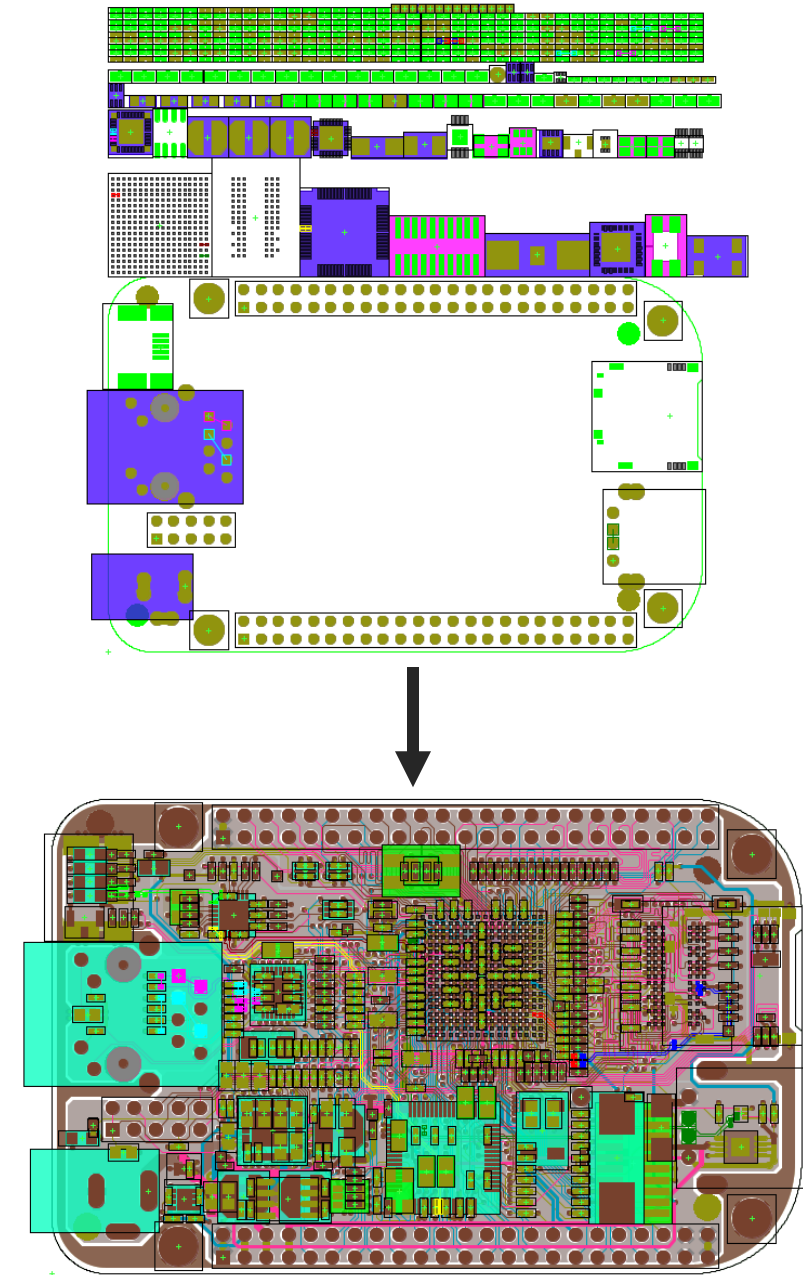
PCB Placement



10^{262}

Design is Difficult

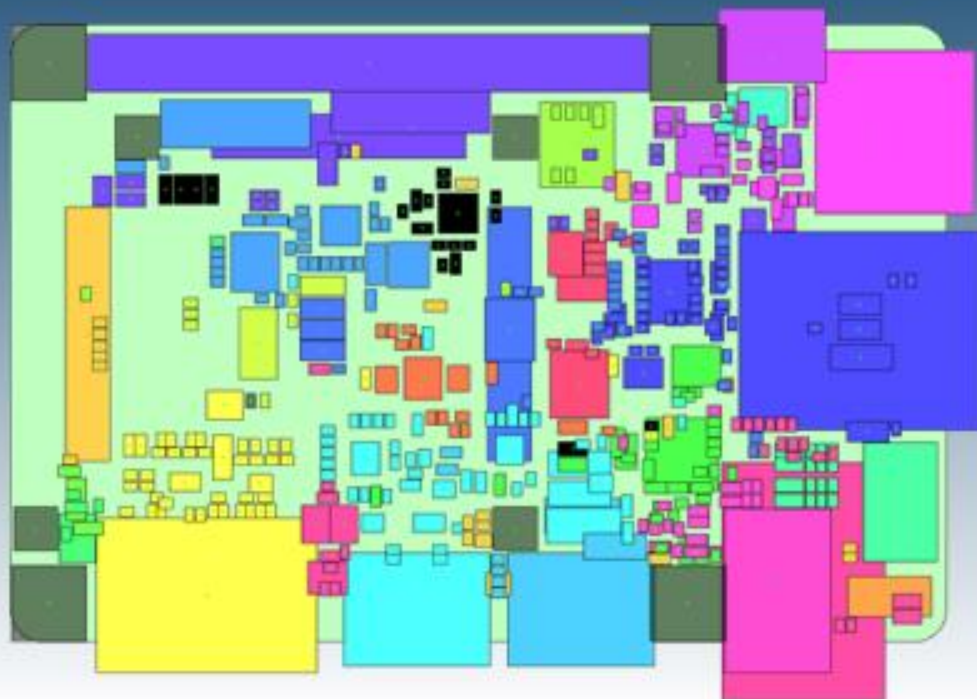
- Computationally Intractable
 - Placement is either ill-posed or NP-Hard
 - Routing is NP-hard (even determining if there is a solution is difficult)
- Tradeoff between speed and quality of solutions
 - Exact solutions take time or are rigid
 - Approximate solutions only generalize within distribution
 - Have to design algorithms that are not general purpose, but flexible



X AI

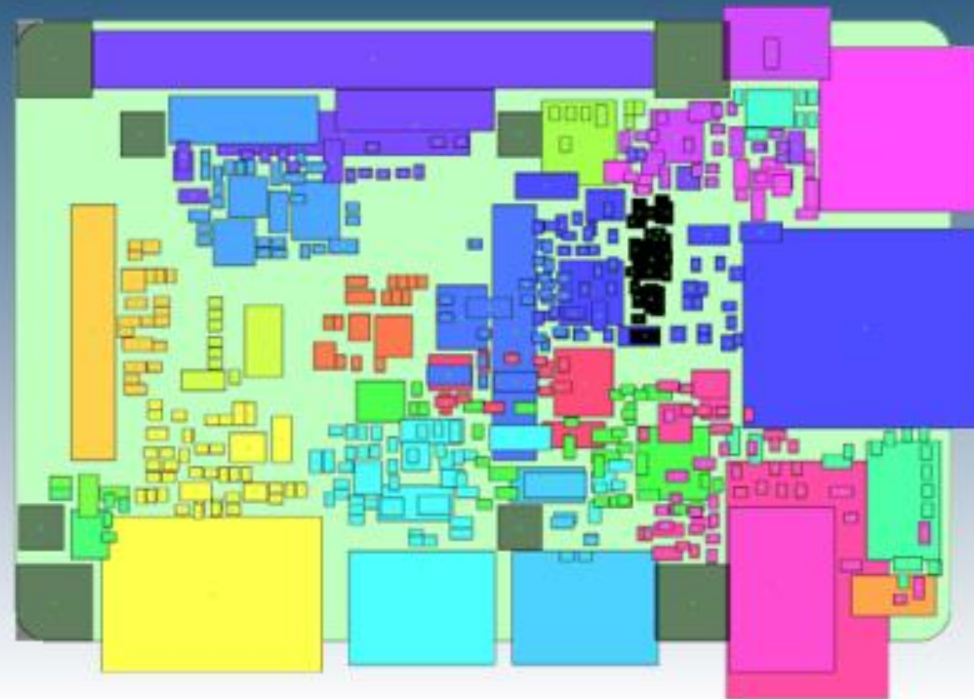
Artificial Intelligence to Reduce Time to Design PCBs – Google Coral Board

Human Placement



~ 3 days Placement

Allegro X: AI Clustering



75 minutes
14% better wirelength
Incremental setup time 15 mins

What problems can machine learning solve?

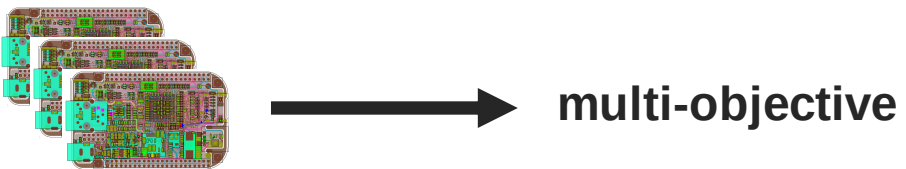
- *ML for prediction problems*



- *ML for design problems*

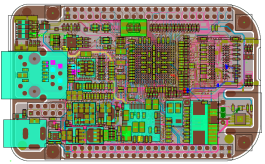


- *ML at scale*



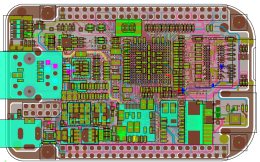
What problems can machine learning solve?

- *ML for prediction problems*



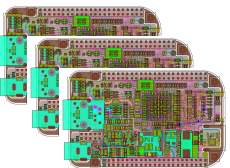
metrics

- *ML for design problems*



actions

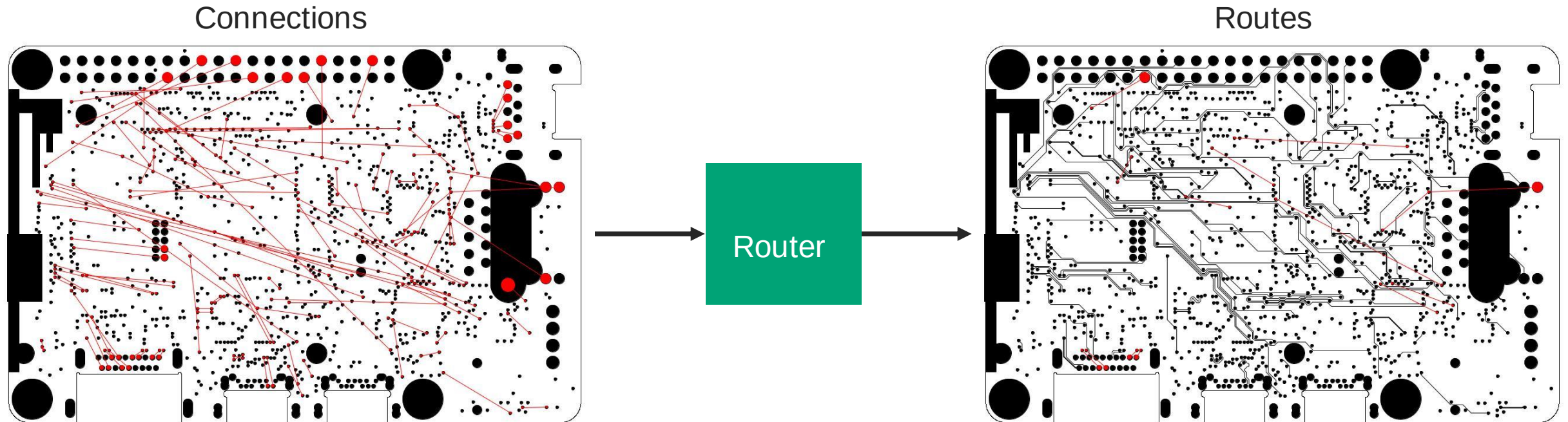
- *ML at scale*



multi-objective

Route Prediction Problems

Can we predict the disconnects before we route?

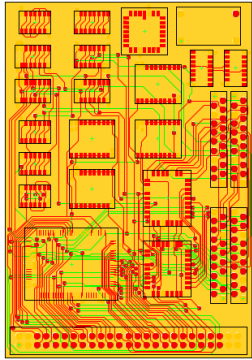


Route Prediction Problems

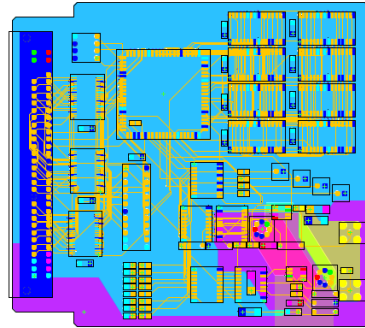
Building a Dataset

	Board 1	Board 2	Board 3	Board 4	Board 5	Board 6
Training Set Size	12,136	12,362	12,320	20,878	12,761	
Test Set Size	2,962	4,038	4,168	3,102	3,260	1,663

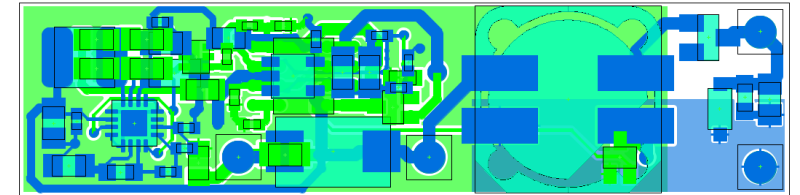
Simple



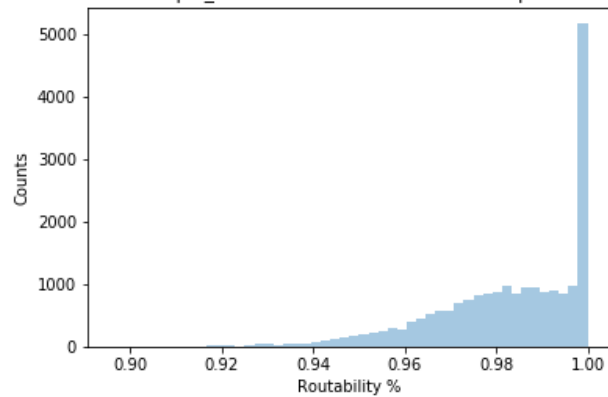
Test9



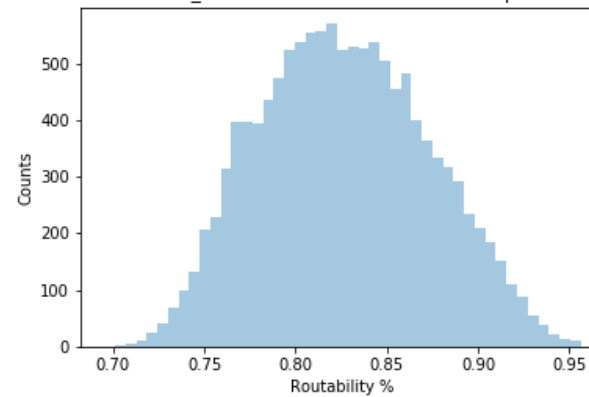
MFC_TI_Flyback_Rev2



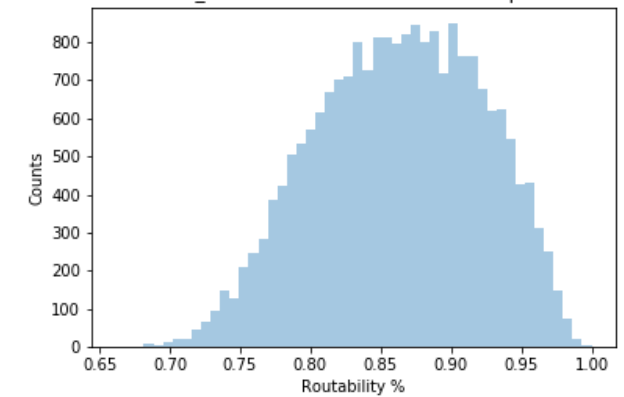
simple_dataset distribution: 21289 samples



test9_dataset distribution: 12362 samples

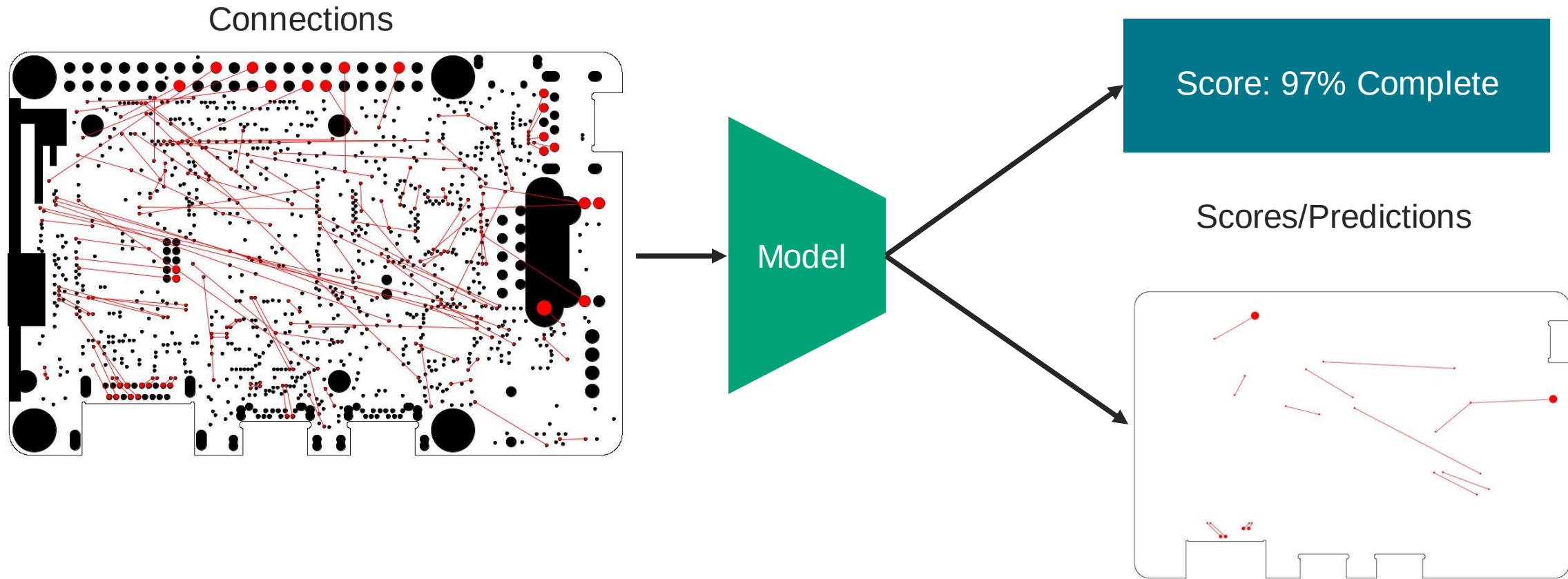


mfc_dataset distribution: 20878 samples



Route Prediction Problems

Can we predict the disconnects before we route?



What problems can machine learning solve?

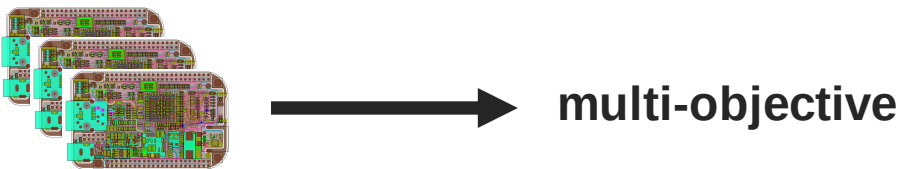
- *ML for prediction problems*



- *ML for design problems*



- *ML at scale*



Why Reinforcement Learning?

- General framework for making decisions
- Transfer knowledge from previous problems we have solved in the past
- Learn to solve certain classes of problems efficiently



The Problem

- RL problems are typically cast as a *Markov Decision Process*

The Problem

- RL problems are typically cast as a *Markov Decision Process*

Markov : The current state is all the knowledge I need to make predictions about the next state.



The Problem

- RL problems are typically cast as a **Markov Decision Process**

Markov : The current state is all the knowledge I need to make predictions about the next state.



Decision : Which action should I take to get towards my goal state?



The Problem

We can take advantage of this structure

- RL problems are typically cast as a **Markov Decision Process**

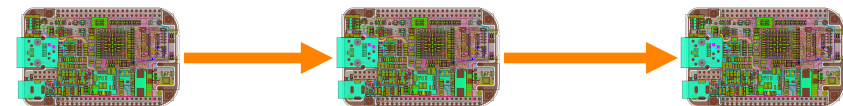
Markov : The current state is all the knowledge I need to make predictions about the next state.



Decision : Which action should I take to get towards my goal state?



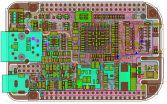
Process : More than one action will be required to reach my goal.



The Solution

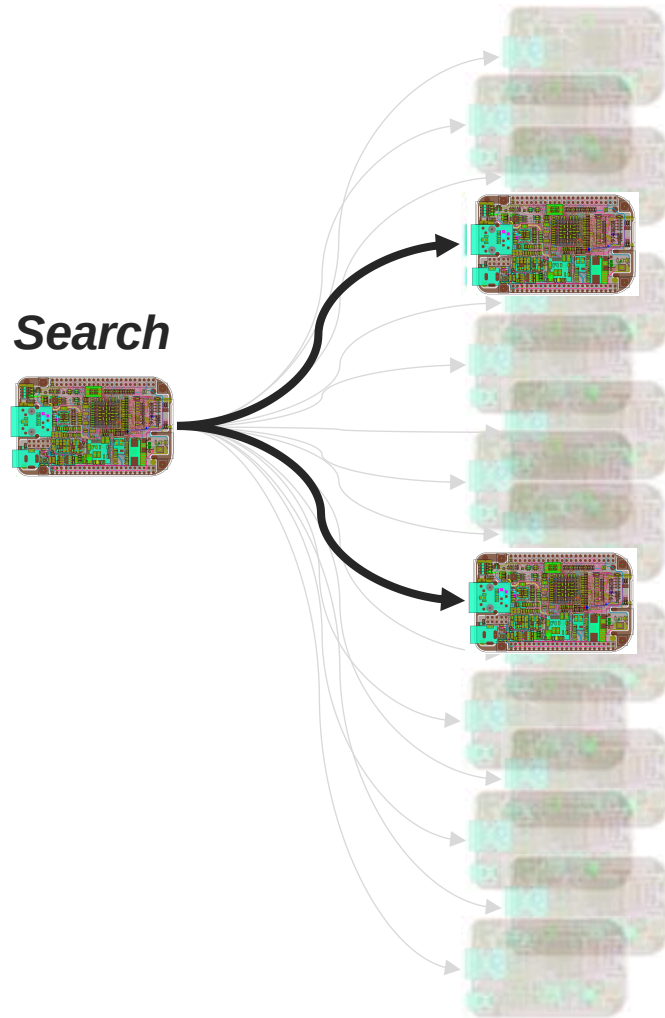
Monte Carlo Tree Search

Search



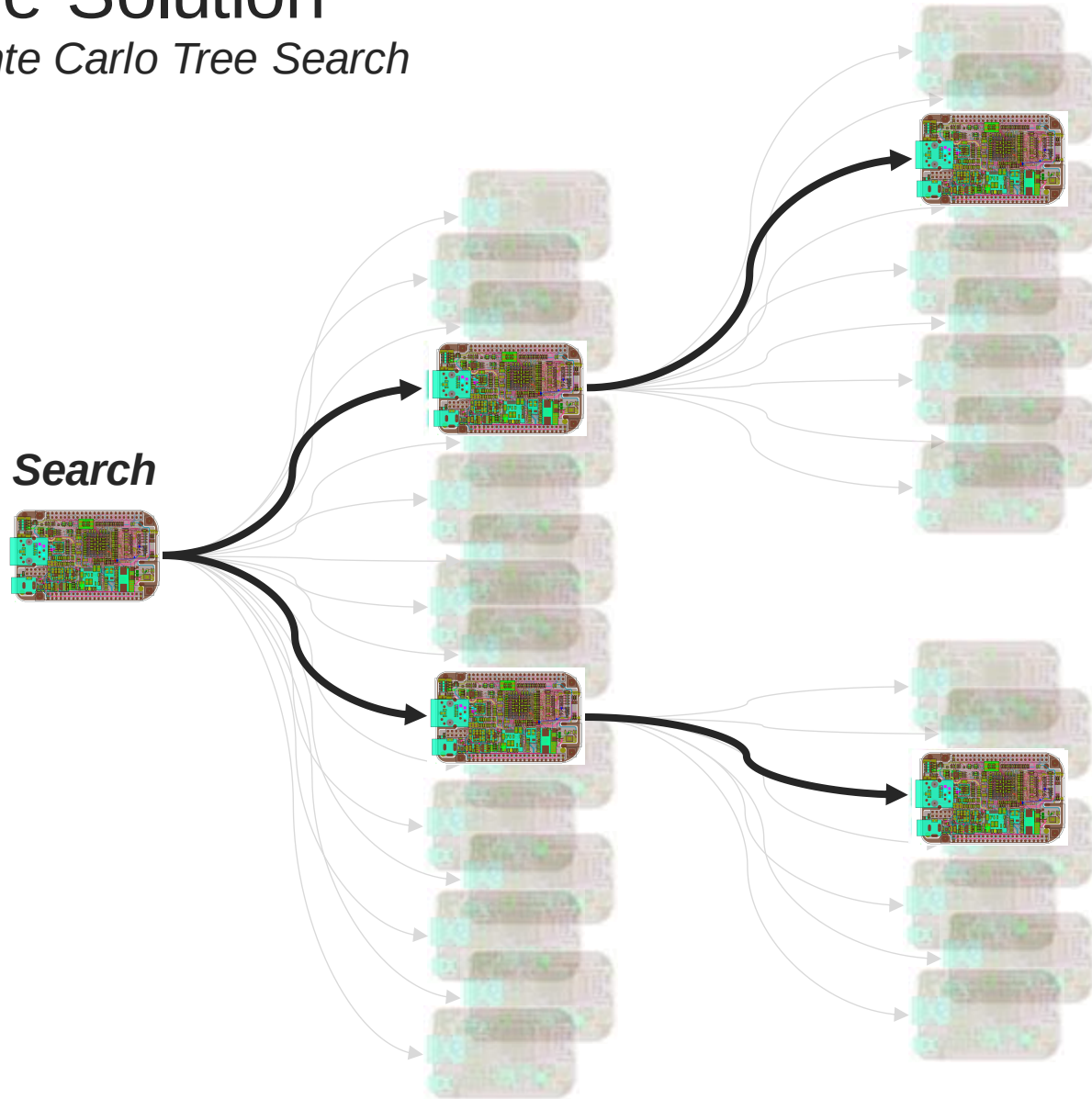
The Solution

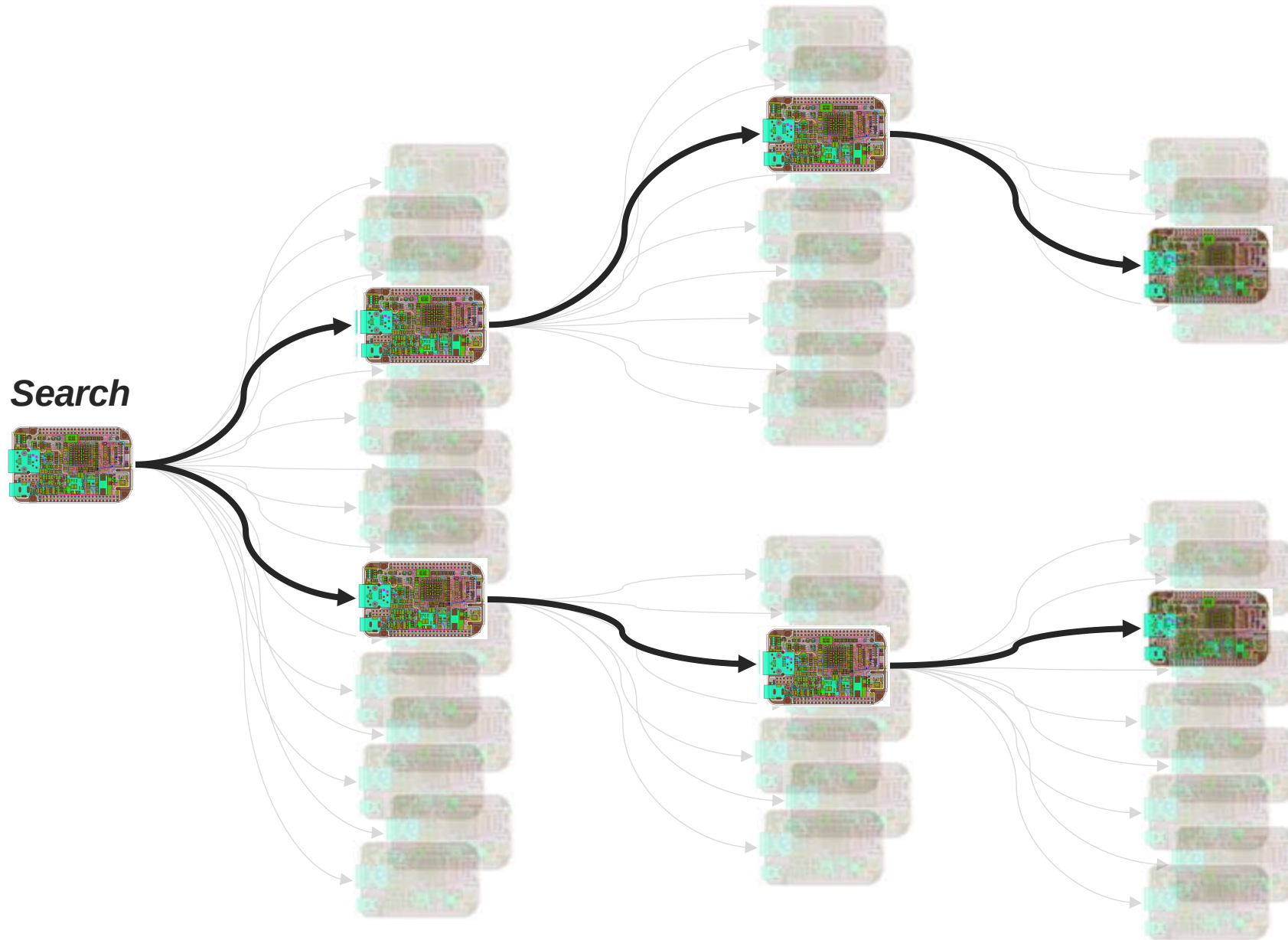
Monte Carlo Tree Search



The Solution

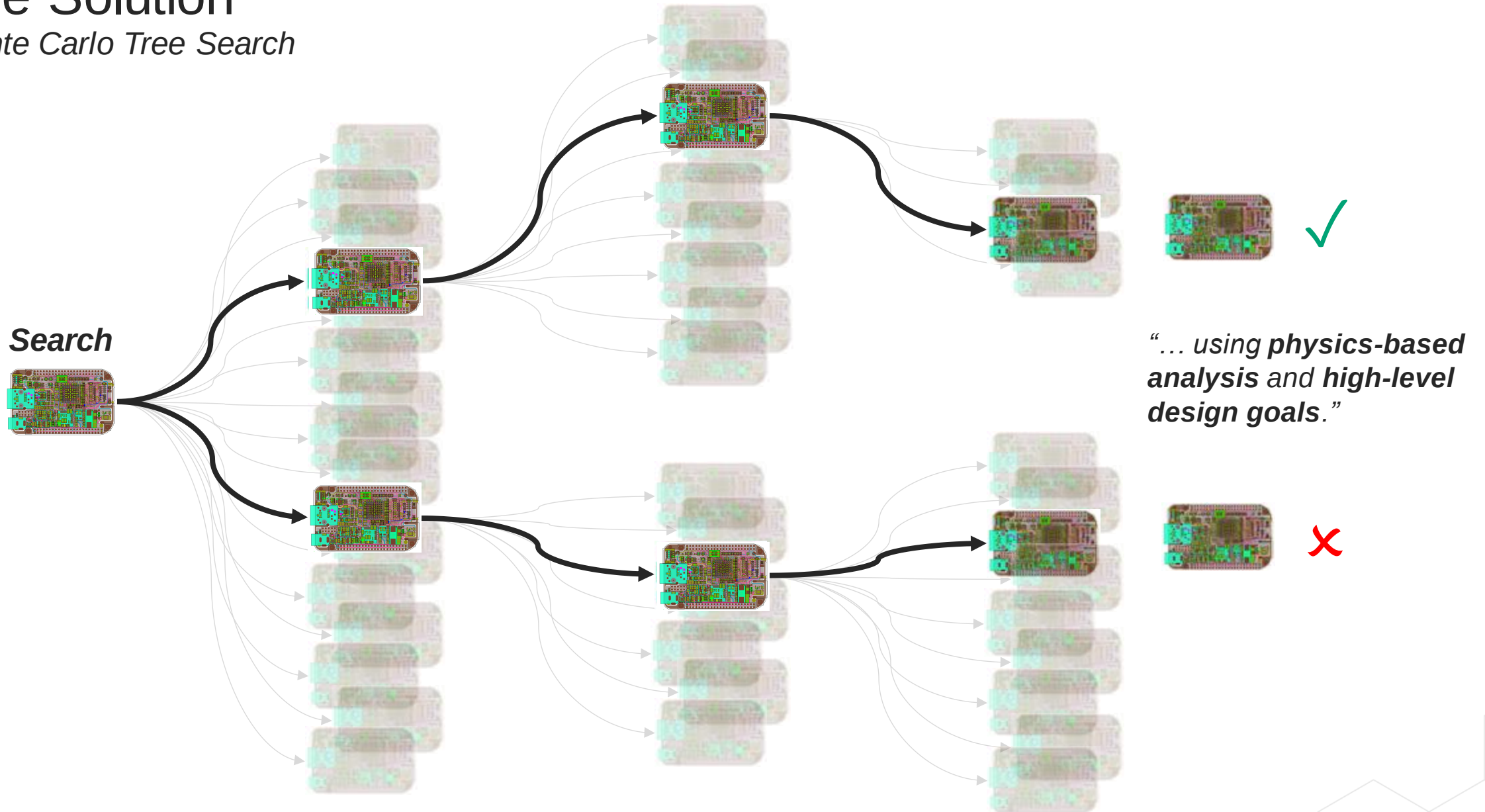
Monte Carlo Tree Search





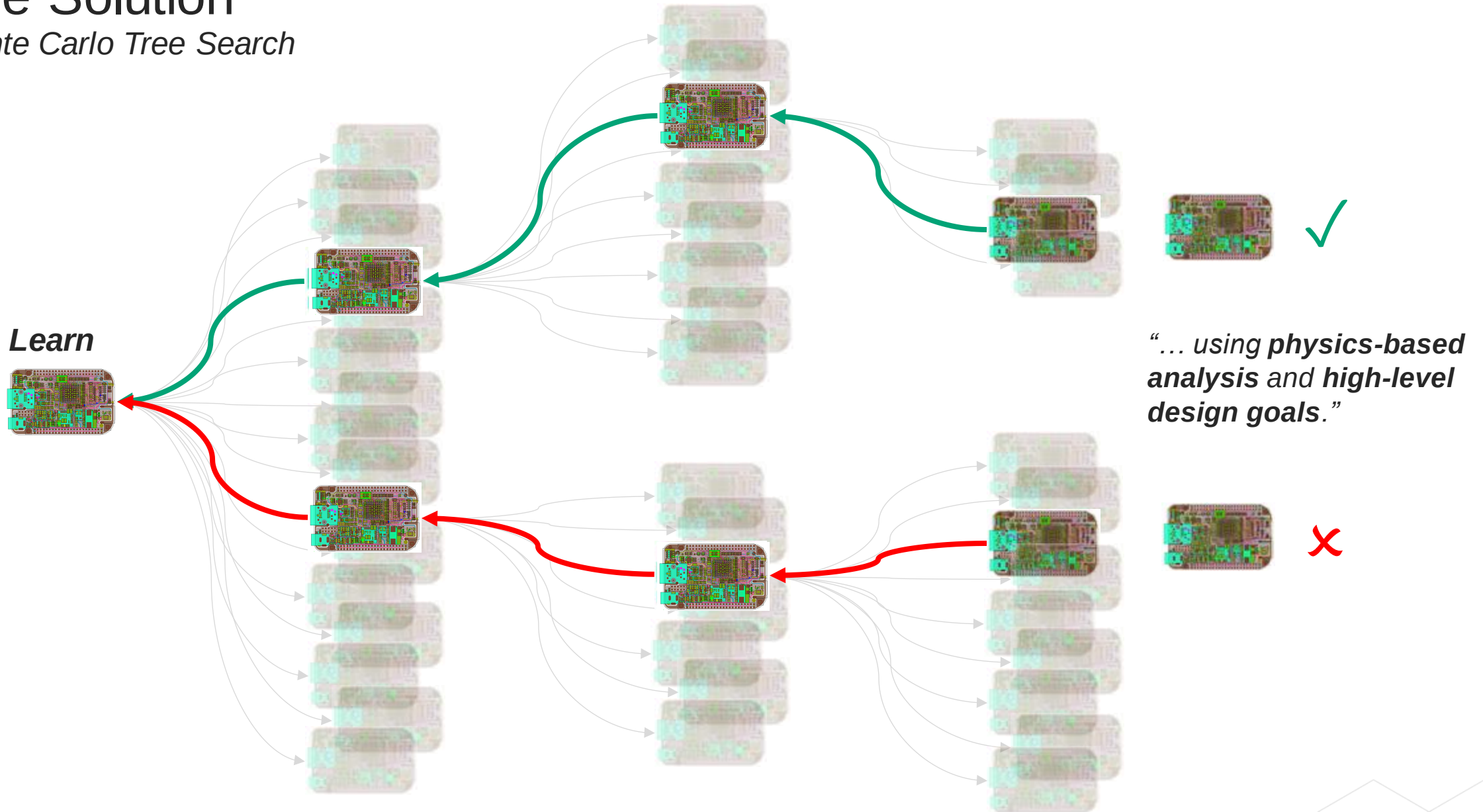
The Solution

Monte Carlo Tree Search



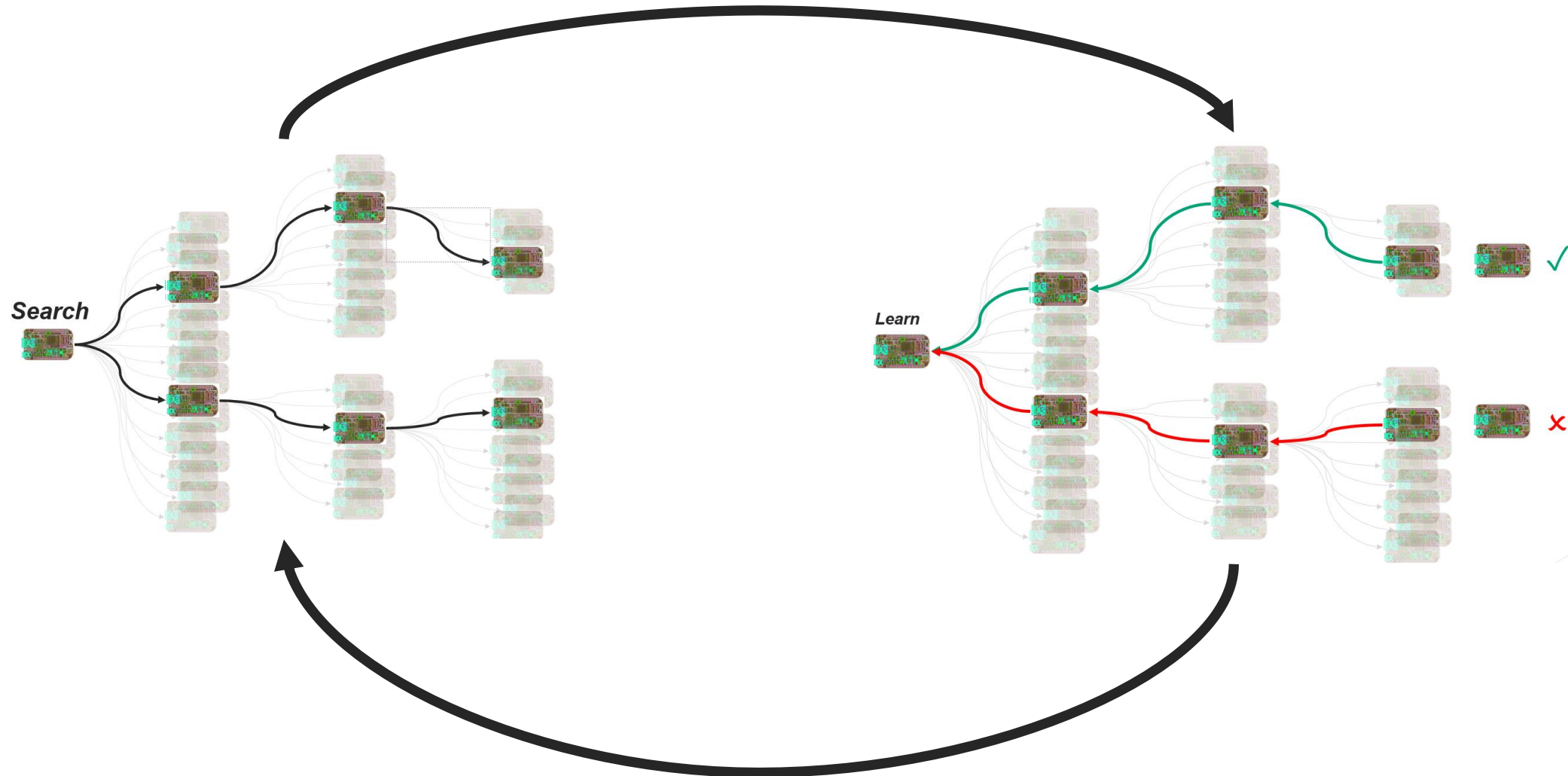
The Solution

Monte Carlo Tree Search



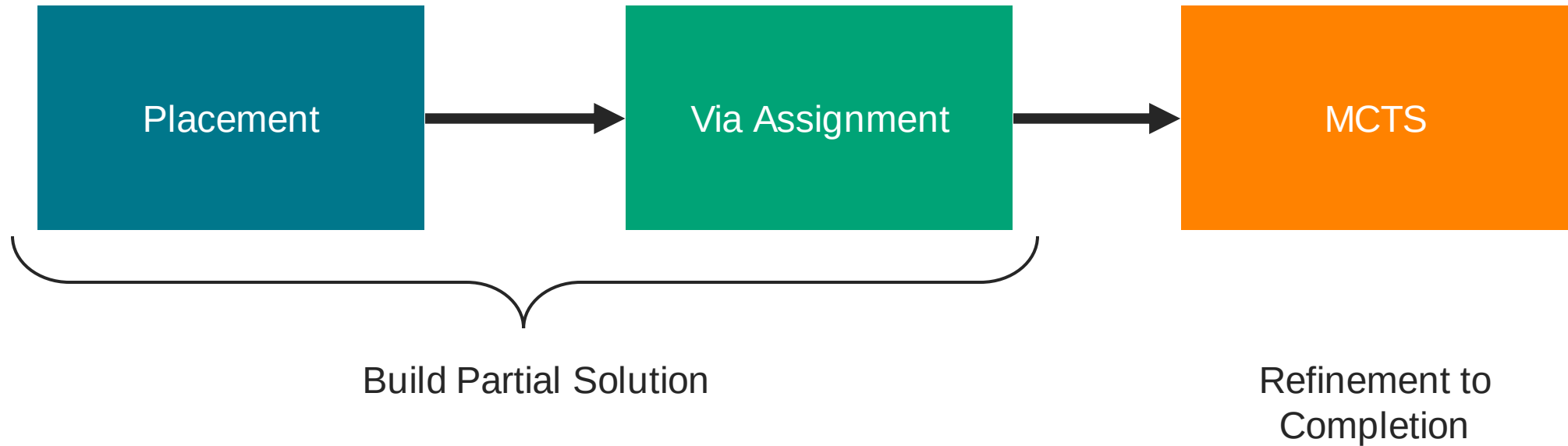
MCTS for PCB Design

Feedback Loop



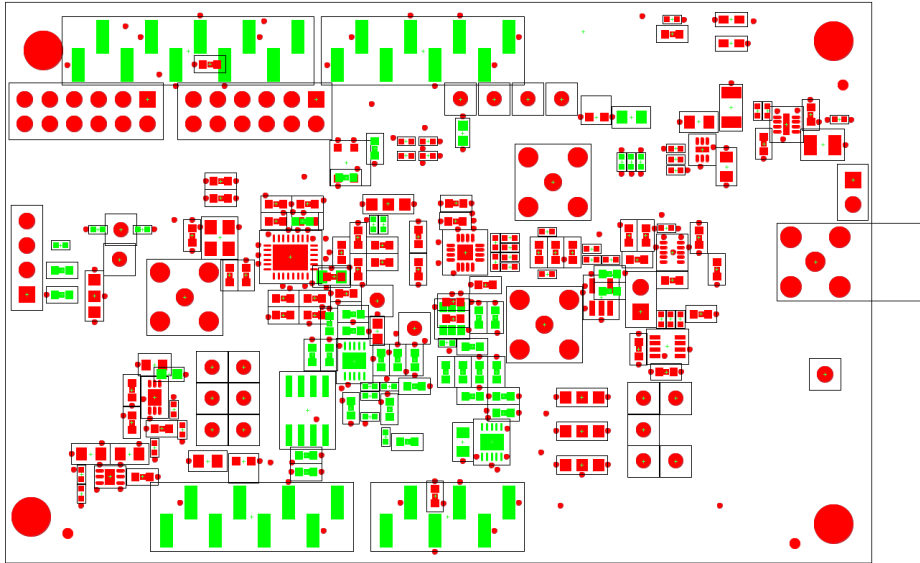
MCTS In Practice

Refinement of Partial Boards

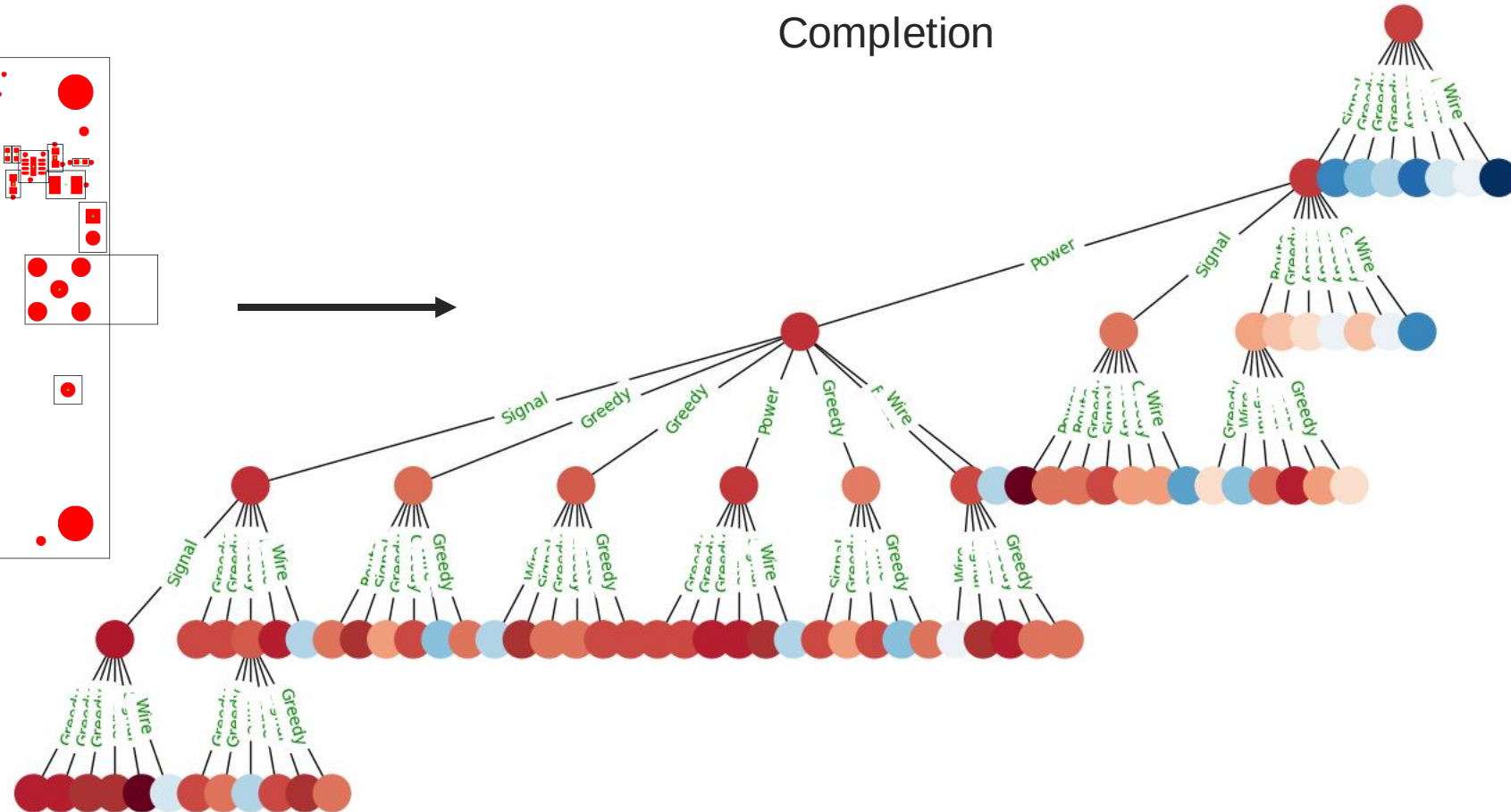


MCTS In Practice

Build Partial Solution



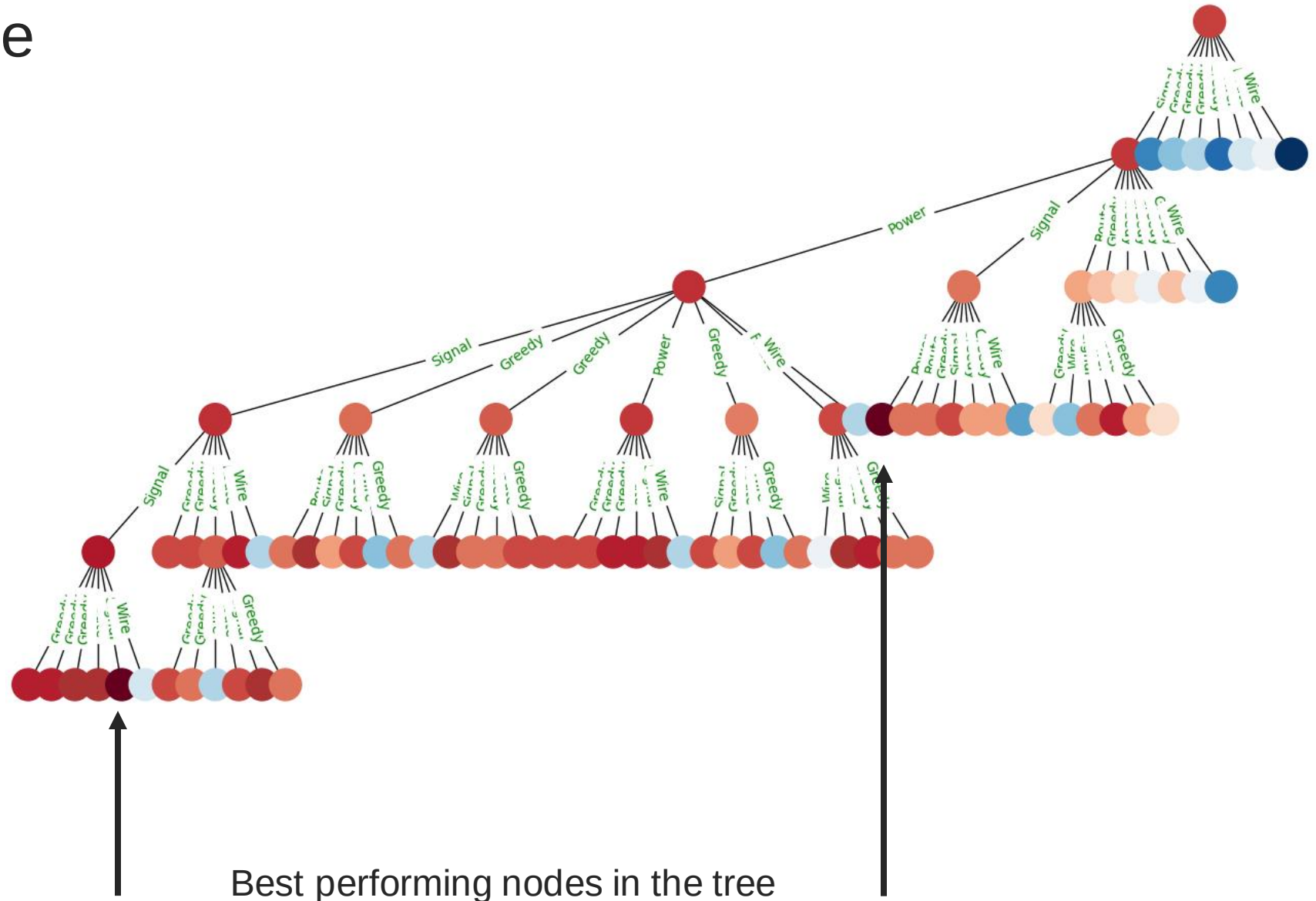
Refinement to Completion



MCTS In Practice

Search Tree

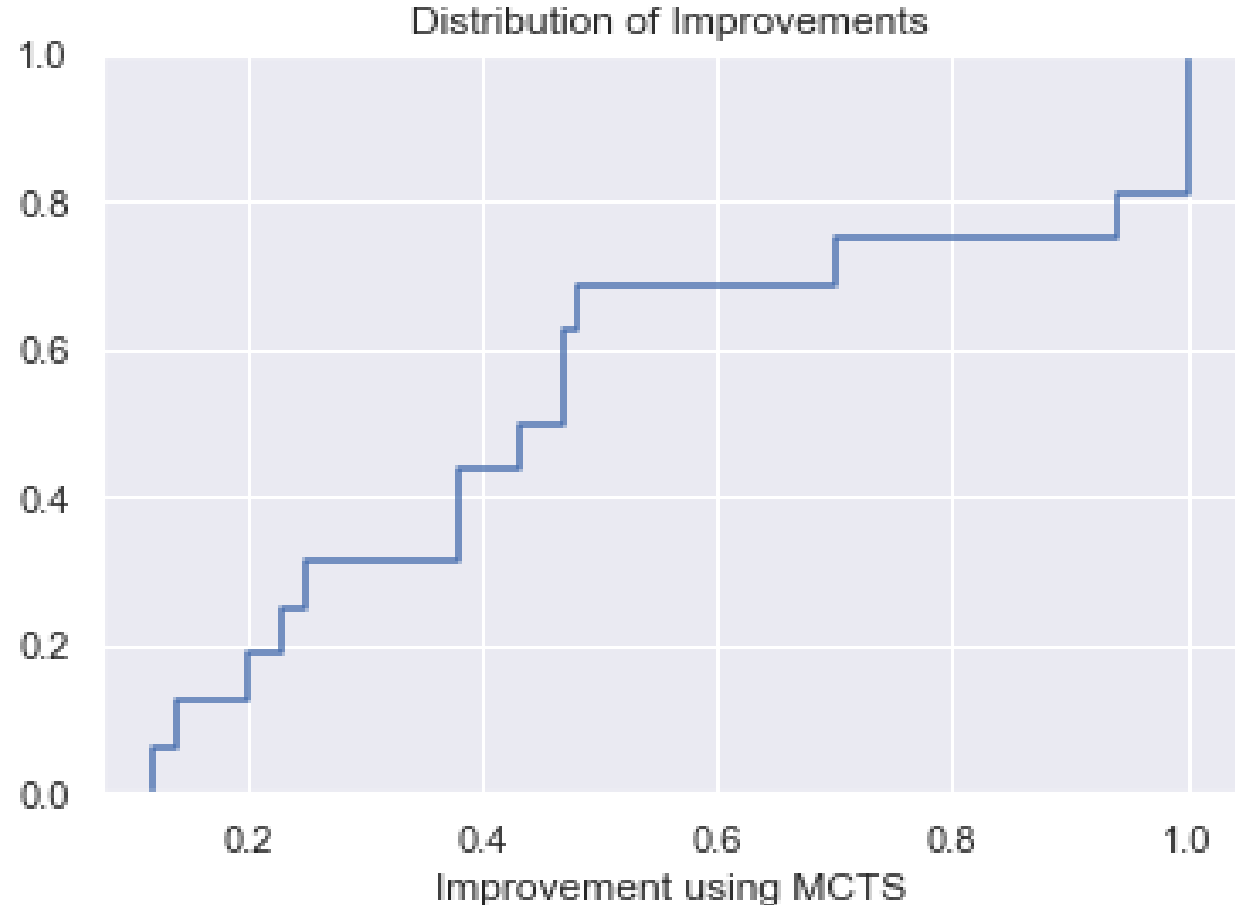
- MCTS adaptively chooses which search directions are worth pursuing
- It might find multiple solutions that are the same value



MCTS In Practice

Benchmark Results

- Suite of approximately 30 focus designs
- On average a design gets over 50% improvement over just routing
- Some designs can only reach route completion with MCTS-based feedback
- Next steps are to identify a more expressive actions space to help bring all designs to completion

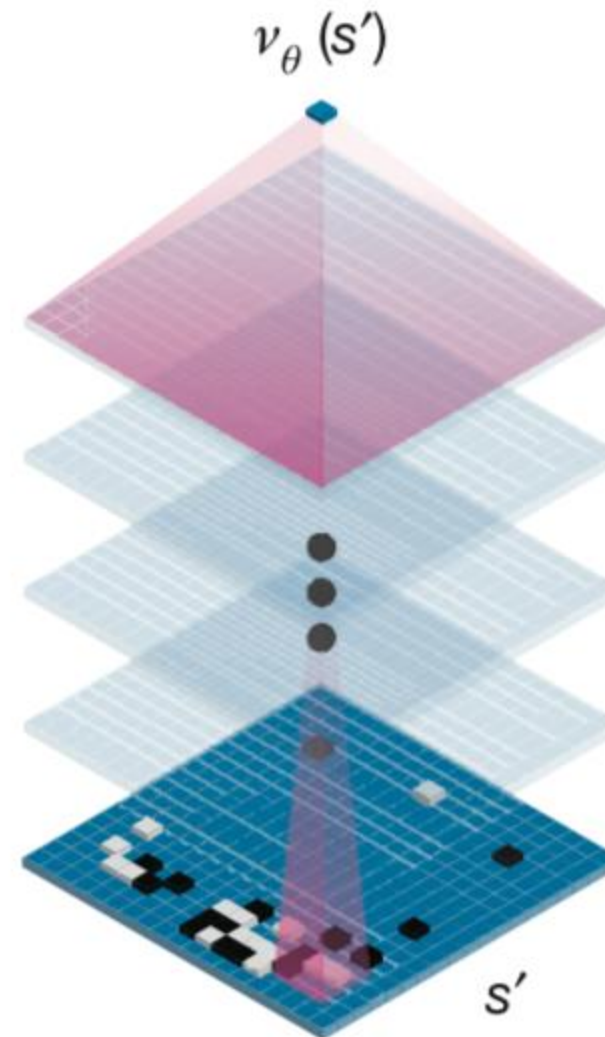
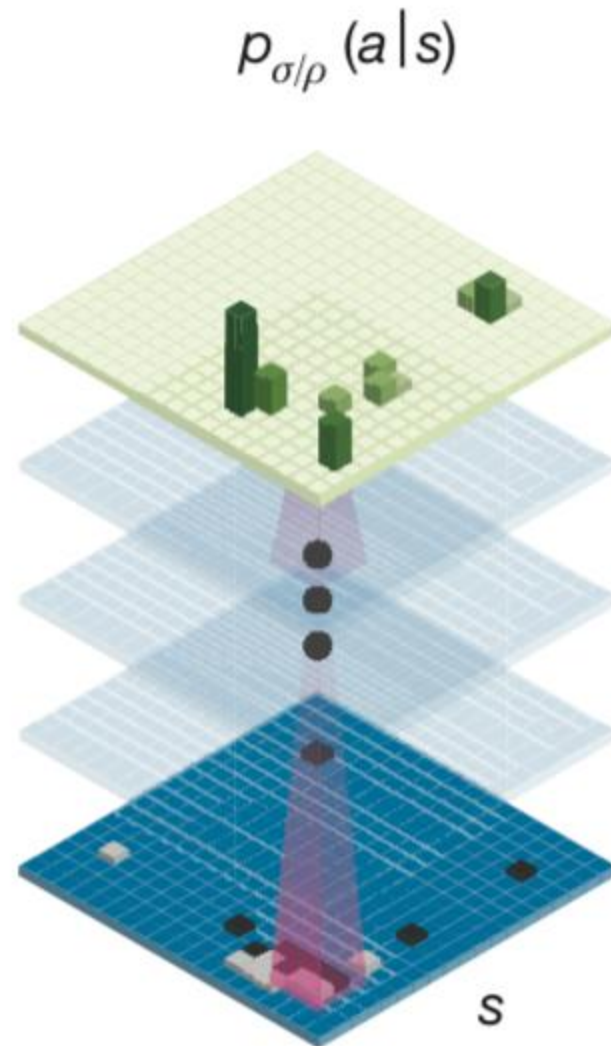


AlphaGO

Learning to search

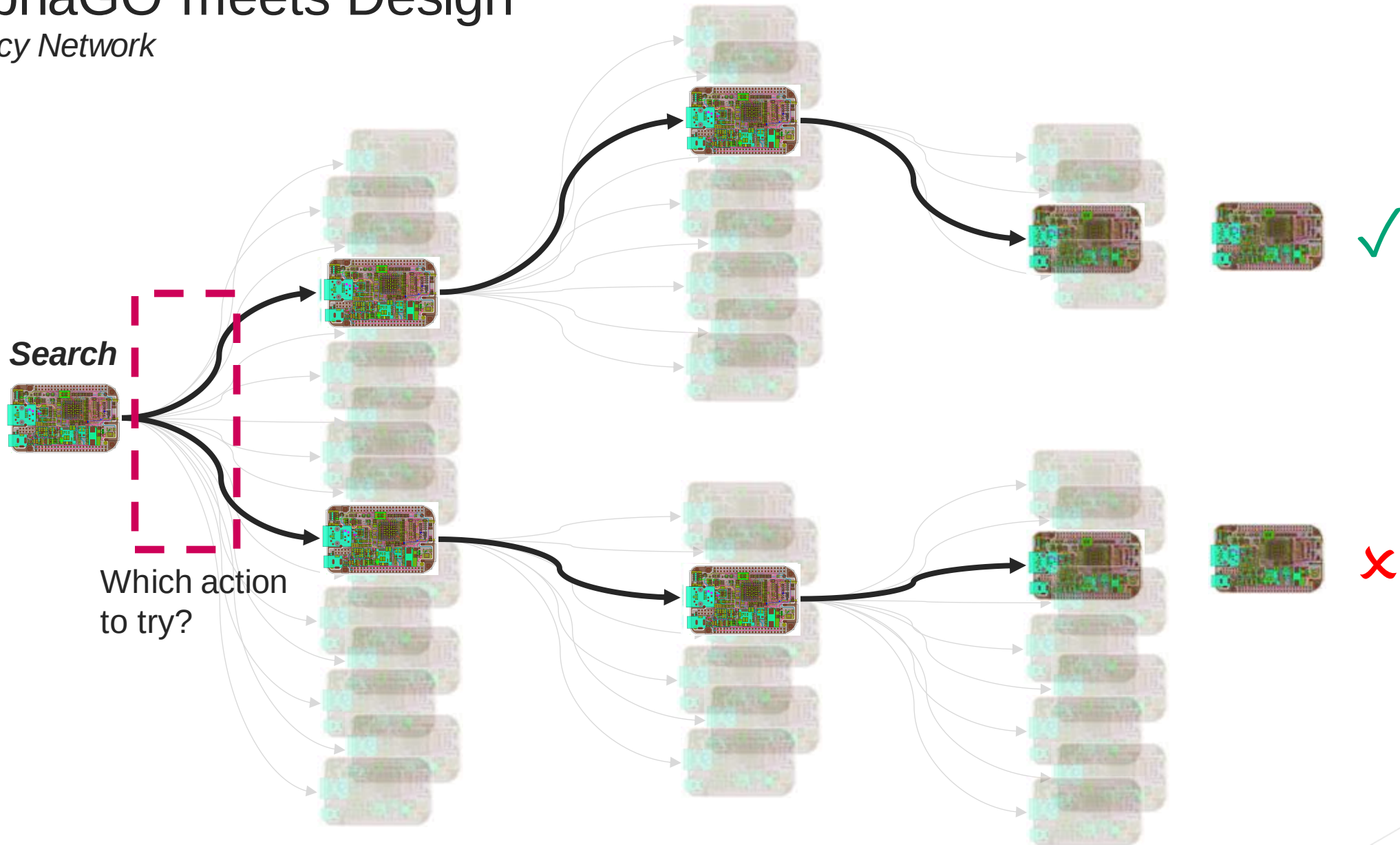
Policy network

Value network



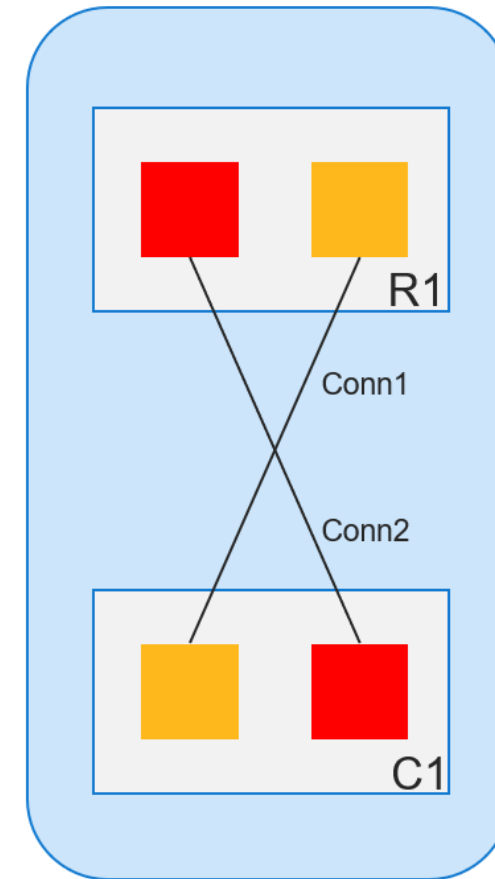
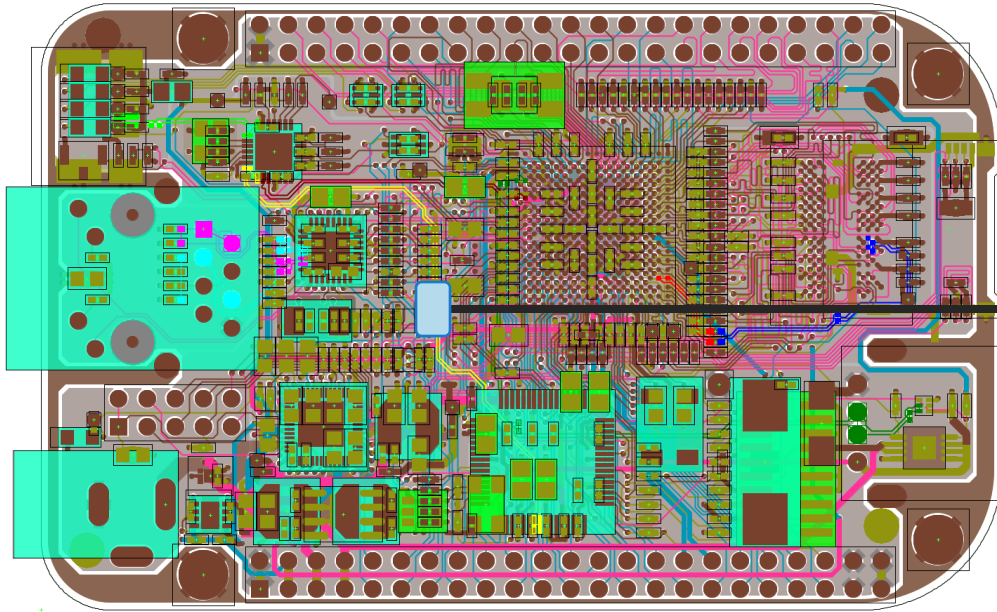
AlphaGO meets Design

Policy Network



MCTS for PCB Design

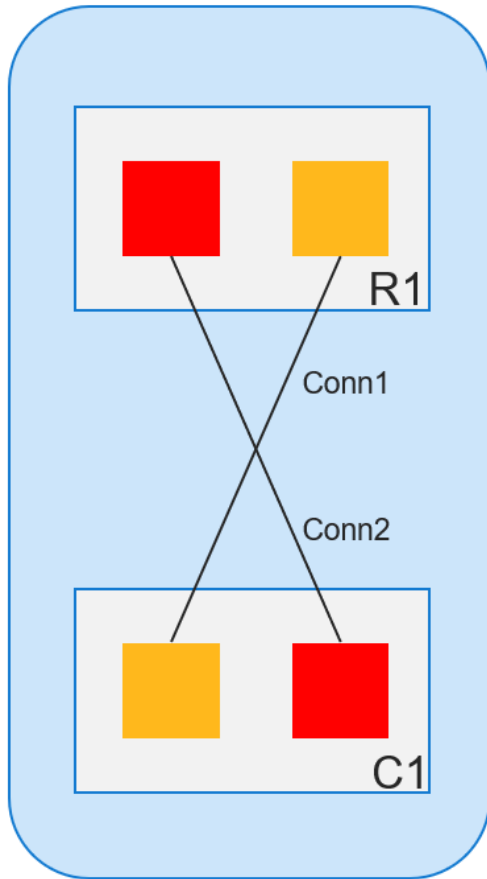
Policy Network



State

MCTS for PCB Design

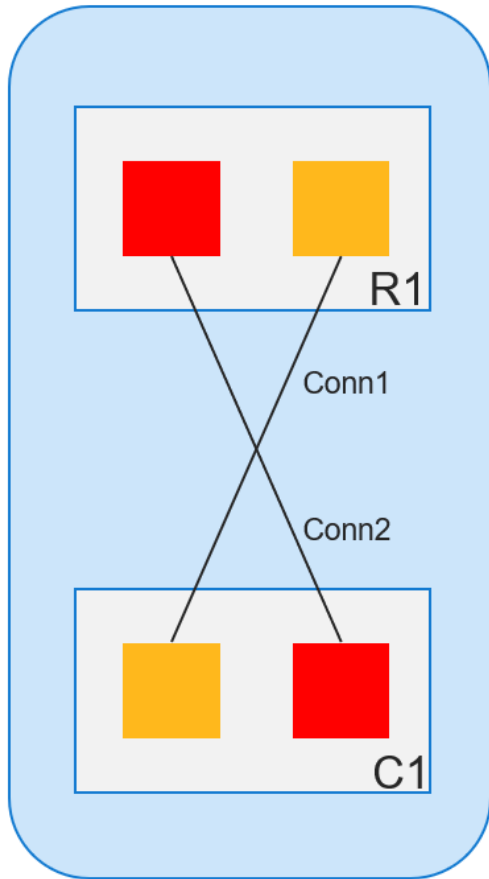
Policy Network



State

MCTS for PCB Design

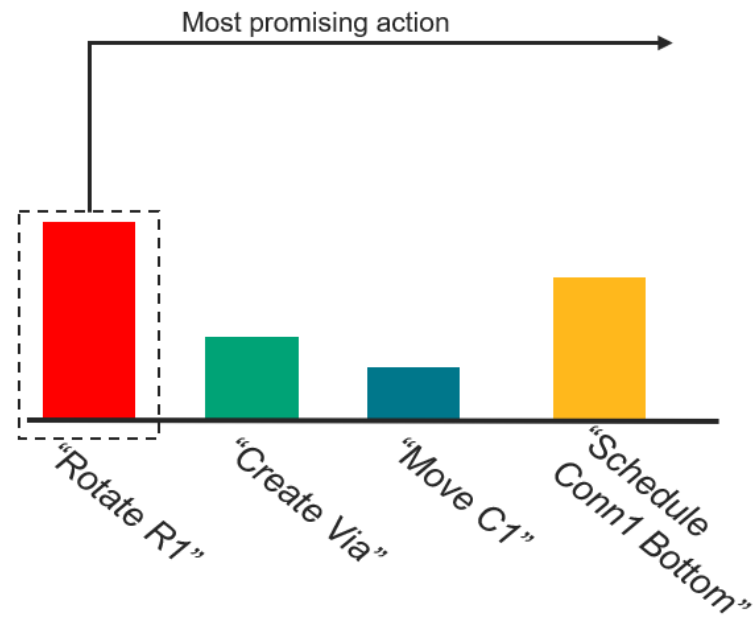
Policy Network



State



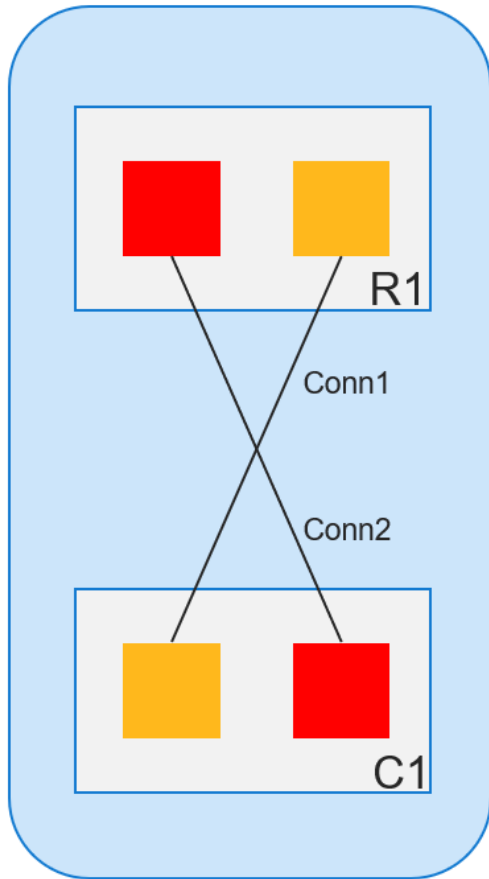
Predict Value



Select Action

MCTS for PCB Design

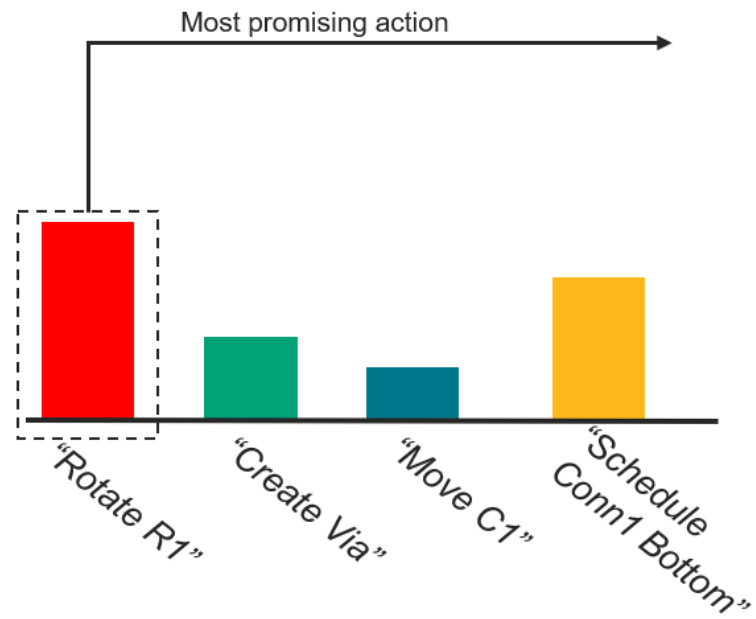
Policy Network



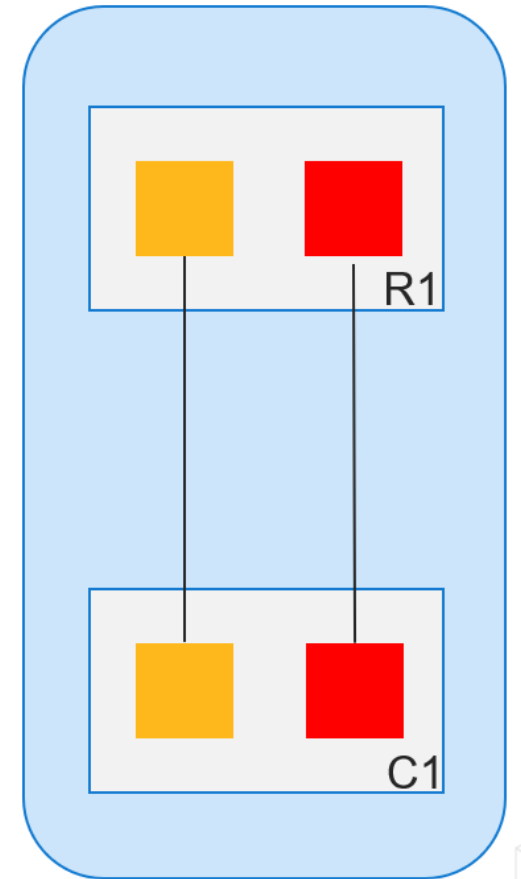
State



Predict Value



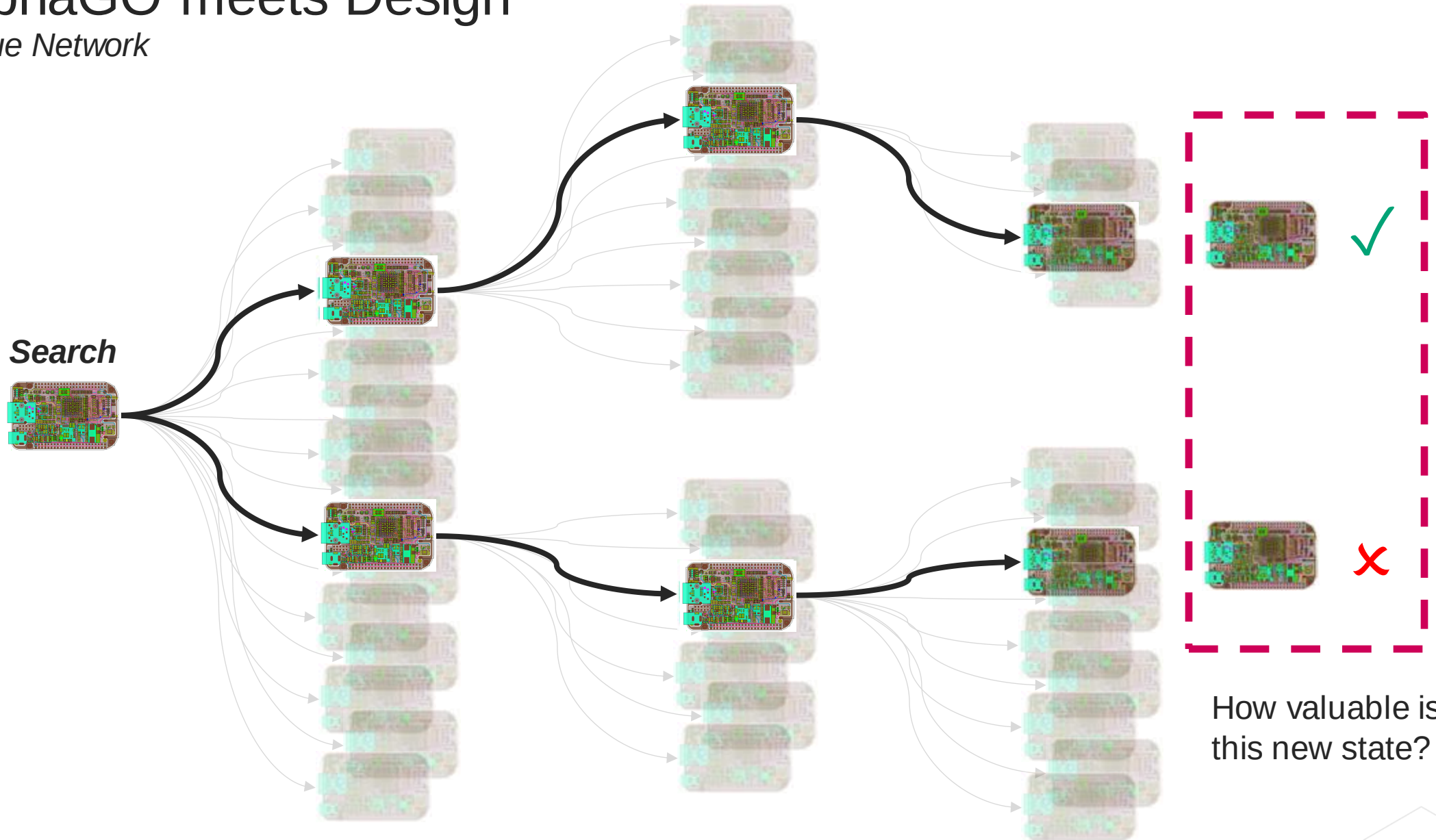
Select Action



Update

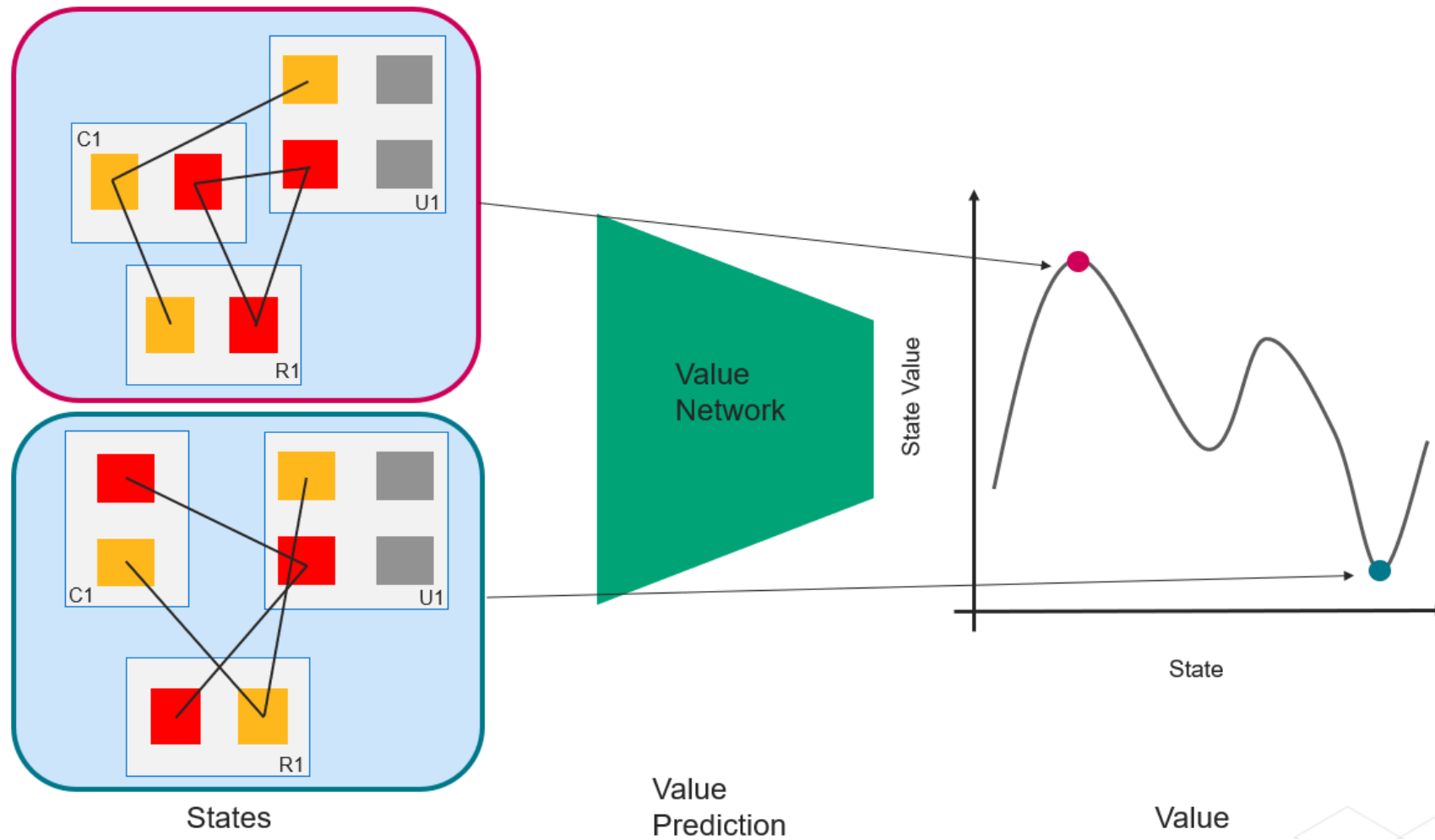
AlphaGO meets Design

Value Network



MCTS for PCB Design

Value Network



What problems can machine learning solve?

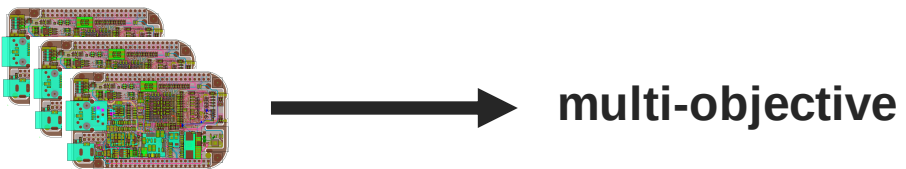
- *ML for prediction problems*



- *ML for design problems*



- *ML at scale*



Lessons in Scaling Laws

Let's dream a little bit

DALL·E 2 can create original, realistic images and art from a text description. It can combine concepts, attributes, and styles.

TEXT DESCRIPTION

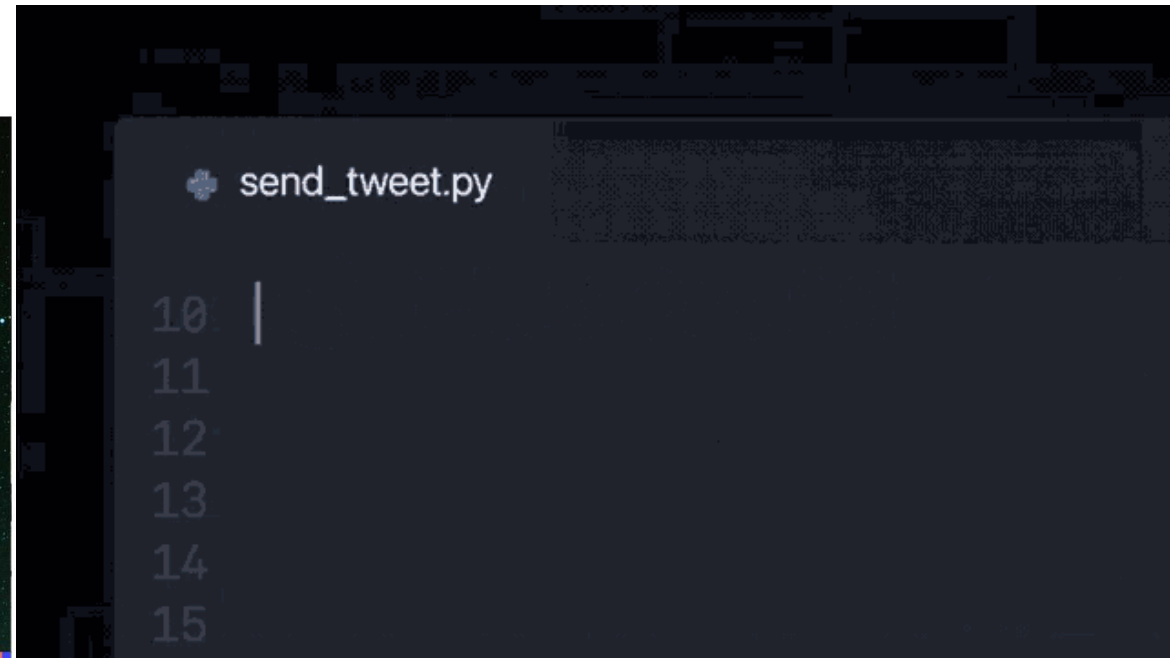
An astronaut Teddy bears A bowl of soup

riding a horse lounging in a tropical resort in space playing basketball with cats in space

in a photorealistic style in the style of Andy Warhol as a pencil drawing



DALL·E 2



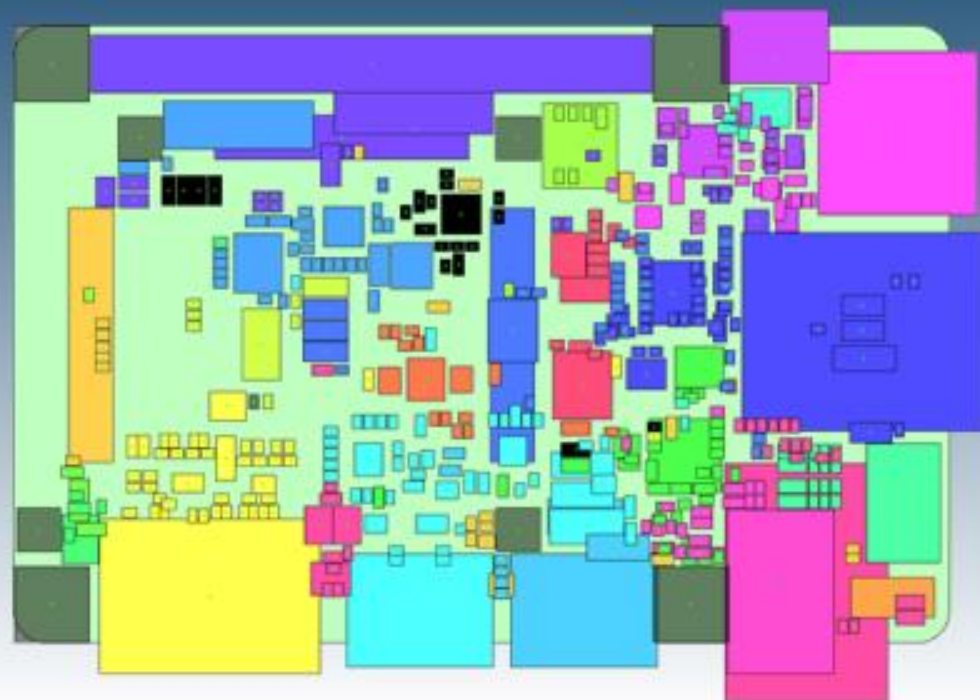
<https://openai.com/dall-e-2/>

<https://dev.to/thenomadevel/w-hat-is-github-copilot-will-it-take-your-jobs-33p5>

X AI

Artificial Intelligence to Reduce Time to Design PCBs – Google Coral Board

Human Placement



~ 3 days Placement

Allegro X: AI Clustering



75 minutes
14% better wirelength
Incremental setup time 15 mins



cādence®

© 2022 Cadence Design Systems, Inc. All rights reserved worldwide. Cadence, the Cadence logo, and the other Cadence marks found at <https://www.cadence.com/go/trademarks> are trademarks or registered trademarks of Cadence Design Systems, Inc. Accellera and SystemC are trademarks of Accellera Systems Initiative Inc. All Arm products are registered trademarks or trademarks of Arm Limited (or its subsidiaries) in the US and/or elsewhere. All MIPI specifications are registered trademarks or service marks owned by MIPI Alliance. All PCI-SIG specifications are registered trademarks or trademarks of PCI-SIG. All other trademarks are the property of their respective owners.

