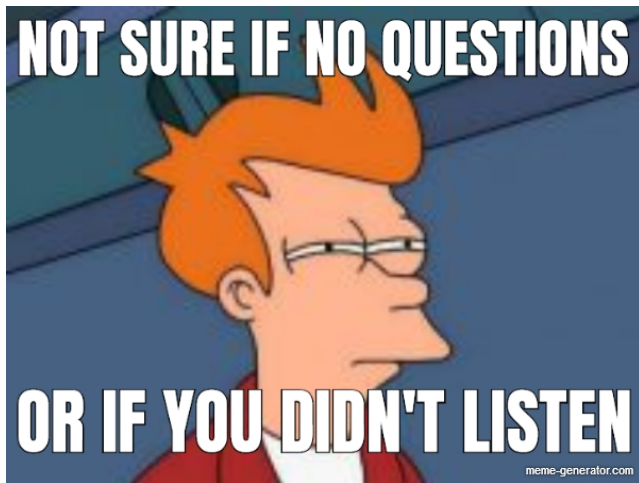


Neural planning

Michaela Urbanovská

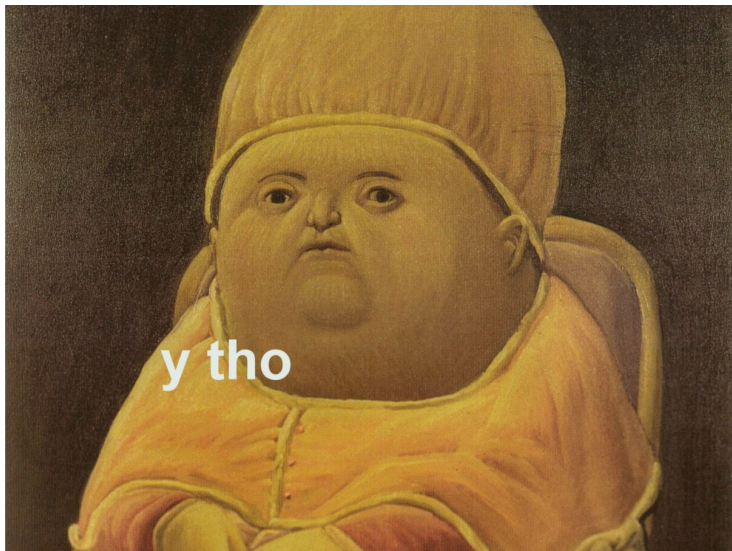
PUI Tutorial
Week 6

- Any questions regarding the lecture?



- What is that?

- What is that?
- That's just how I call it to be honest...
- Combining **Classical Planning** aaaaaaaand **Deep Learning**

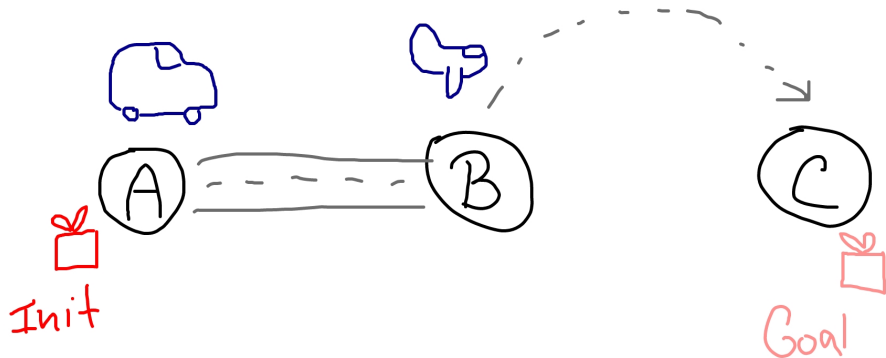




- Classical planning is great because
 - it can be domain-independent
 - you have powerful heuristics
 - there's a lot of state of the art
 - you have optimized planner implementations

- Classical planning is great because
 - it can be domain-independent
 - you have powerful heuristics
 - there's a lot of state of the art
 - you have optimized planner implementations
- ...but it's less great because
 - you **need a model**
 - you often have to hand-encode the model
 - different planner may need different representation
 - you need languages like PDDL

Neural planning



Neural planning

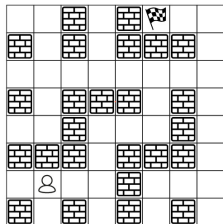


- Neural networks are great
 - when you have a lot of data
 - because they work on unstructured data
 - because they don't need a model!
- ...but they're less great
 - when you have a difficult task
 - when you don't have data
 - because they don't learn algorithms

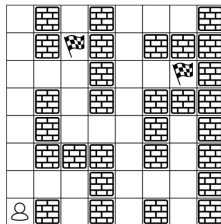
- What if... we used them both?
- Classical planning: algorithms
- Neural networks: replace parts of algorithms that require the costly model
- Different approaches as well
 - learn policy function
 - learn PDDL from the graphic representation
 - ...many more directions

- State space + state-transition function
- **Expansion network** that works with graphic representation
- Heuristic function
- **Heuristic network** that works with graphic representation

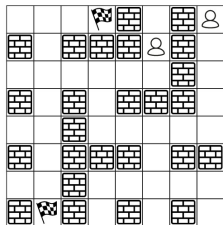
Neural planning - domains



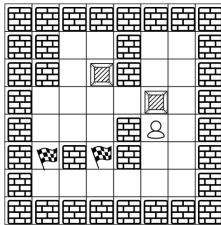
Single-agent maze



Multi-goal maze



Multi-agent maze


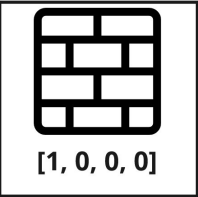




Sokoban

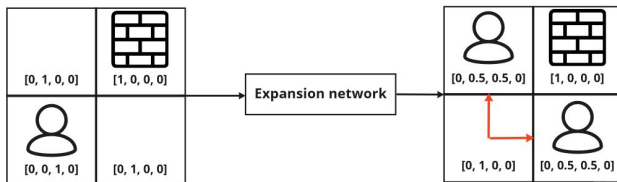
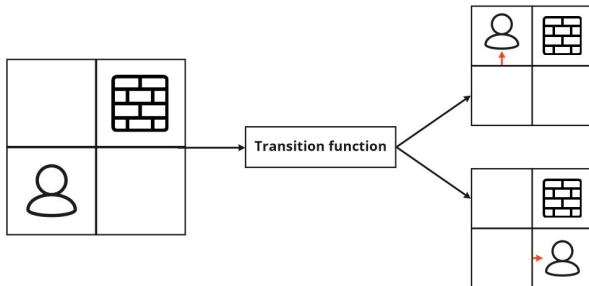
mira

Neural planning

- Grid domains (so far...)
- One-hot encoding of the entities on cells
- Convolutional + recurrent neural networks
- **Scale-free** architectures

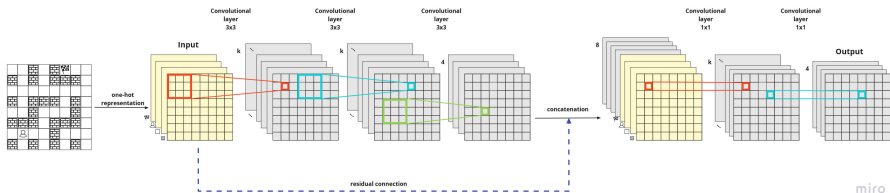
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Expansion network



Expansion network

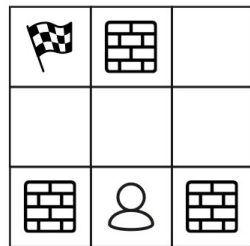
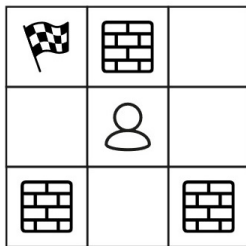
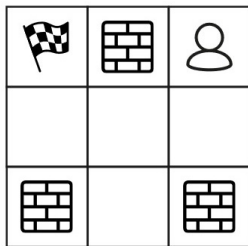
- Convolutional neural network
- 4-neighborhood movement possible
- 3×3 convolutional window to see the surroundings of the agent
- residual connection in the architecture to not lose initial information



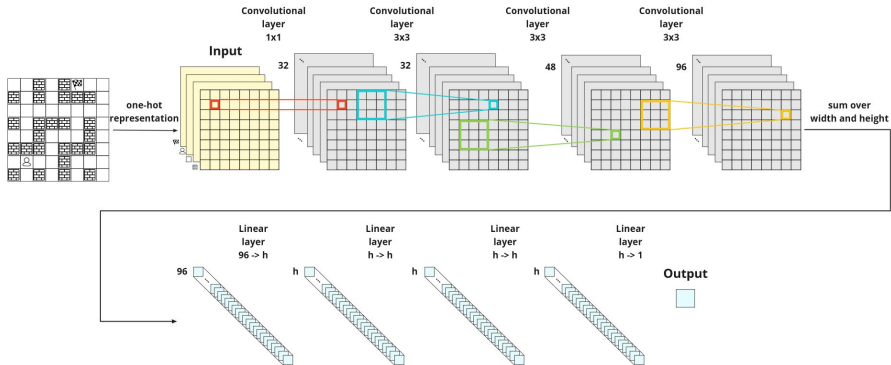
- 3 architectures
- **CNN** - convolutional neural network (most simple)
- **CNN_att** - convolutional neural network using attention
- **RNN** - reasoning recurrent network using MAC cell
- Inspiration in landmarks, relaxations, abstractions...
- Each architecture has intuition

Heuristic network - loss function

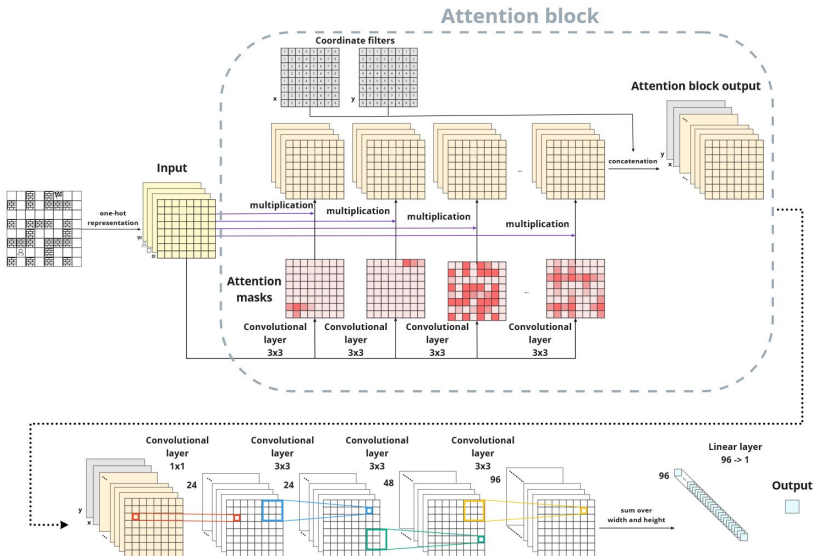
- We're learning **monotonicity**
- Property of any good heuristic
- sample + label pairs aren't enough anymore
- one instance with multiple agent placements



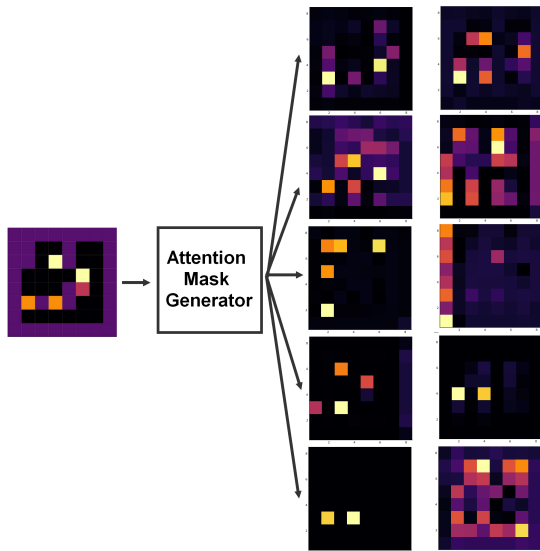
Heuristic network - CNN



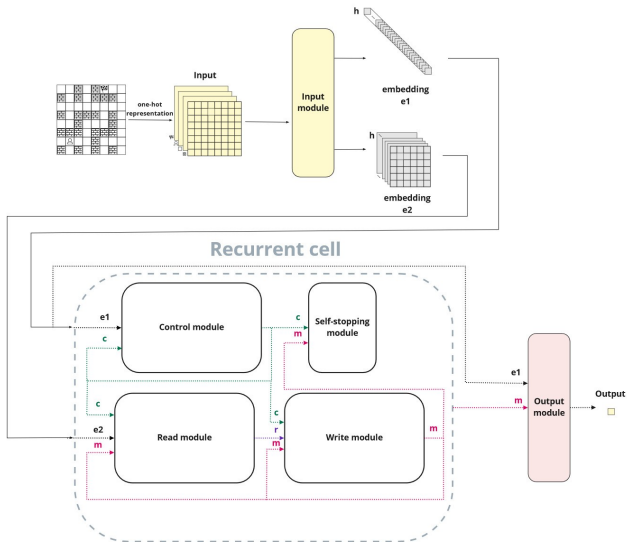
Heuristic network - CNN_att



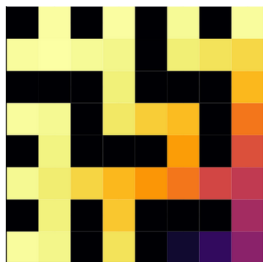
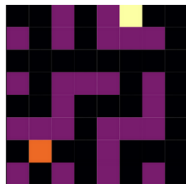
Heuristic network - CNN_att Sokoban attention masks



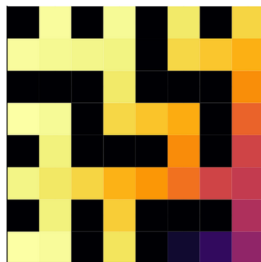
Heuristic network - RNN



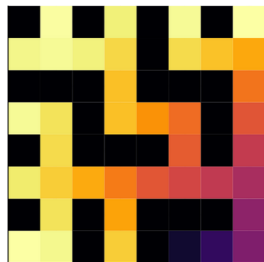
Architecture comparison - 8×8



CNN

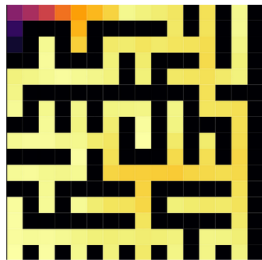
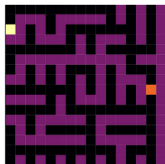


CNN att

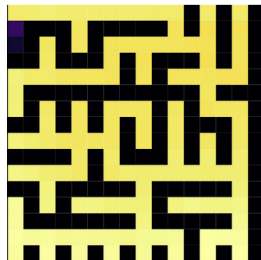


RNN

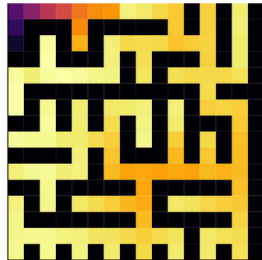
Architecture comparison - 16×16



CNN

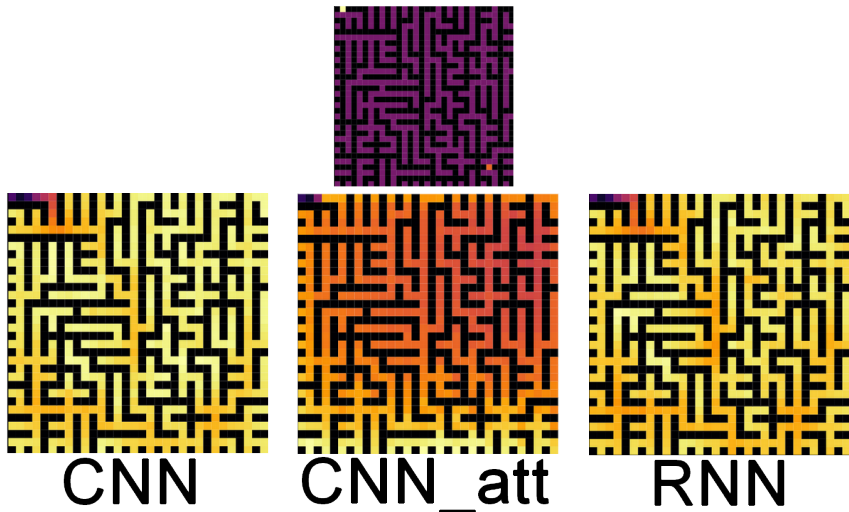


CNN_att

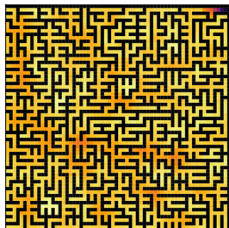


RNN

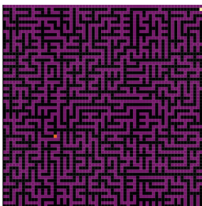
Architecture comparison - 32×32



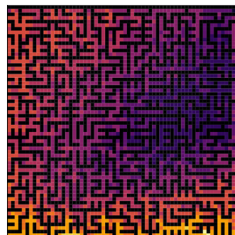
Architecture comparison - 64×64



CNN



CNN_att



RNN

- We tested all architectures against Euclidean distance, h^{LM-Cut} and h^{FF} heuristics
- Results on par on small domains
- Time advantage in large complex domains
- Less informed values in large state-spaces
- Slower expansion can slow down the search too much

- More domains!
- Bootstrapping
- Different approaches entirely?
- Explainability of the architectures

- Planning can be combined with other disciplines
- Neural networks aren't almighty
- Combination is the key!
- If you're interested in this direction (or any other planning direction) let us know!
 - Thesis topics
 - Possibly summer jobs
 - Other cooperation :)



[Feedback form link](#)

