

CS 598CM: ML for Compilers and Architecture

Instructor: Charith Mendis



Brief Announcements

- Recordings and Zoom
- **Reading List:** Live on the website!
- **Paper Selections:** Due on **September 7th**
- **Paper Reviews:** We will use hotCRP to facilitate review writing. Instructions to come soon!
- **Resources and tutorials:** Will be up by Friday

Recap

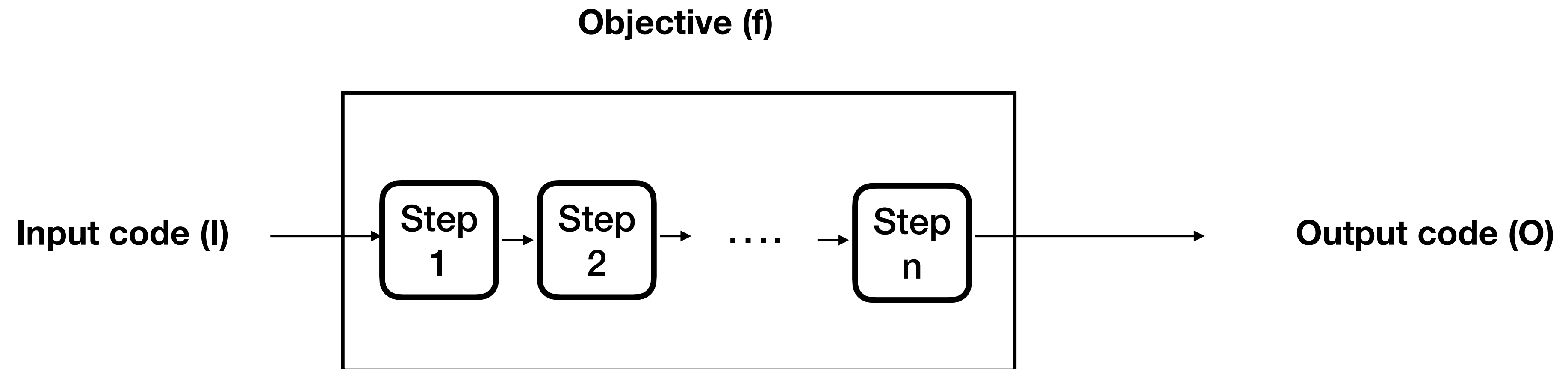
- Compiler Stages
 - Lexer => Parser => Sema => Optimization => Code Generation
- Two types of compiler optimizations
- Phase ordering problem

Lecture 3:

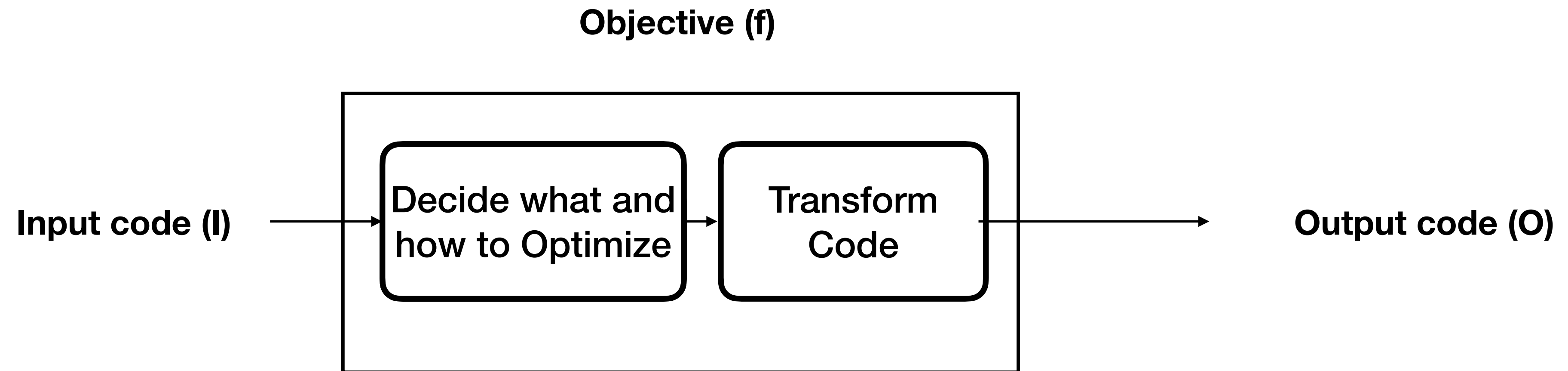
Compiler Optimizations

Optimizations + DSLs

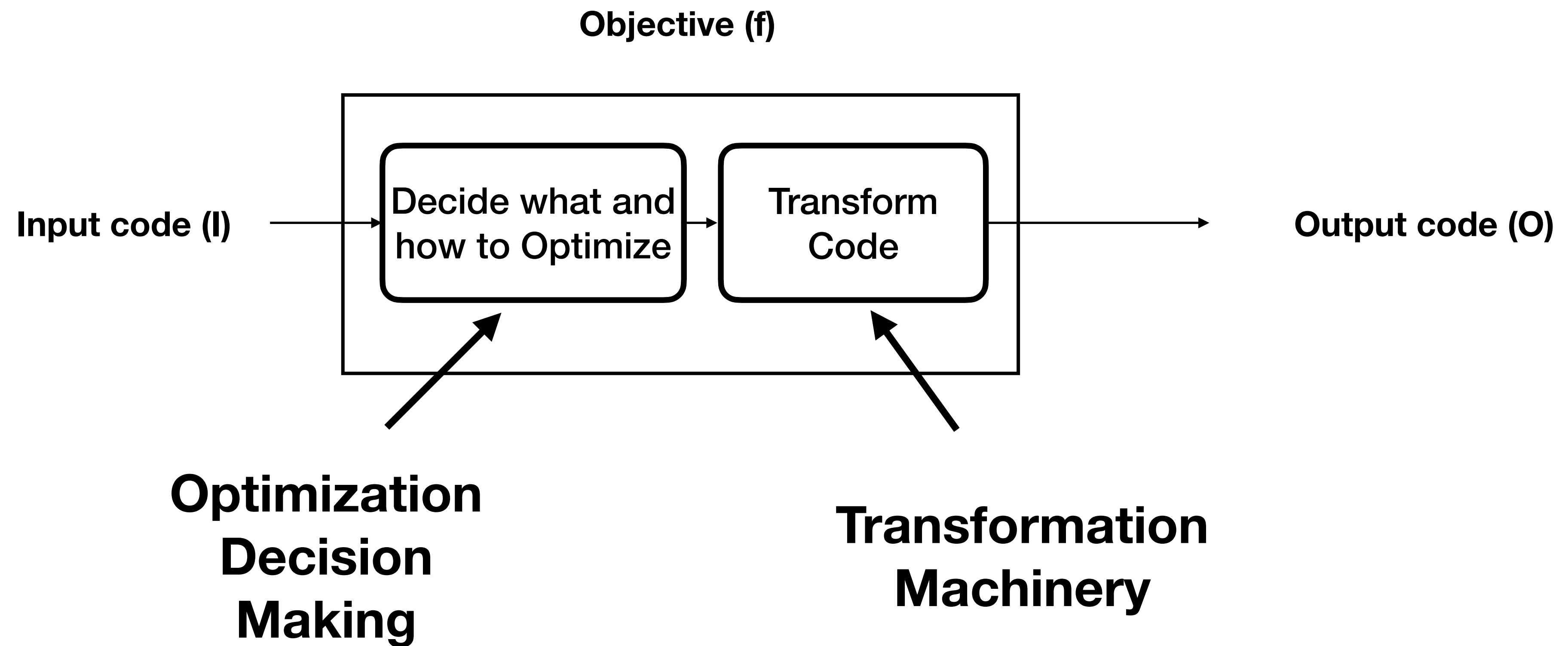
Anatomy of an Optimization Pass



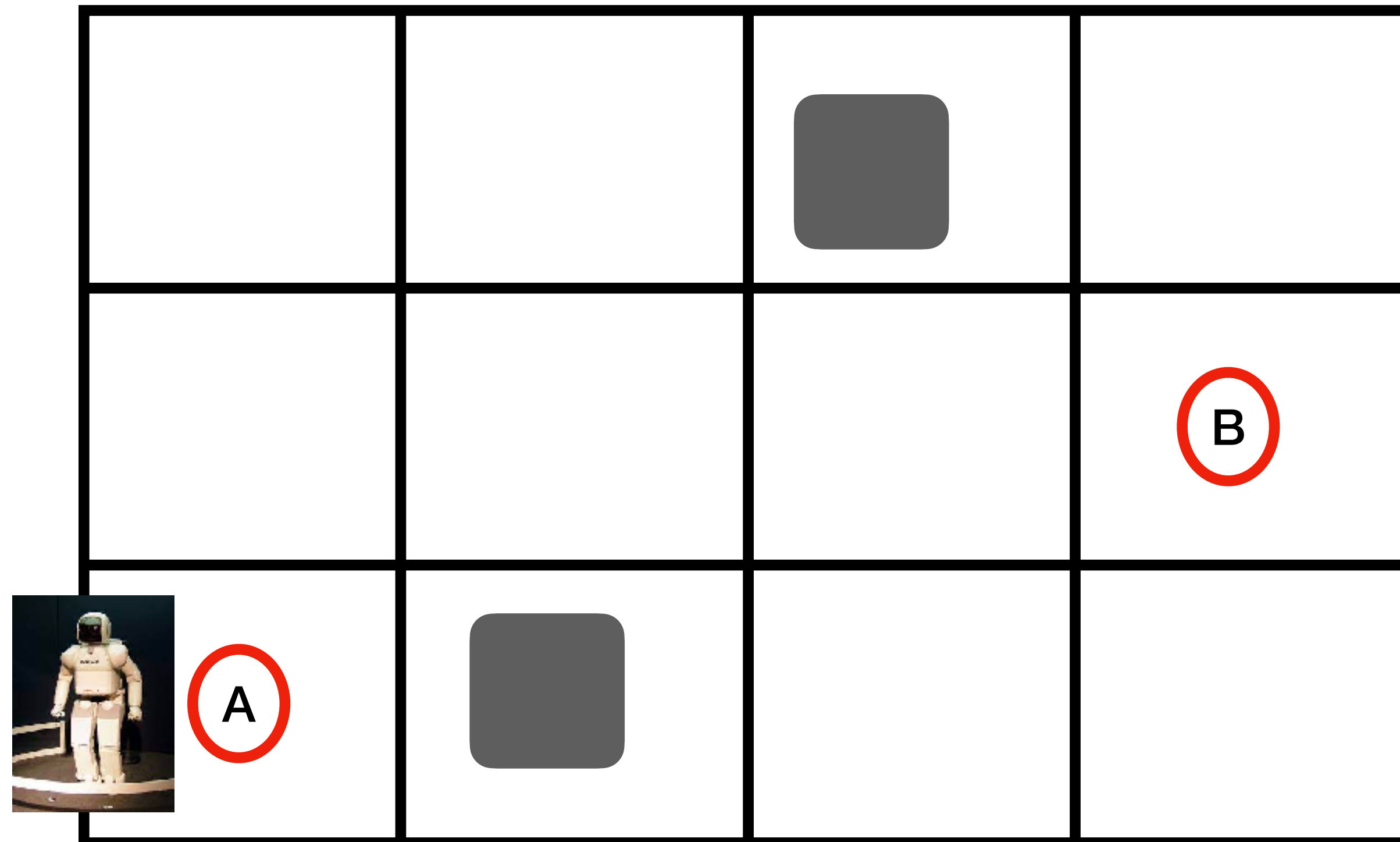
Anatomy of an Optimization Pass



Anatomy of an Optimization Pass

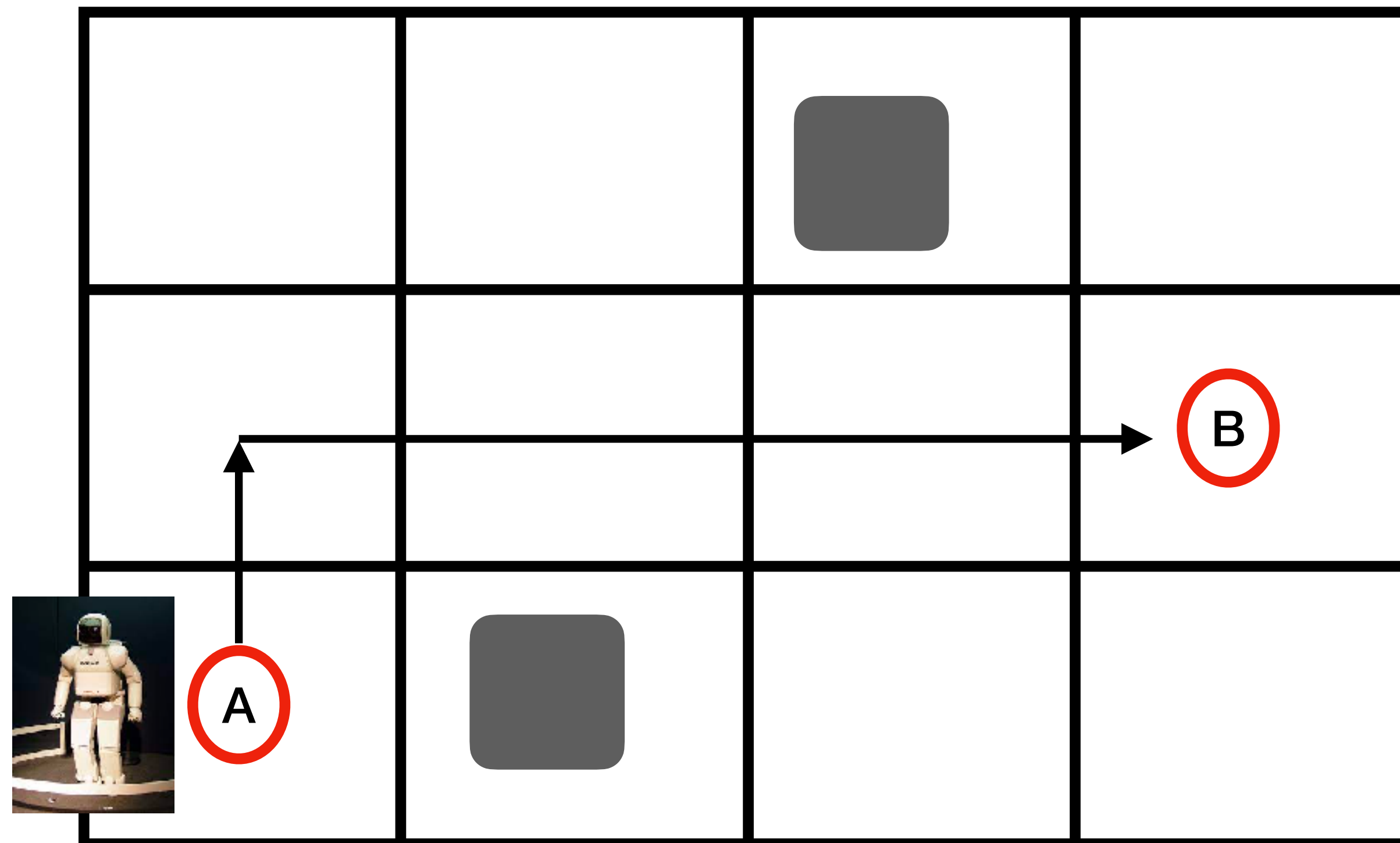


Robot Analogy



Task: Move from A to B cheaply

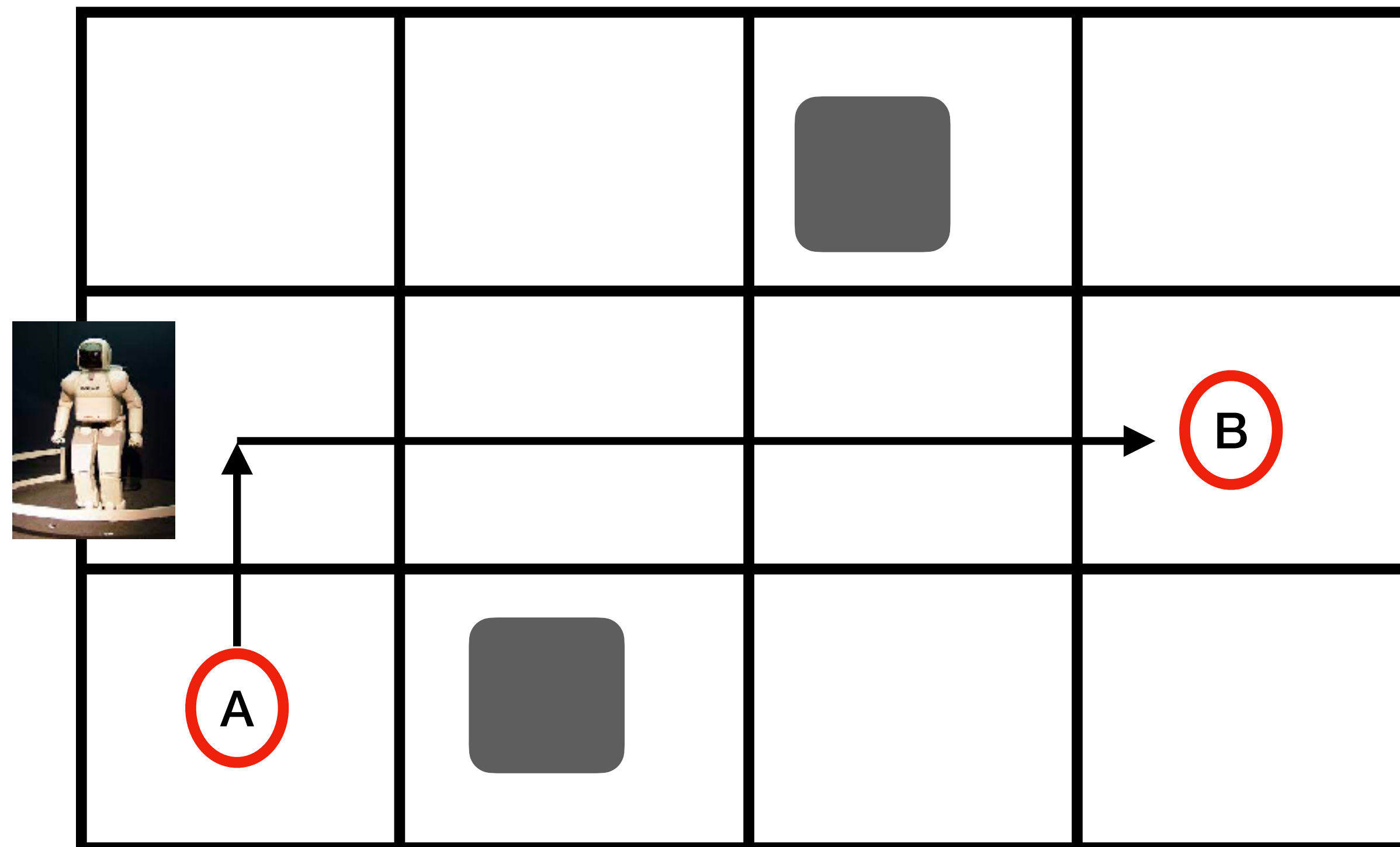
Robot Analogy



Task: Move from A to B cheaply

1. Plan

Robot Analogy

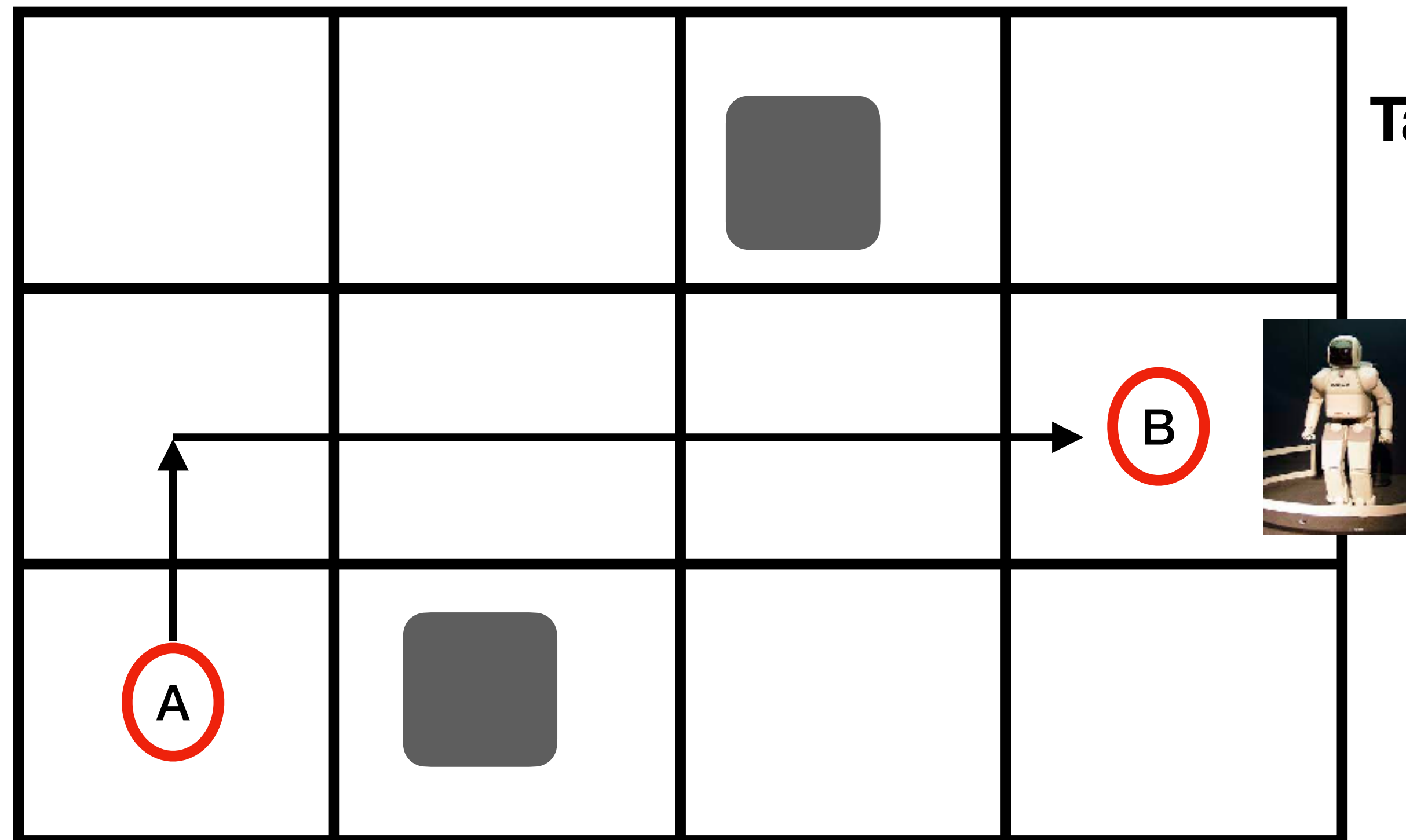


Task: Move from A to B cheaply

1. Plan

2. Execute

Robot Analogy

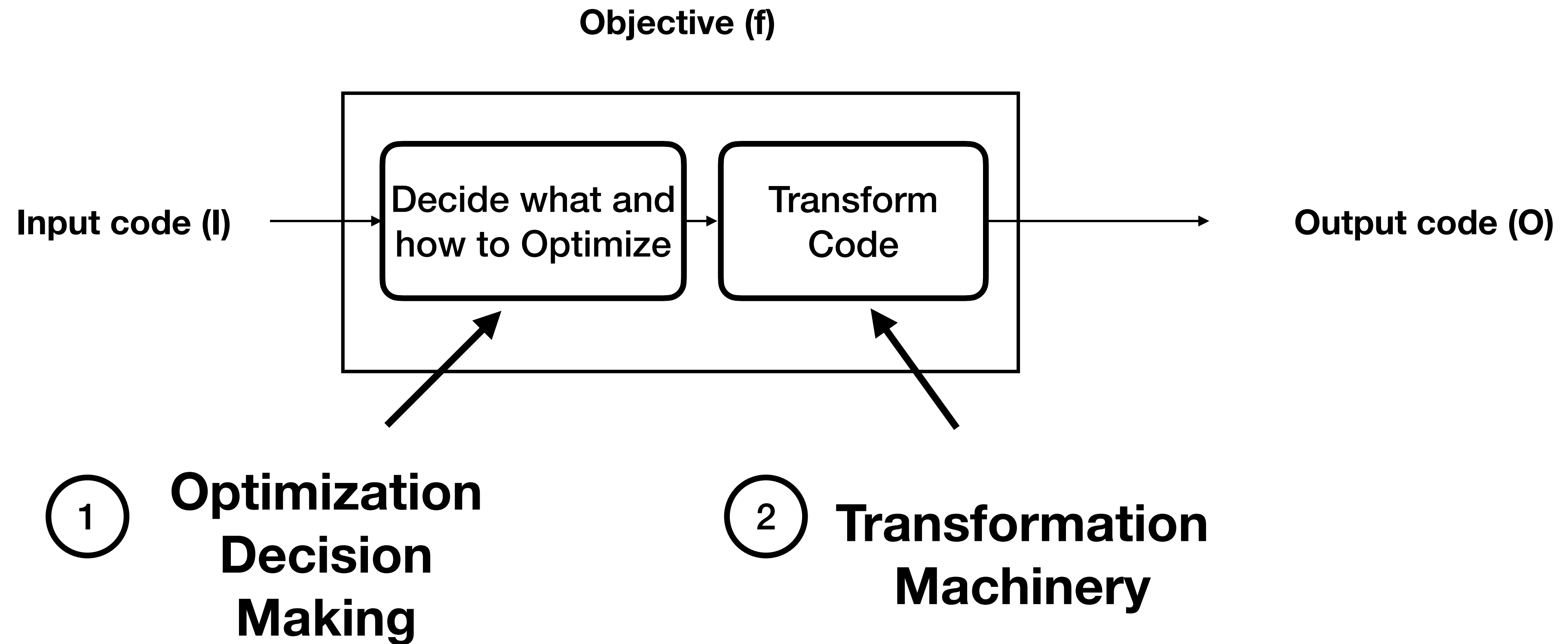


Task: Move from A to B cheaply

1. Plan

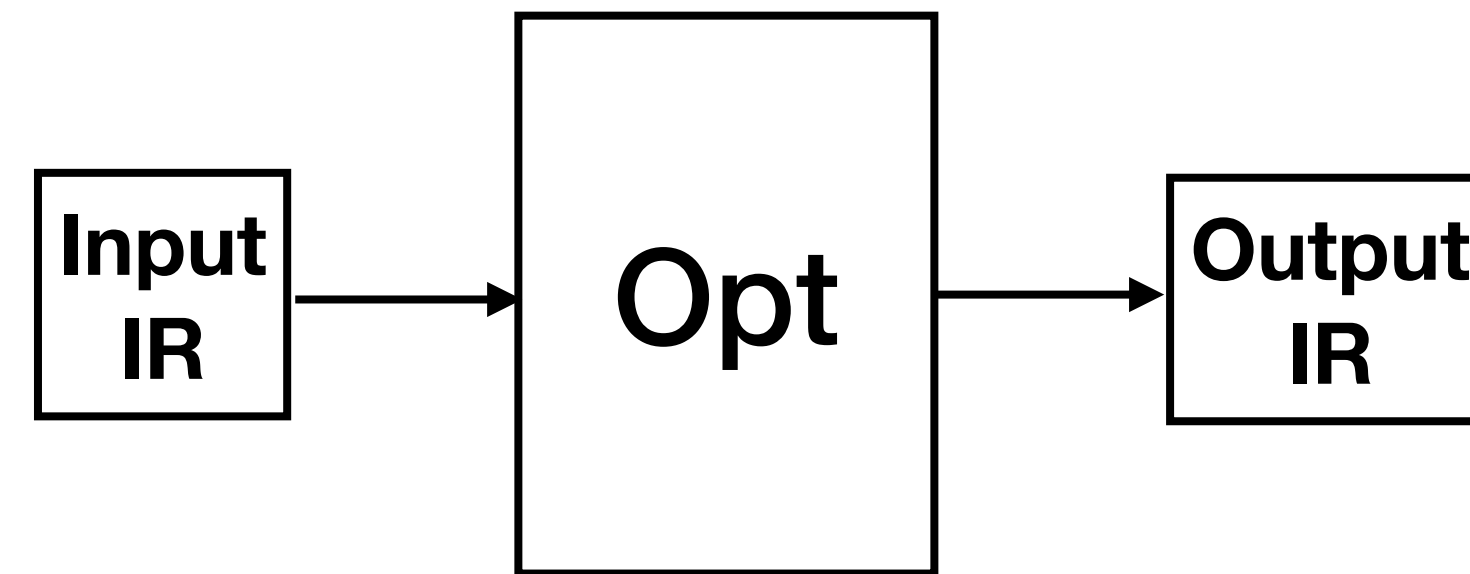
2. Execute

Anatomy of an Optimization Pass

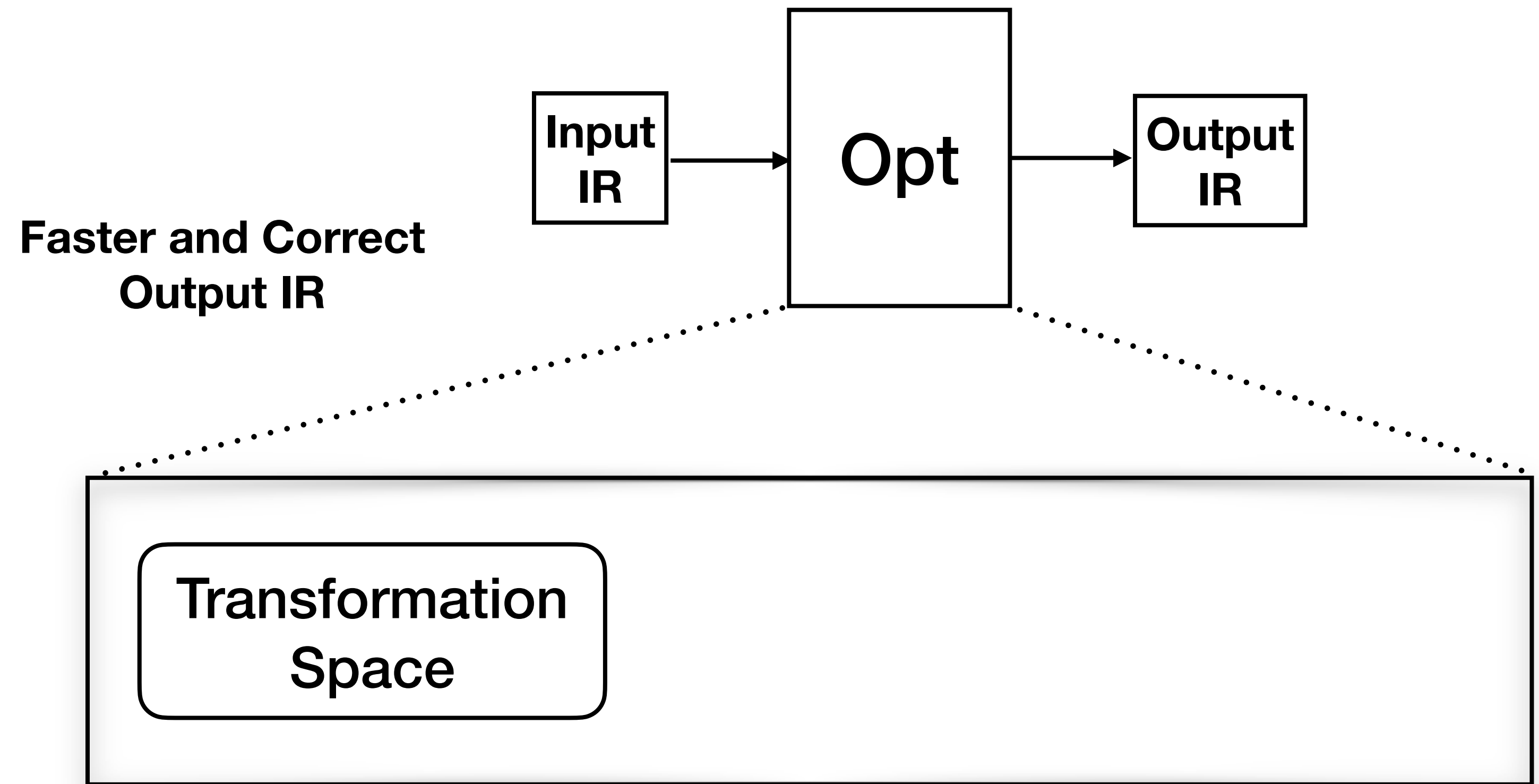


Optimization Decision Making

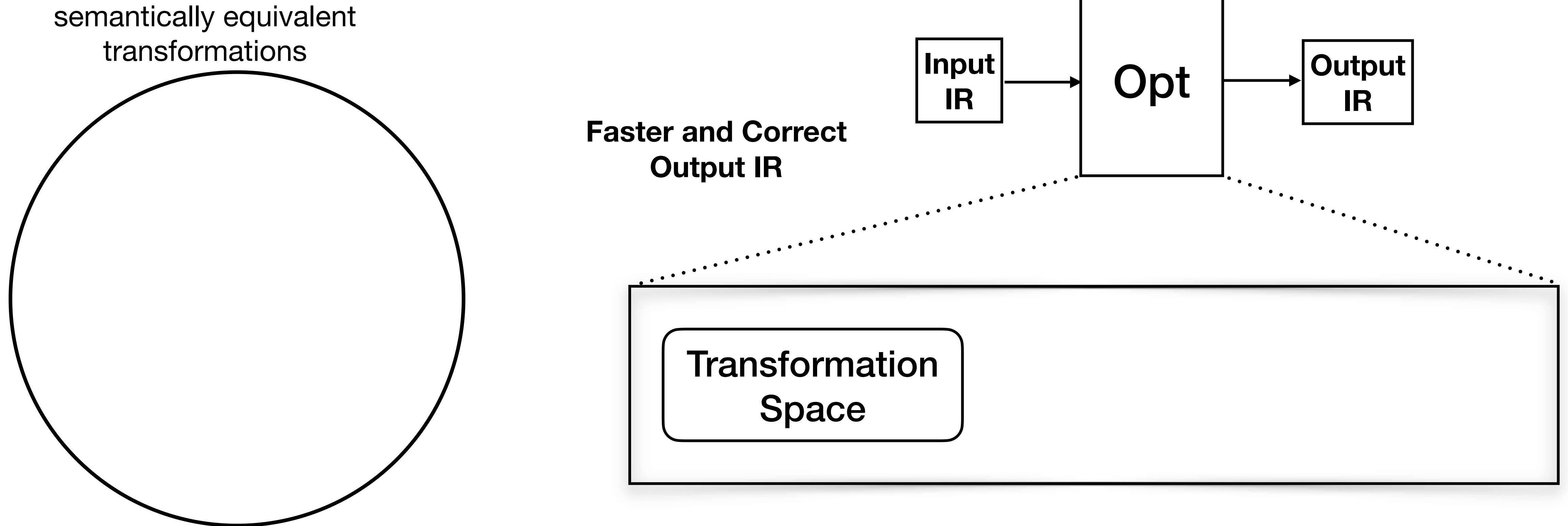
**Faster and Correct
Output IR**



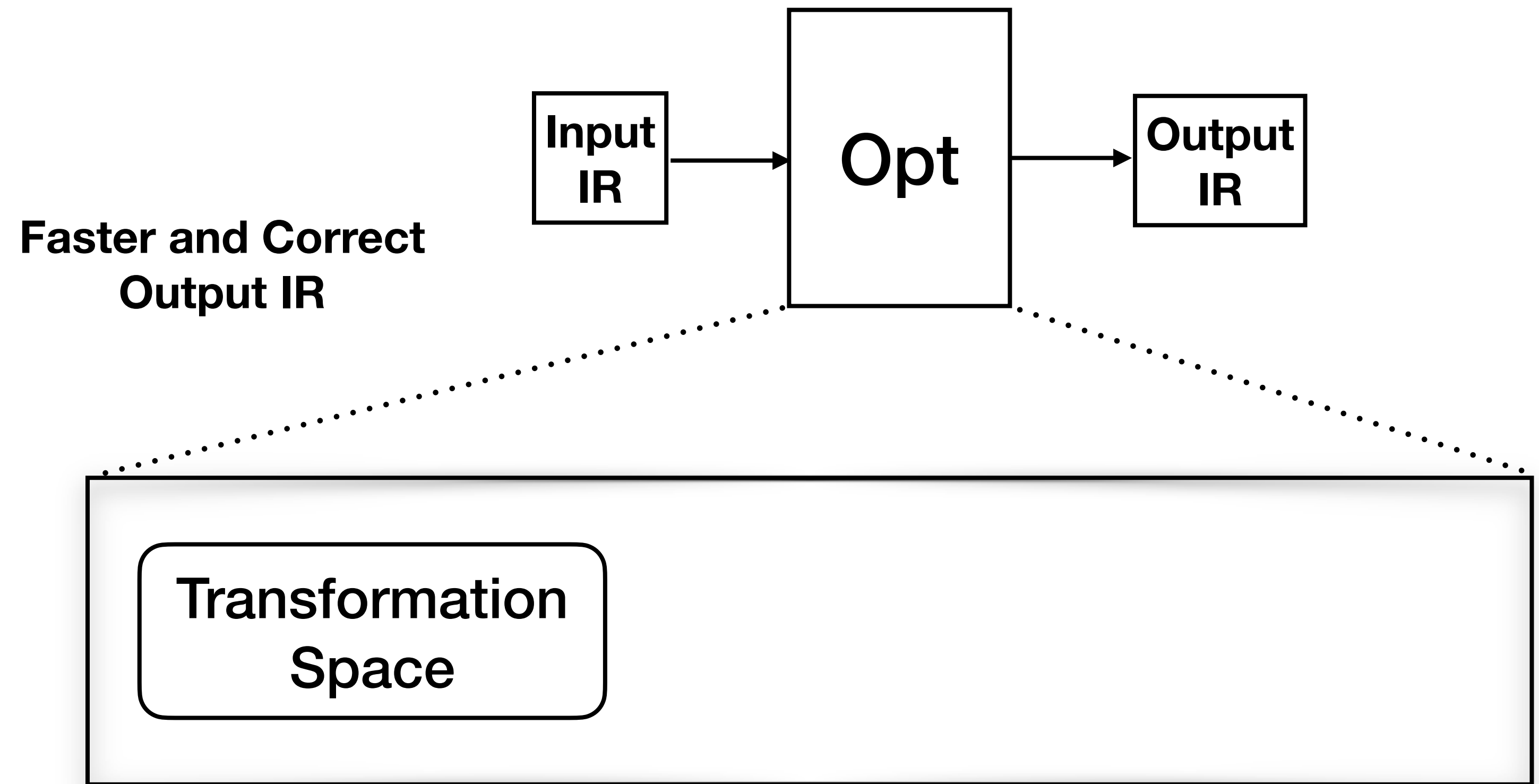
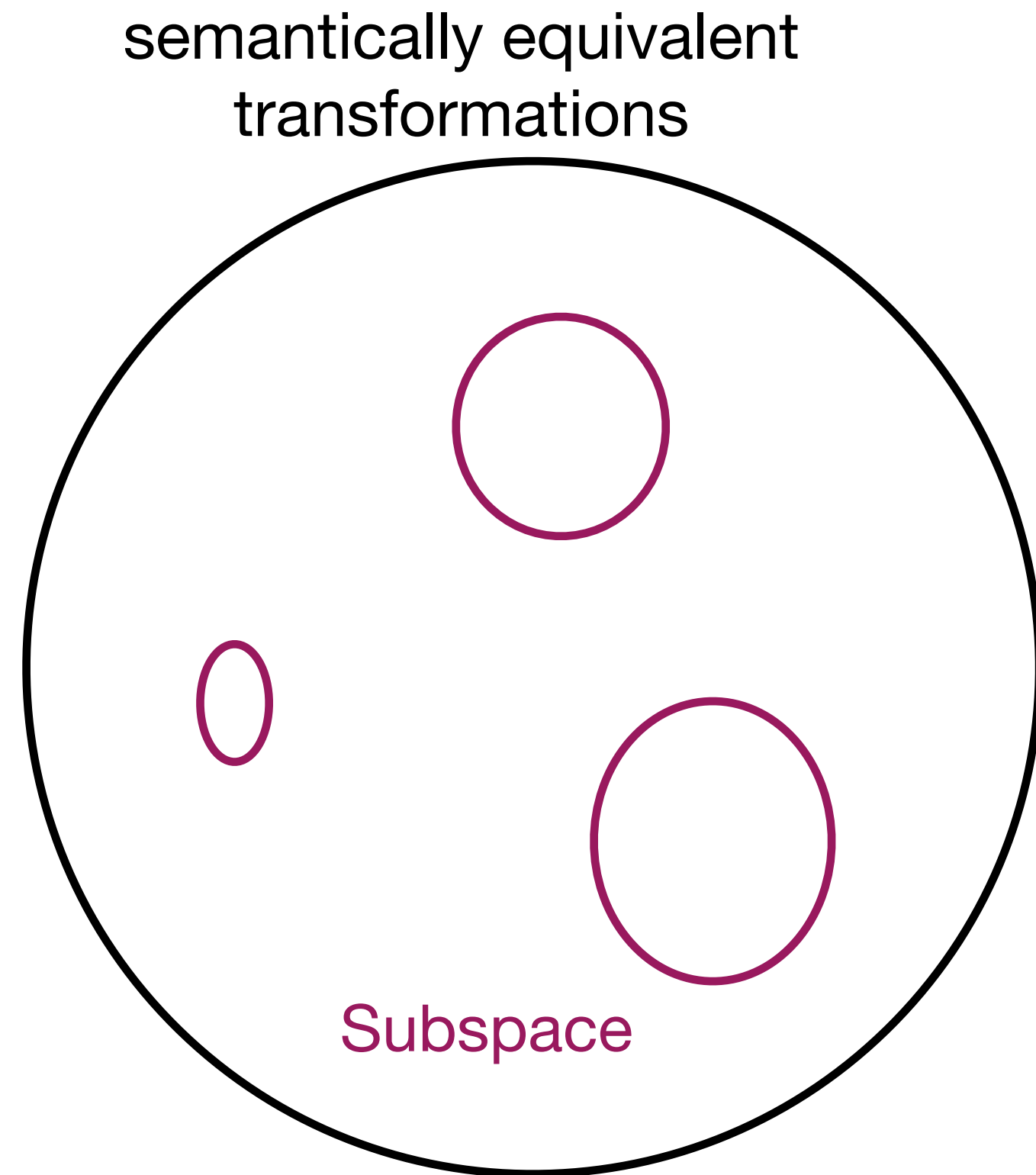
Optimization Decision Making



Optimization Decision Making

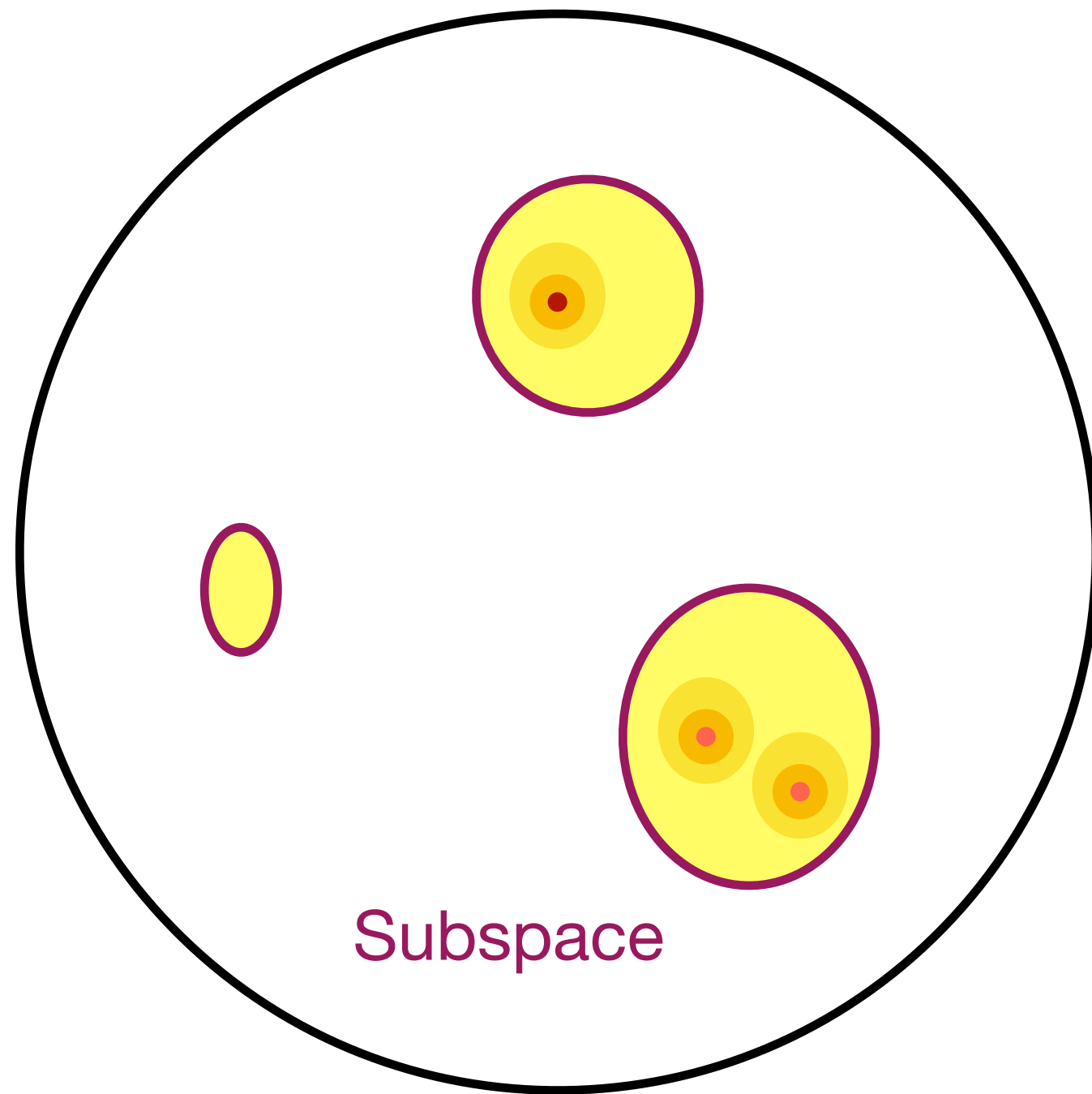


Optimization Decision Making

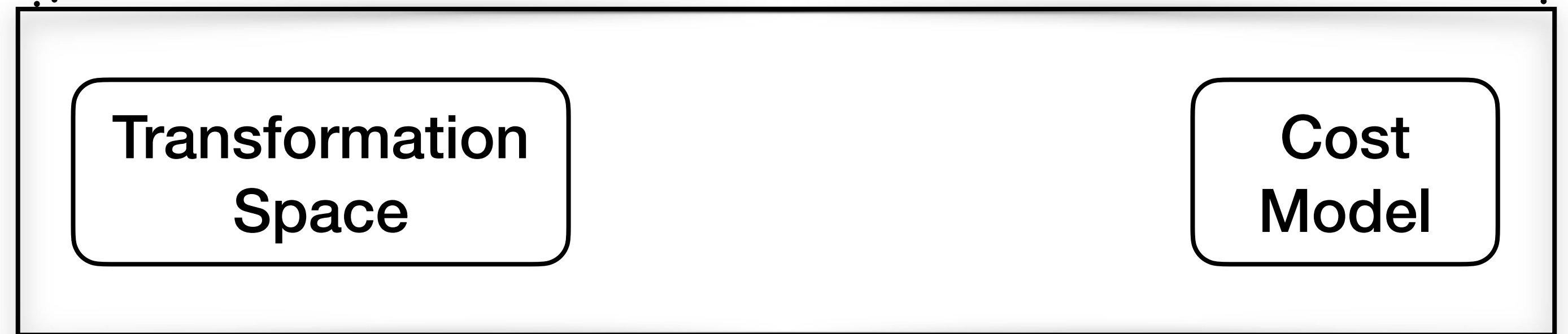
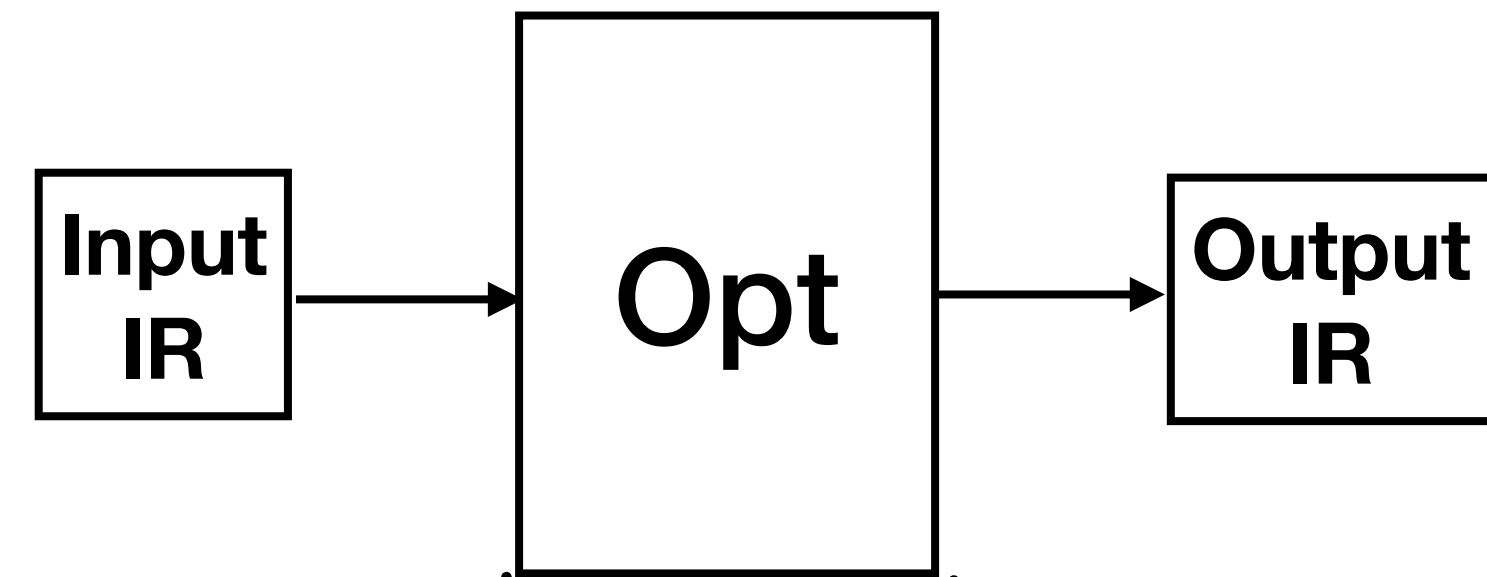


Optimization Decision Making

semantically equivalent transformations

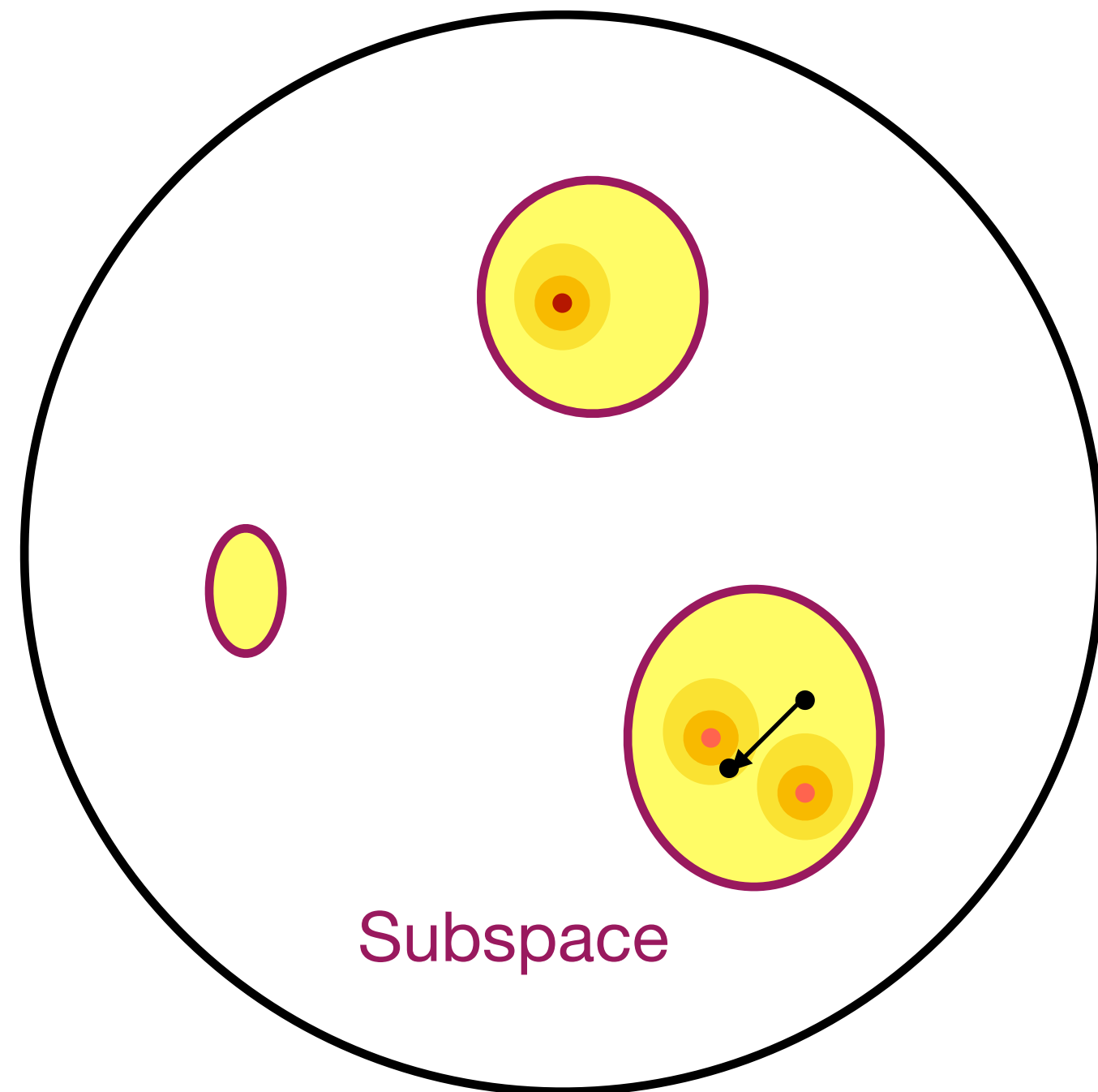


Faster and Correct Output IR

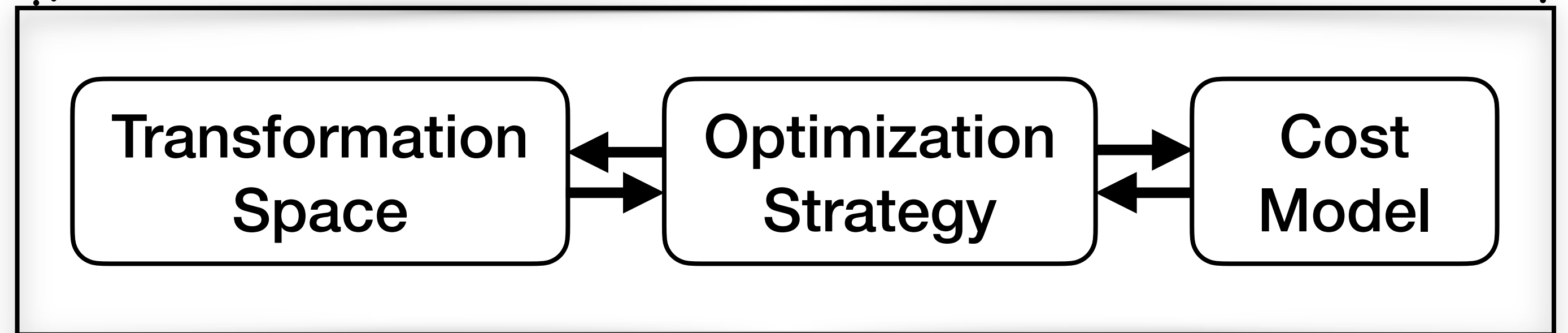
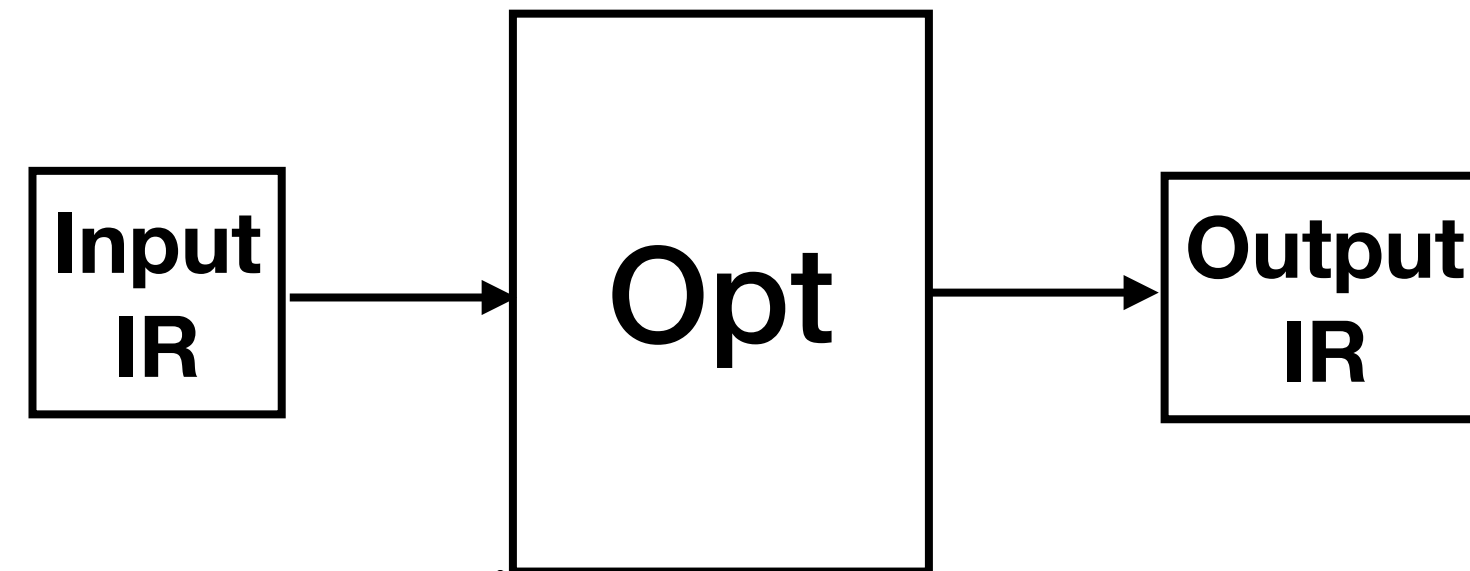


Optimization Decision Making

semantically equivalent transformations

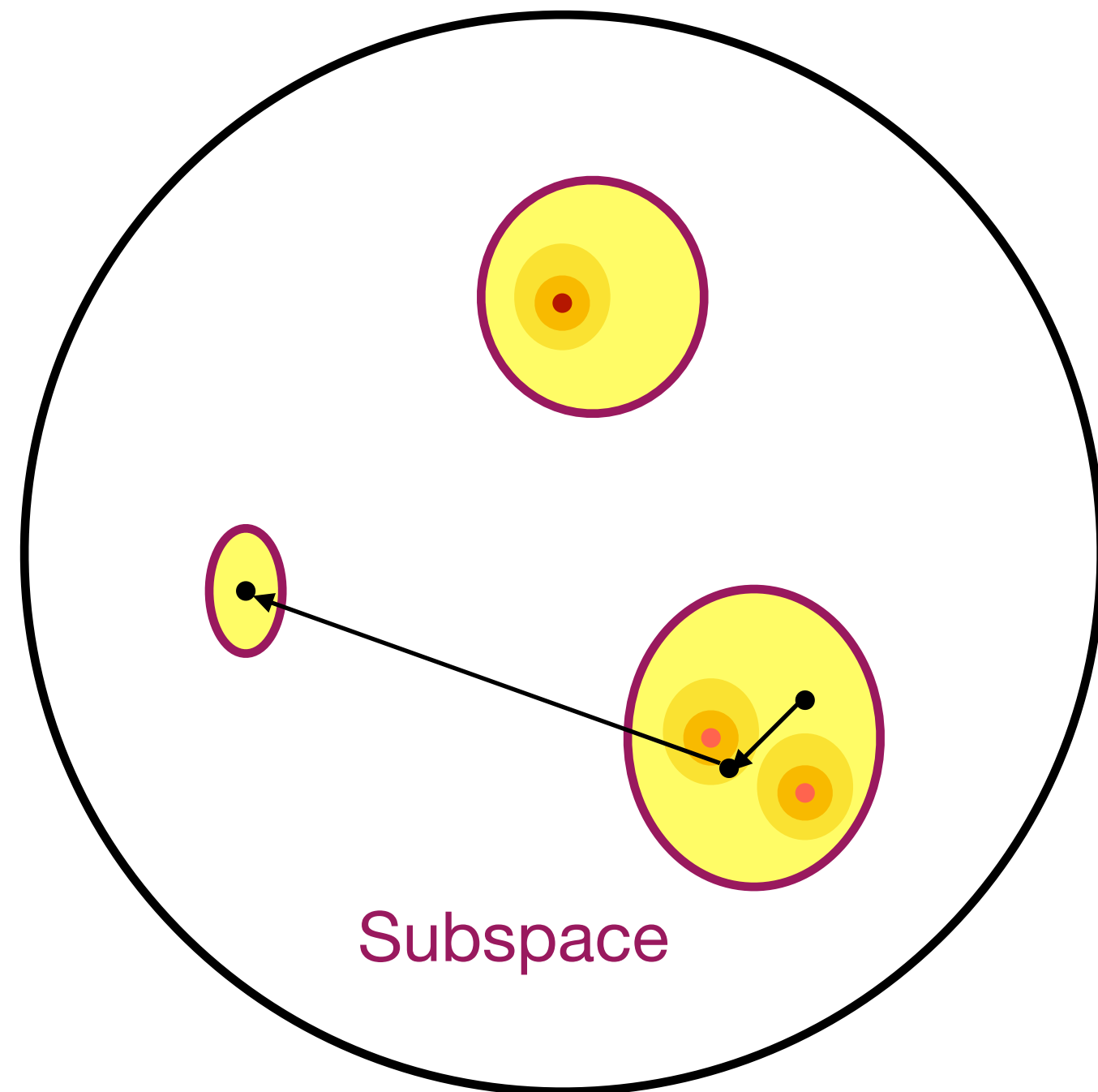


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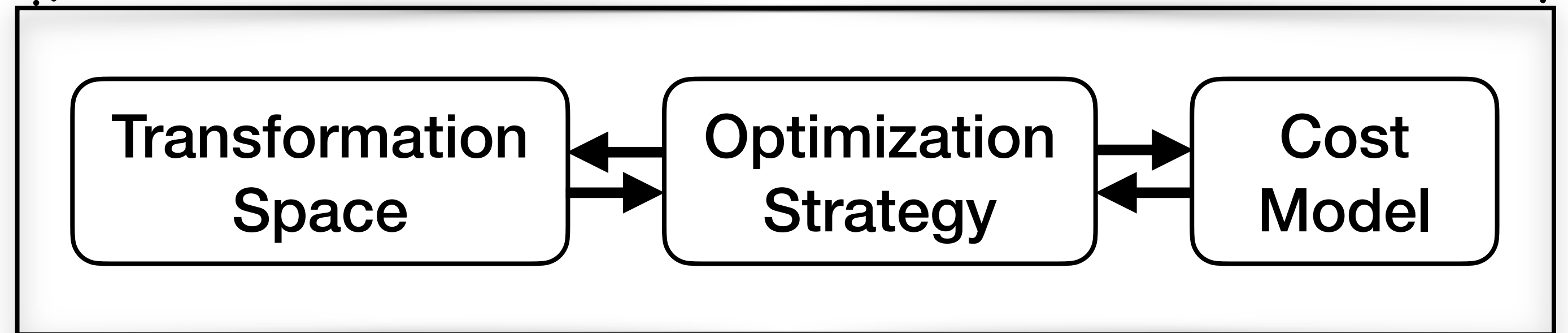
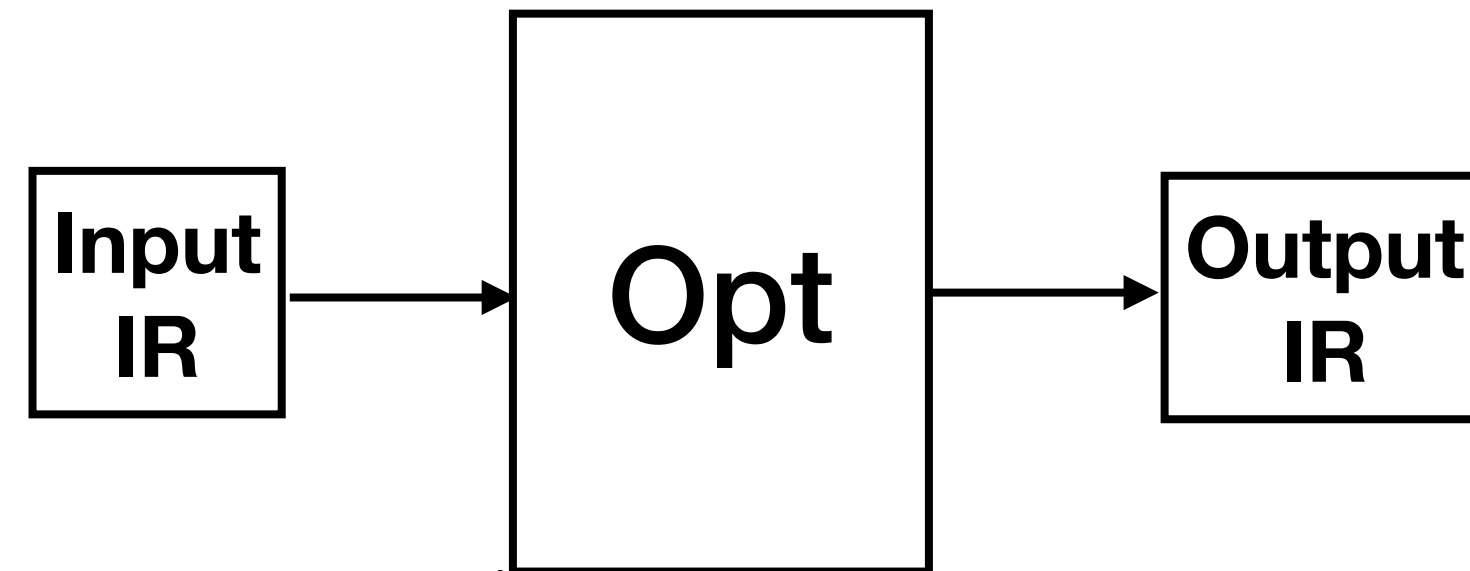


Optimization Decision Making

semantically equivalent transformations

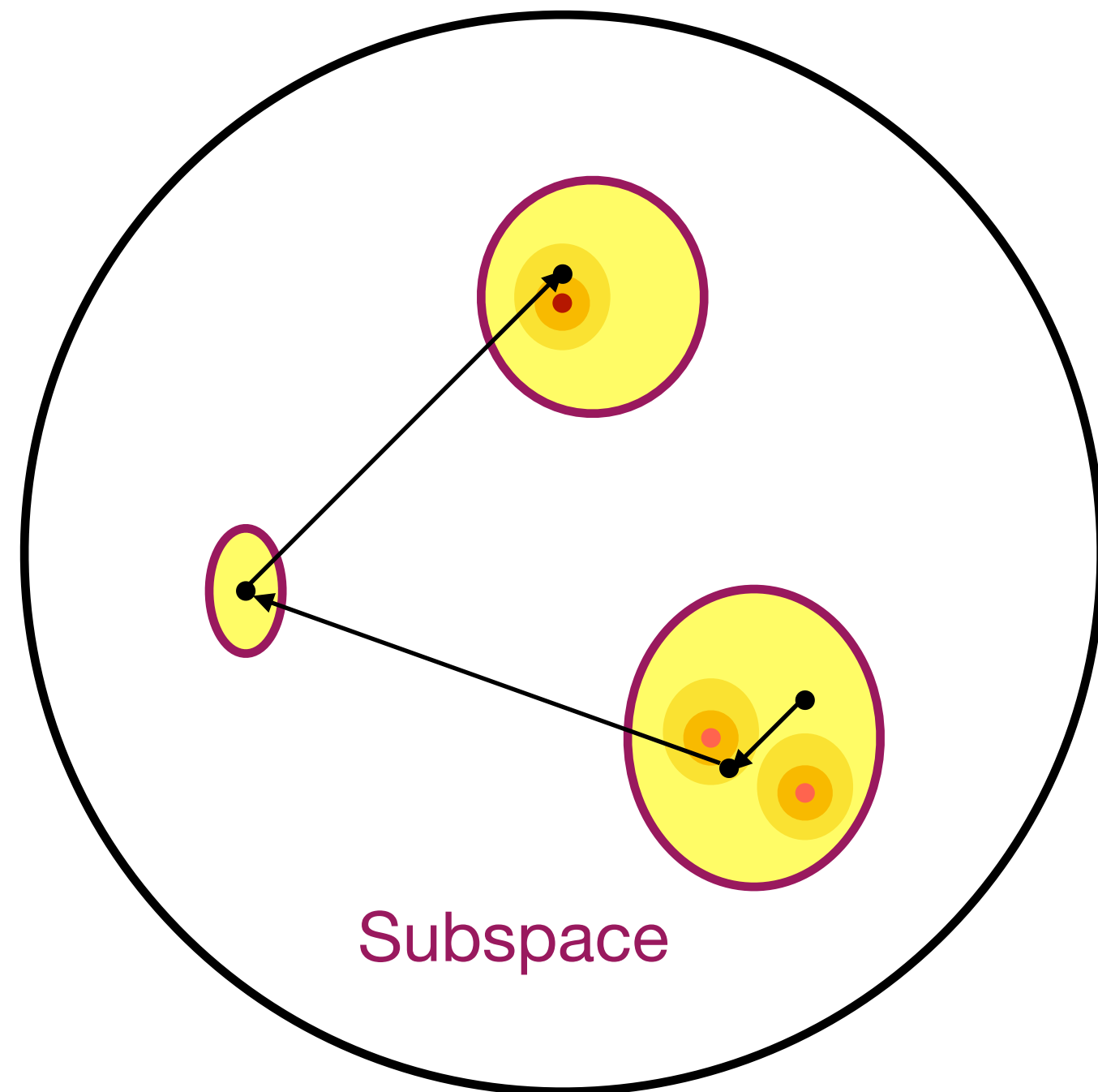


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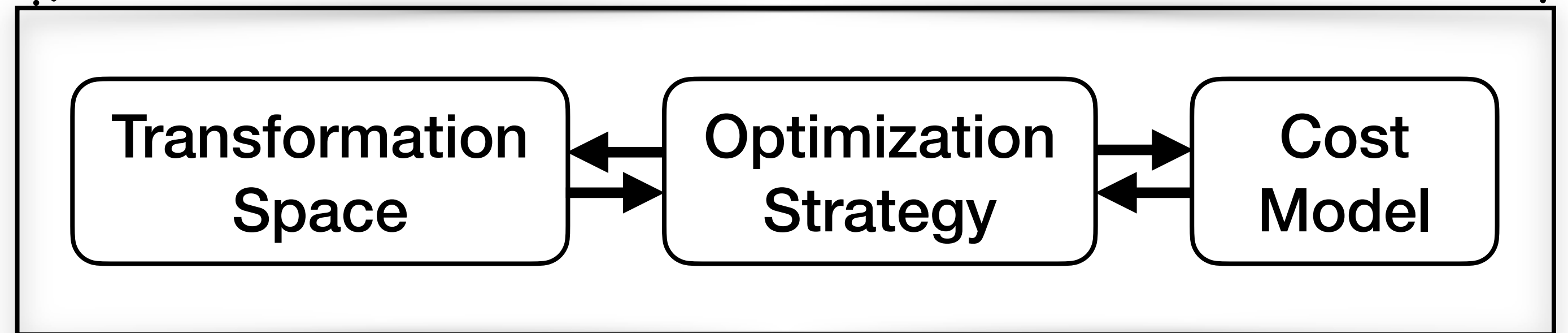
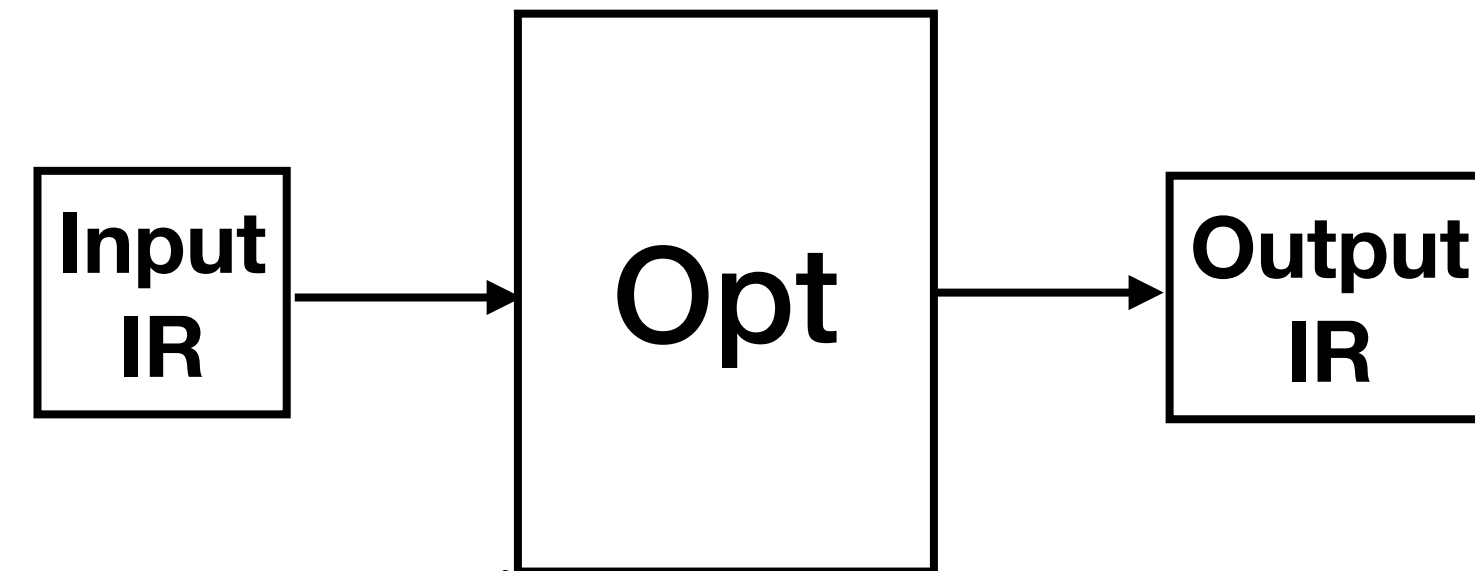


Optimization Decision Making

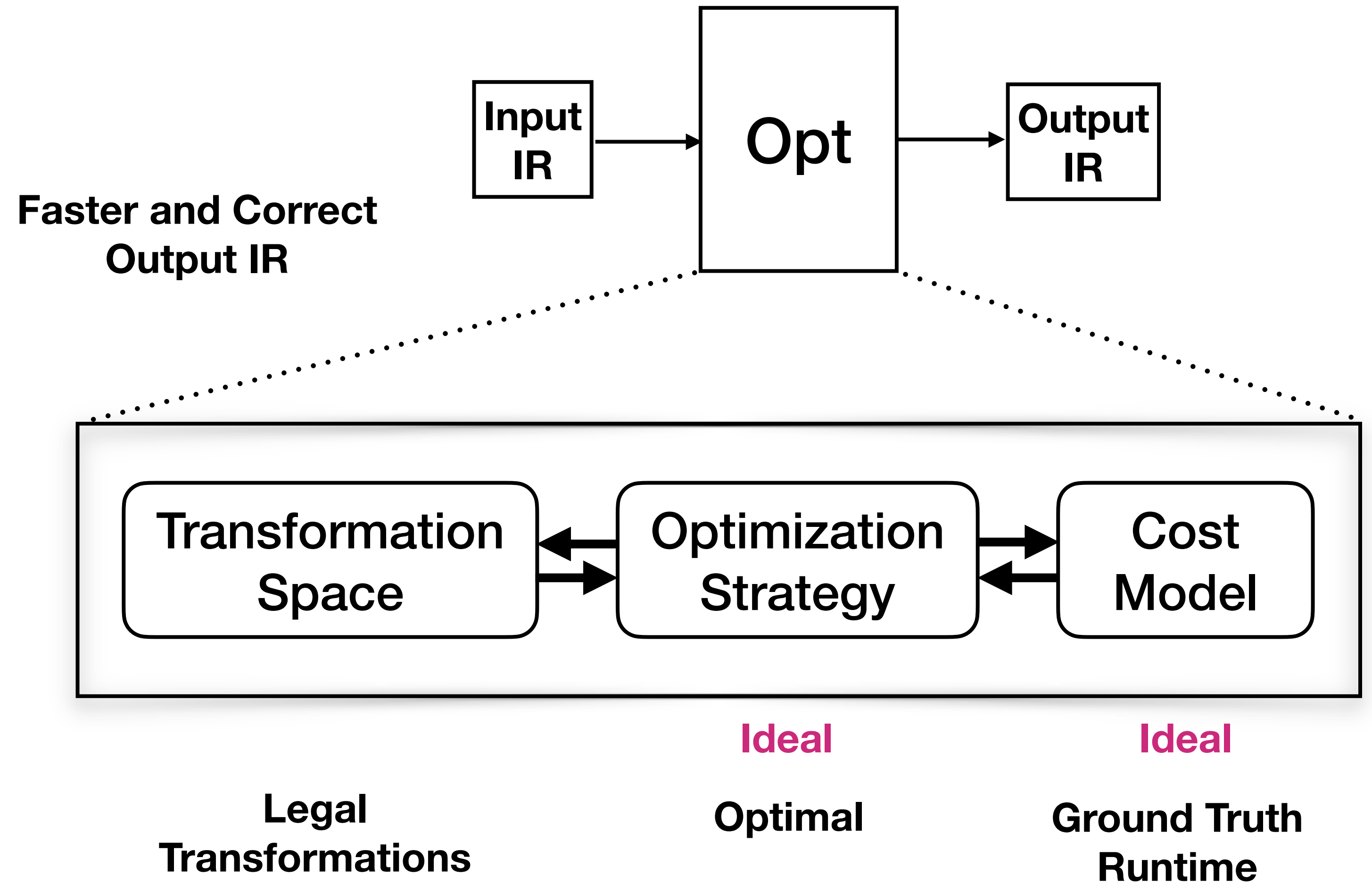
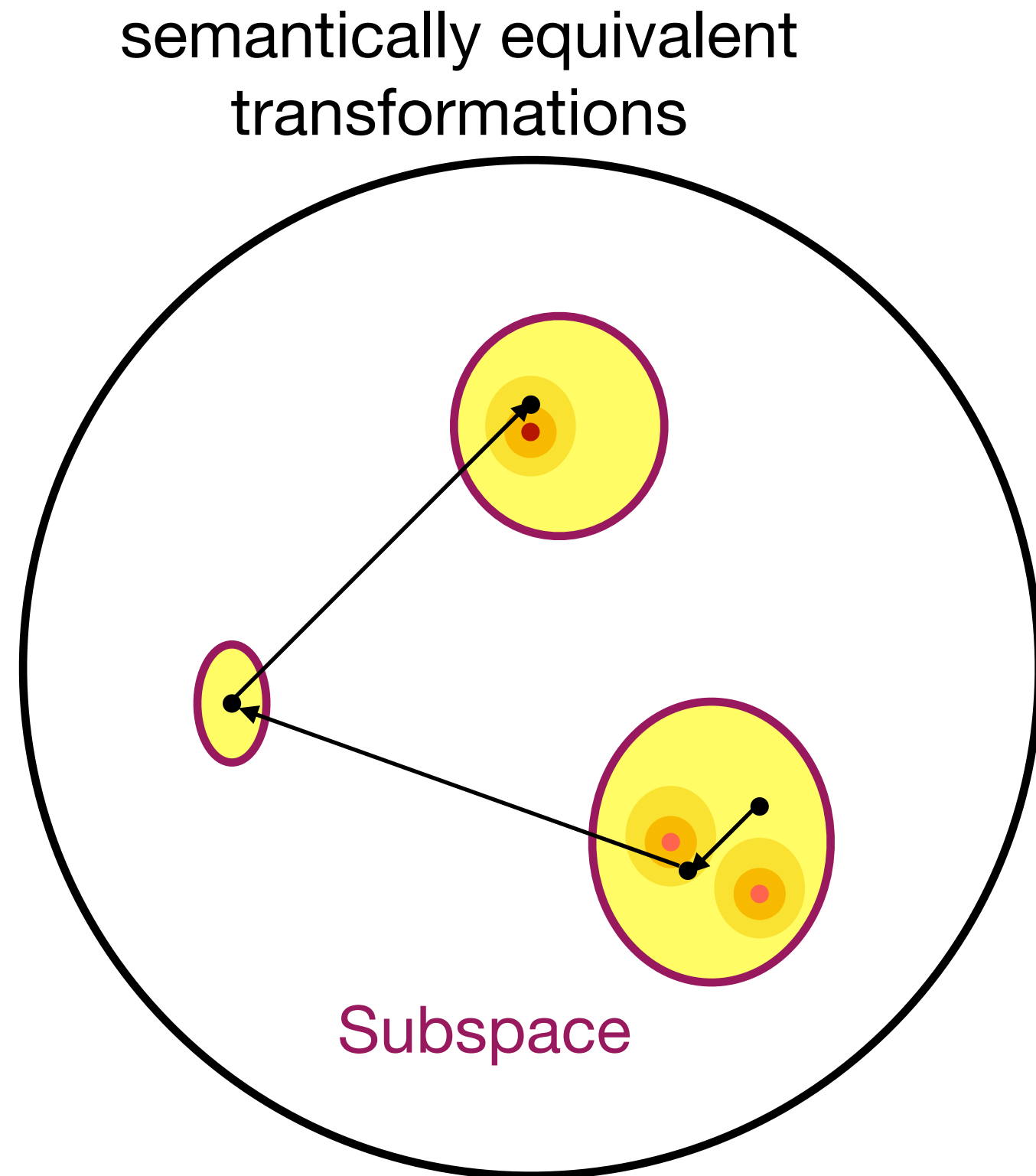
semantically equivalent transformations



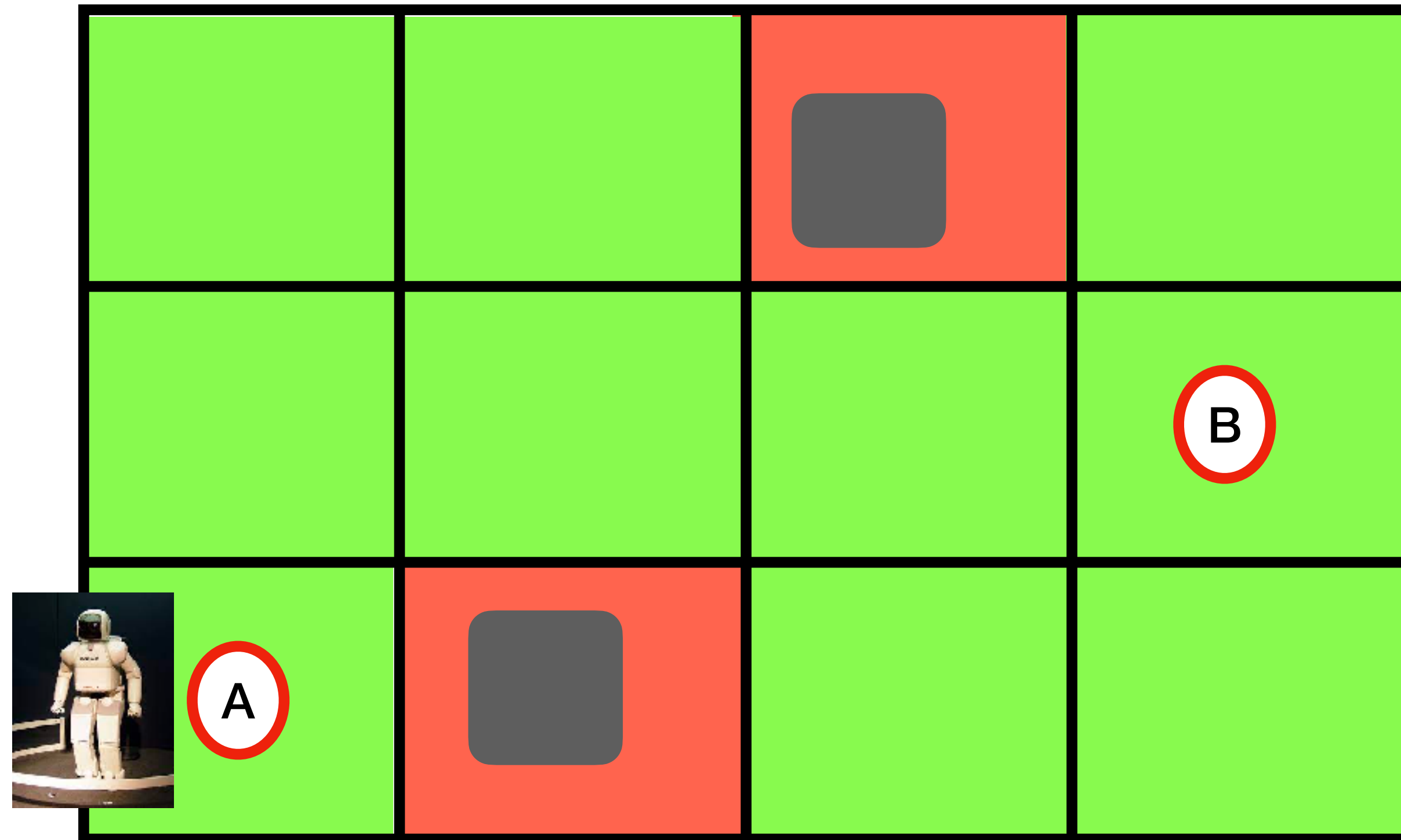
Faster and Correct Output IR



Optimization Decision Making



Robot Analogy



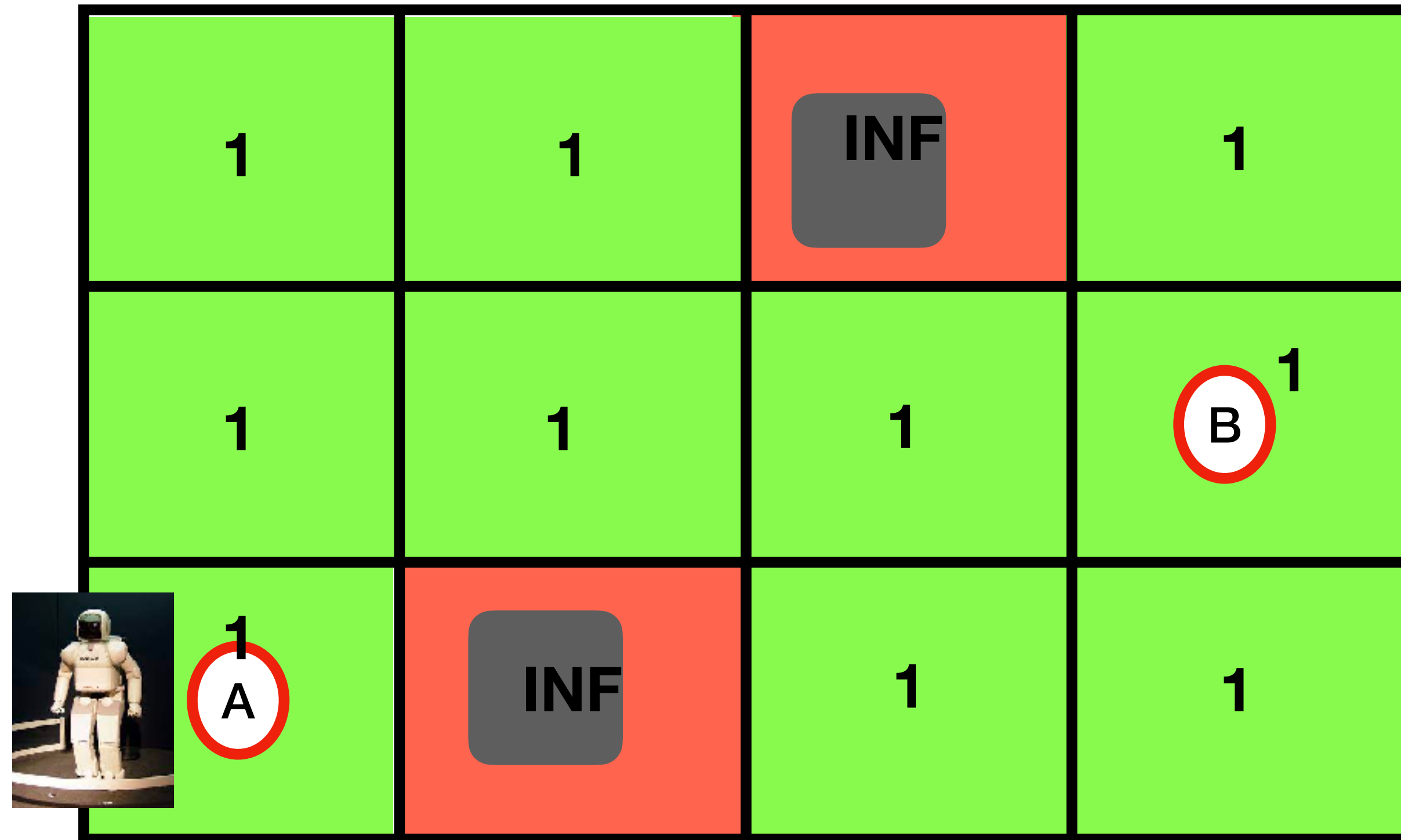
Task: Move from A to B cheaply

1. Plan

2. Execute



Robot Analogy



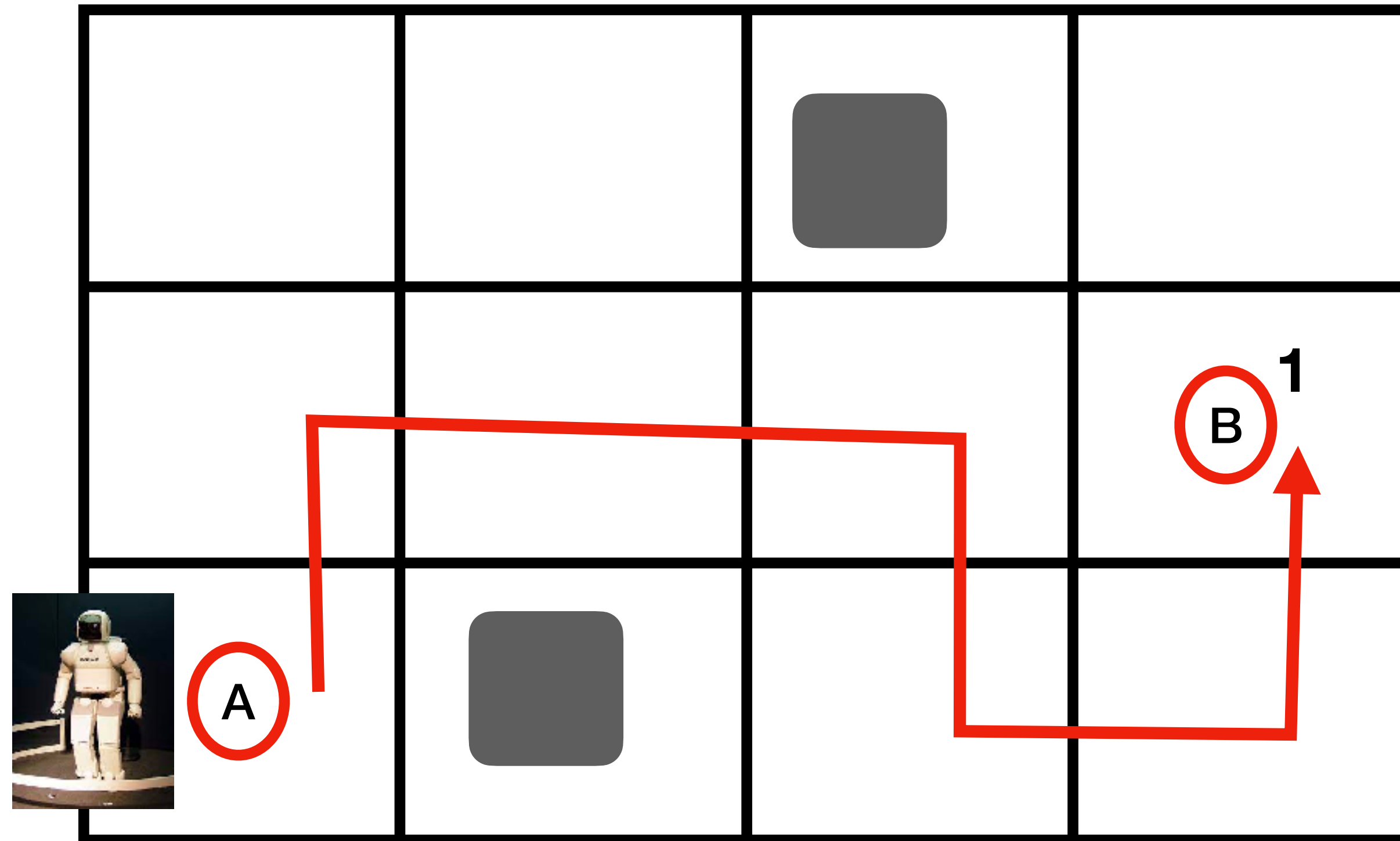
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Robot Analogy

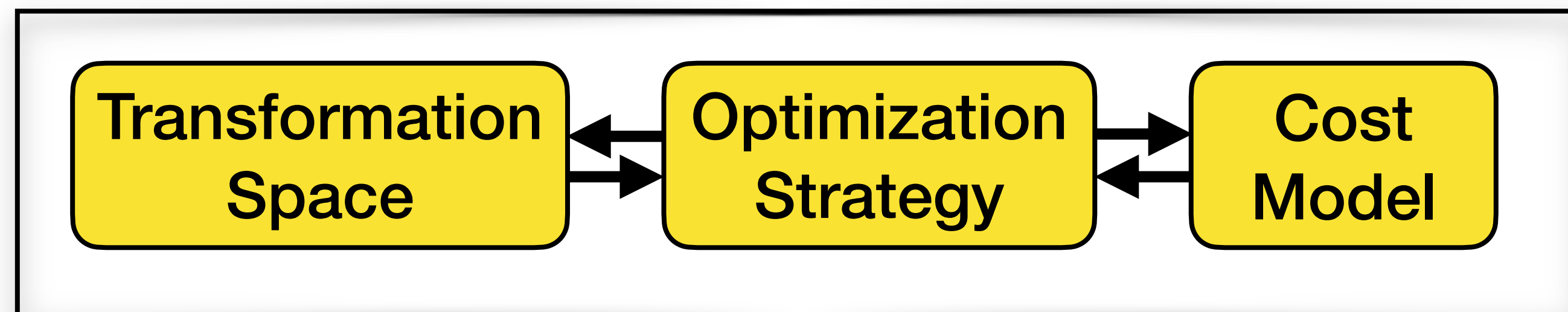


Task: Move from A to B cheaply

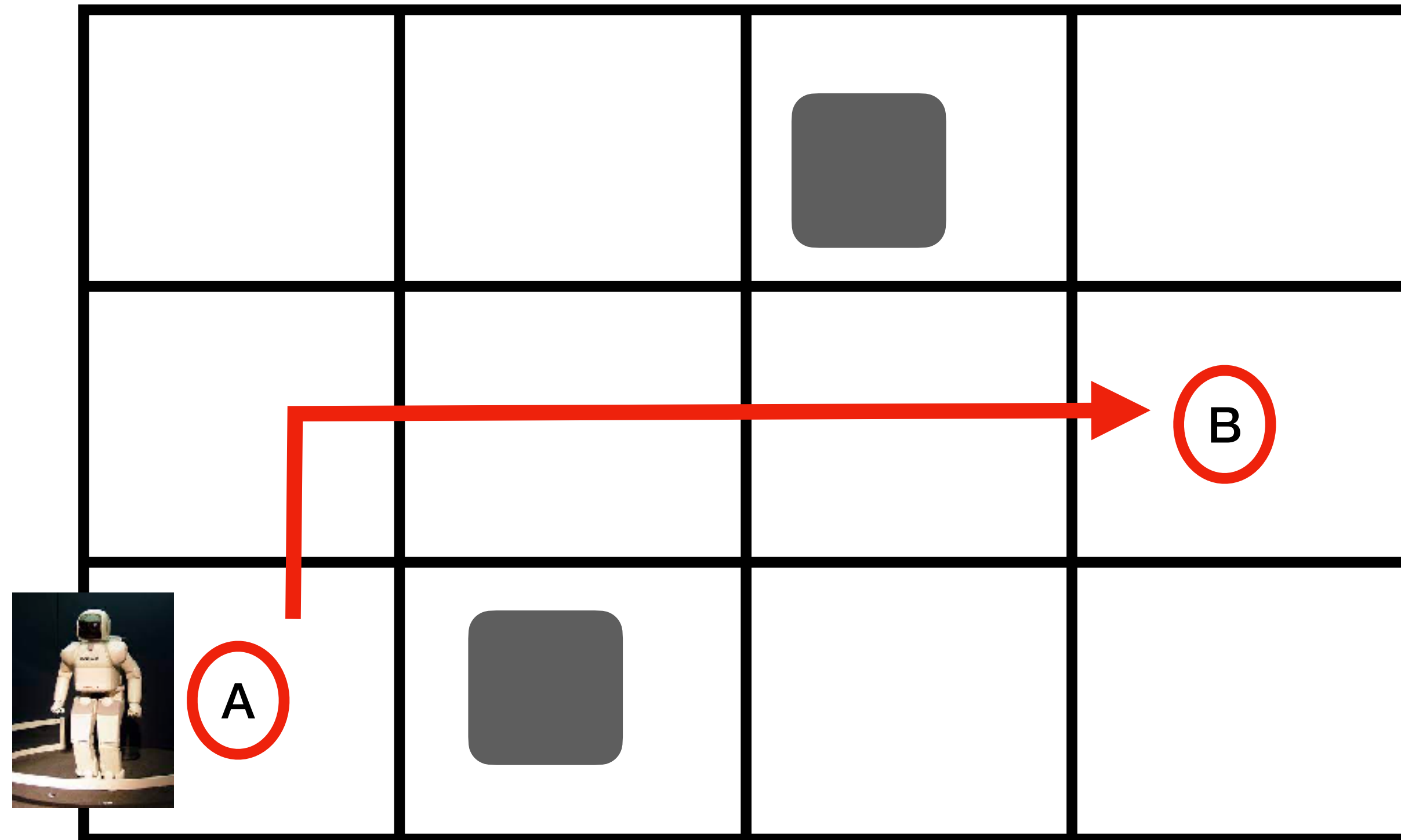
1. Plan

2. Execute

Cost: 7



Robot Analogy

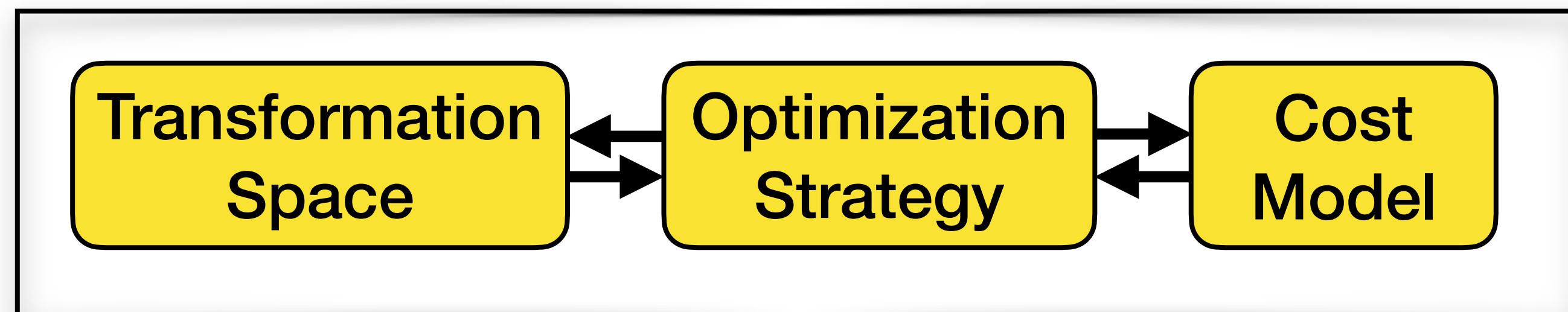


Task: Move from A to B cheaply

1. Plan

2. Execute

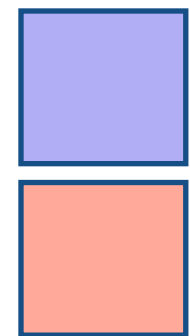
Cost: 5



Vectorization

Independent and Isomorphic statements can be vectorized

Scalar Code



```
a[0] = b[0] + c[0]
a[1] = b[1] + c[1]
```

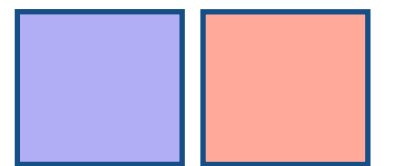
Vector Packs



Vector Code

Single Instruction Multiple Data (SIMD)

```
{a[0], a[1]} = {b[0], b[1]} + {c[0], c[1]}
```



Statement Packing Problem

- Find **independent** and **isomorphic** statements
- Not all vector packs can exist with each other
- Need to select the most profitable packing strategy

S1	:	A1	=	L[5]	/	L[2]
S2	:	A2	=	L[6]	/	L[3]
S3	:	A3	=	L[7]	/	L[4]
S4	:	A4	=	L[1]	-	A2
S5	:	A5	=	L[2]	-	A3
S6	:	A6	=	L[3]	-	A1

{S1, S2}

{S2, S3}

{S1, S3}

{S4, S5}

{S5, S6}

{S4, S6}

Transformation Space

Statement packing strategy 1

Scalar code

```
S1 : A1 = L[5] / L[2]
S2 : A2 = L[6] / L[3]
S3 : A3 = L[7] / L[4]
S4 : A4 = L[1] - A2
S5 : A5 = L[2] - A3
S6 : A6 = L[3] - A1
```

Vector code

```
S1 : A1 = L[5] / L[2]
S2 : A2 = L[6] / L[3]

S3 : A3 = L[7] / L[4]

S4 : A4 = L[1] - A2
S5 : A5 = L[2] - A3
S6 : A6 = L[3] - A1
```

Instruction Breakdown

0 vector
0 packing
0 unpacking

There are costs associated with vectorization

Scalar code

```
S1 : A1 = L[5] / L[2]
S2 : A2 = L[6] / L[3]
S3 : A3 = L[7] / L[4]
S4 : A4 = L[1] - A2
S5 : A5 = L[2] - A3
S6 : A6 = L[3] - A1
```

Vector code

```
SV1 : {A1, A2} = {L[5], L[6]} / {L[2], L[3]}
S3   : A3 = L[7] / L[4]
S4   : A4 = L[1] - A2
SV2 : {A5, A6} = {L[2], L[3]} - {A3, A1}
```

Non-isomorphic

Instruction Breakdown

4 vector
0 packing
0 unpacking

There are costs associated with vectorization

Scalar code

```
S1 : A1 = L[5] / L[2]
S2 : A2 = L[6] / L[3]
S3 : A3 = L[7] / L[4]
S4 : A4 = L[1] - A2
S5 : A5 = L[2] - A3
S6 : A6 = L[3] - A1
```

Vector code

```
SV1 : {A1, A2} = {L[5], L[6]} / {L[2], L[3]}
SU1 : A1 = unpack(SV1, 1)
S3 : A3 = L[7] / L[4]
S4 : A4 = L[1] - A2
SV2 : {A5, A6} = {L[2], L[3]} - {A3, A1}
```

Instruction Breakdown

4 vector
0 packing
1 unpacking

There are costs associated with vectorization

Scalar code

```
S1 : A1 = L[5] / L[2]
S2 : A2 = L[6] / L[3]
S3 : A3 = L[7] / L[4]
S4 : A4 = L[1] - A2
S5 : A5 = L[2] - A3
S6 : A6 = L[3] - A1
```

Vector code

```
SV1 : {A1, A2} = {L[5], L[6]} / {L[2], L[3]}
SU1 : A1 = unpack(SV1, 1)

S3 : A3 = L[7] / L[4]
SP1 : {A3, A1} = pack(A3, A1)
S4 : A4 = L[1] - A2
SV2 : {A5, A6} = {L[2], L[3]} - {A3, A1}
```

Instruction Breakdown

4 vector
1 packing
1 unpacking

There are costs associated with vectorization

Scalar code

```
S1 : A1 = L[5] / L[2]
S2 : A2 = L[6] / L[3]
S3 : A3 = L[7] / L[4]
S4 : A4 = L[1] - A2
S5 : A5 = L[2] - A3
S6 : A6 = L[3] - A1
```

Vector code

```
SV1 : {A1, A2} = {L[5], L[6]} / {L[2], L[3]}
SU1 : A1 = unpack(SV1, 1)
SU2 : A2 = unpack(SV1, 2)
S3   : A3 = L[7] / L[4]
SP1 : {A3, A1} = pack(A3, A1)
S4   : A4 = L[1] - A2
SV2 : {A5, A6} = {L[2], L[3]} - {A3, A1}
```

Instruction Breakdown

4 vector
1 packing
2 unpacking

Statement packing strategy 2

Scalar code

```
S1 : A1 = L[5] / L[2]
S2 : A2 = L[6] / L[3]
S3 : A3 = L[7] / L[4]
S4 : A4 = L[1] - A2
S5 : A5 = L[2] - A3
S6 : A6 = L[3] - A1
```

Vector code

```
SV1 : {A2,A3} = {L[6],L[7]} / {L[3],L[4]}
SU1 : L[2] = unpack(SLV1,2)
S1 : A1 = L[5] / L[2]
SU2 : L[3] = unpack(SLV2,1)
SV2 : {A4,A5} = {L[1],L[2]} - {A2,A3}
S6 : A6 = L[3] - A1
```

Instruction Breakdown

5 vector
0 packing
2 unpacking

Different vectorization schemes have different profitability

Strategy 1

Liu et. al [PLDI'12]

```
SV1 : {A1,A2} = {L[5],L[6]} / {L[2],L[3]}
SU1 : A1 = unpack(SV1,1)
SU2 : A2 = unpack(SV1,2)
S3  : A3 = L[7] / L[4]
SP1 : {A3,A1} = pack(A3,A1)
S4  : A4 = L[1] - A2
SV2 : {A5,A6} = {L[2],L[3]} - {A3,A1}
```

4 vector
1 packing
2 unpacking

Strategy 2

Optimal

```
SV1 : {A2,A3} = {L[6],L[7]} / {L[3],L[4]}
SU1 : L[2] = unpack(SLV1,2)
S1  : A1 = L[5] / L[2]
SU2 : L[3] = unpack(SLV2,1)
SV2 : {A4,A5} = {L[1],L[2]} - {A2,A3}
S6  : A6 = L[3] - A1
```

5 vector
0 packing
2 unpacking

Machine Learning Influence

	Transformation Space	Optimization Strategy	Cost Model
Traditional solutions ↓	Hand-written	<ul style="list-style-type: none">• Greedy / Heuristic• Integer Linear Programming• Dynamic Programming	<ul style="list-style-type: none">• Analytical Linear Non-linear
Automated solutions	Program Logics	Data-driven Imitation Learning 10/21	Data-driven LSTM based Cost Model 10/07

Domain Specific Languages

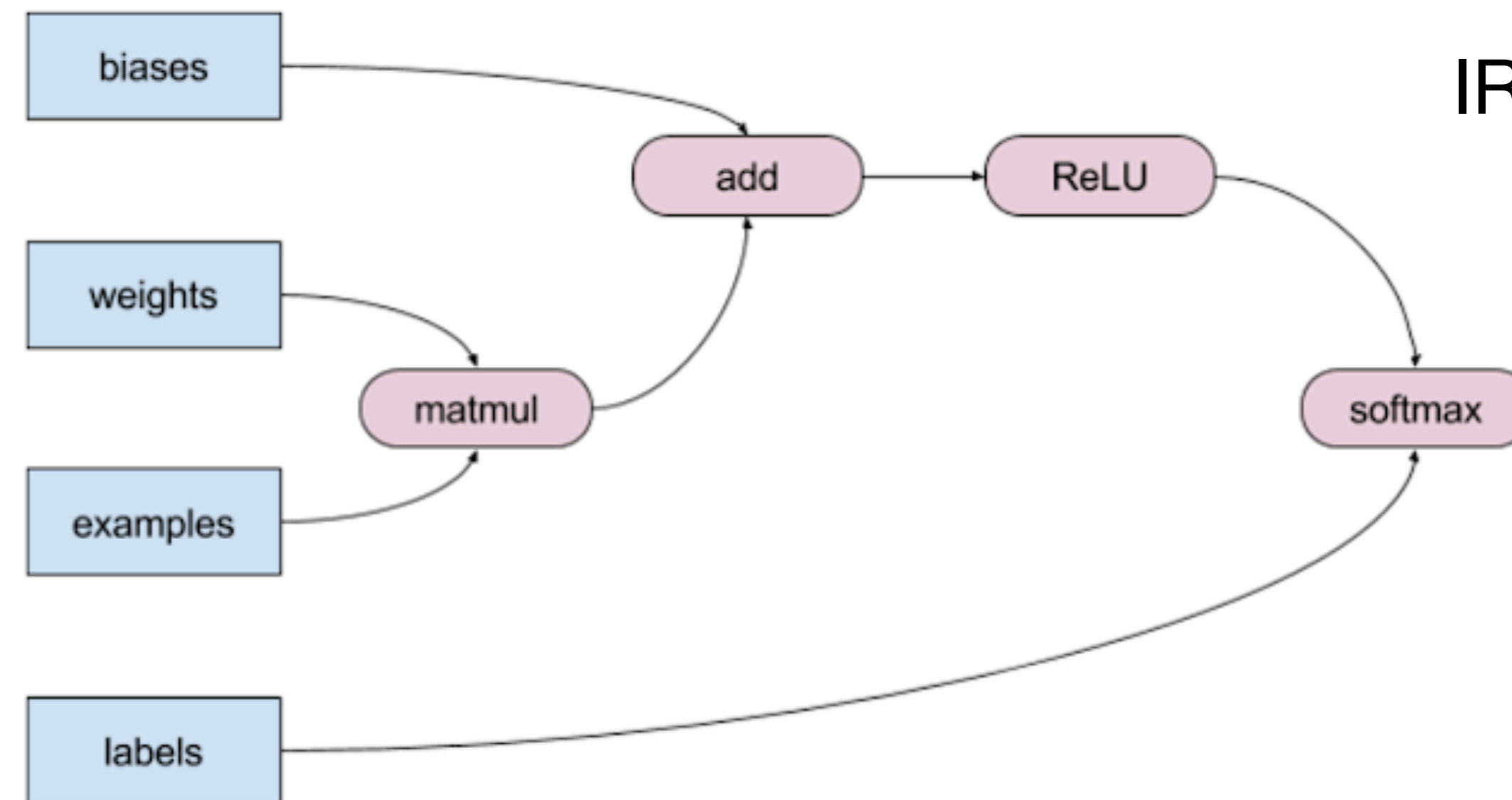
- Programming model specific to one domain
 - Image / Array Processing - Halide, MATLAB
 - Sparse Tensor Computations - TACO
 - Tensor Algebra - Tensorflow, Pytorch (frameworks)
 - Graphs - GraphIt, Gunrock
 - Genomic Computations - Seq
- Usually comes with a set of domain specific optimizations

Halide

- **Main idea:** Separate algorithm specification from optimizations (schedules)
- Halide Video
 - <https://www.youtube.com/watch?v=3uiEyEKji0M&t=3s>
- **Optimization objective:** find the best schedule or optimization sequence for a given Halide algorithm

Tensorflow

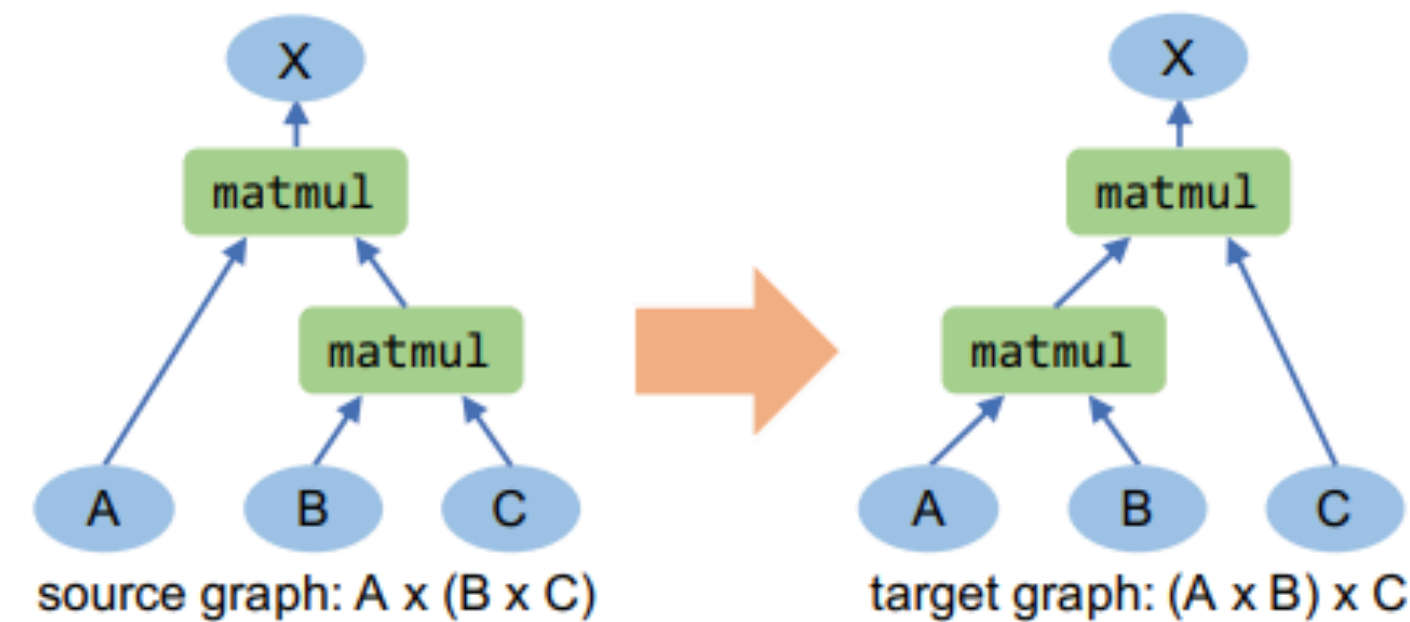
- Model tensor manipulating programs
- Uses the XLA compiler to target GPUs, TPUs and CPUs
- Main abstraction: Computational Graphs



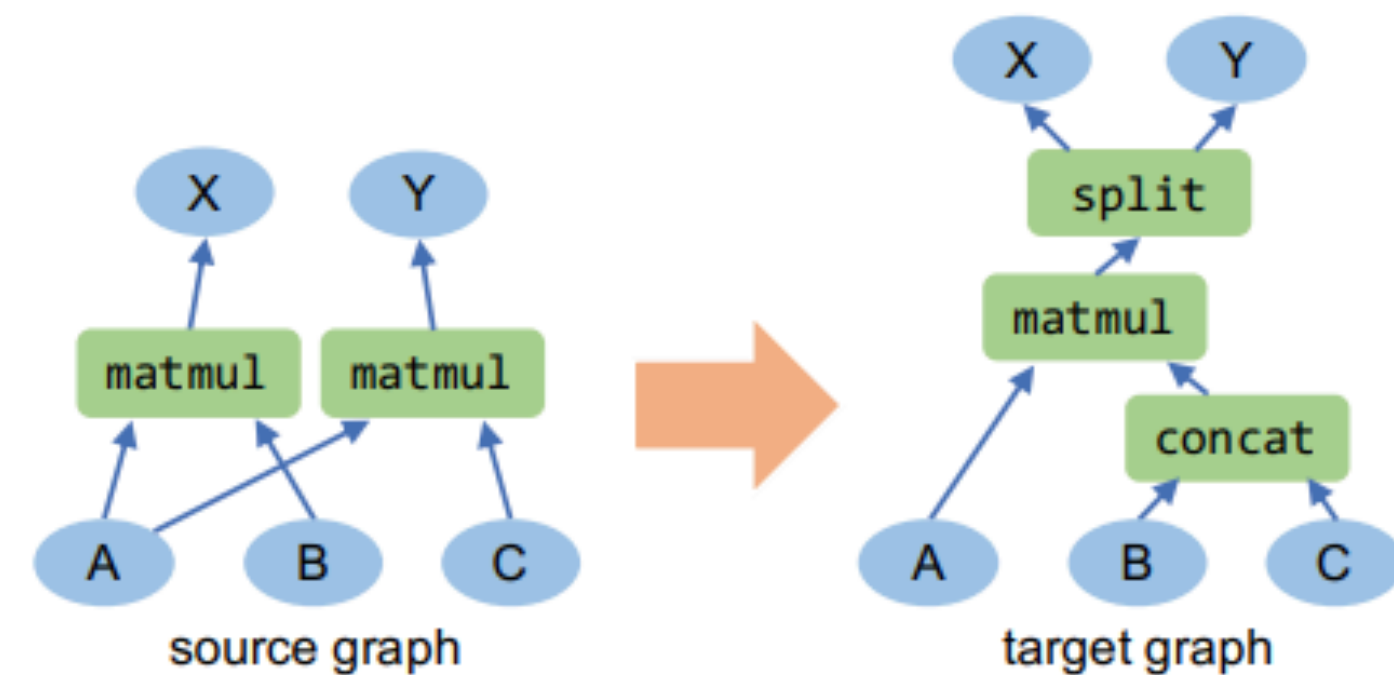
IR: High Level Operations (HLO)

XLA Compiler

- (Most) optimizations can be expressed as computational graph rewrites



(a) Associativity of matrix multiplication.



(b) Fusing two matrix multiplications using concatenation and split.

TASO [SOSP'19]

<https://cs.stanford.edu/~padon/taso-sosp19.pdf>

Machine Learning Influence

Transformation
Space

Optimization
Strategy

Cost
Model

Automated solutions

Program Logics

Data-driven

Data-driven

**10/28: Tree Search
(Halide)**

**11/02: Reinforcement
Learning
(Halide)**

**10/12: GNN based
Cost Model
(XLA)**

Paper Presentation

- Paper presentations assigned on **September 8th**
- **Week before:** Meet instructor to discuss the presentation plan (compulsory!)
 - Use this time to ask questions and discuss the outline
 - Presentation slides are due when reviews are due for that class
 - Submit using the hotCRP system
- **During the class:** Be present in class (compulsory!)
 - Deliver a 20-30 min presentation on the paper
 - Answer questions for the following 15 min
 - Final 30-40 min for open discussion on the paper (lead by the instructor)

Paper Presentation

- **After class:** Summarize the discussion of the paper
 - Submit the summary by the start of the next class (hotCRP)
- First presentation on **September 14th**
 - Matteo Frigo, “A Fast Fourier transform Compiler” (PLDI 1999)
 - 15 min presentation
 - Meet instructor on Thursday 9th to discuss the outline

Any Questions?