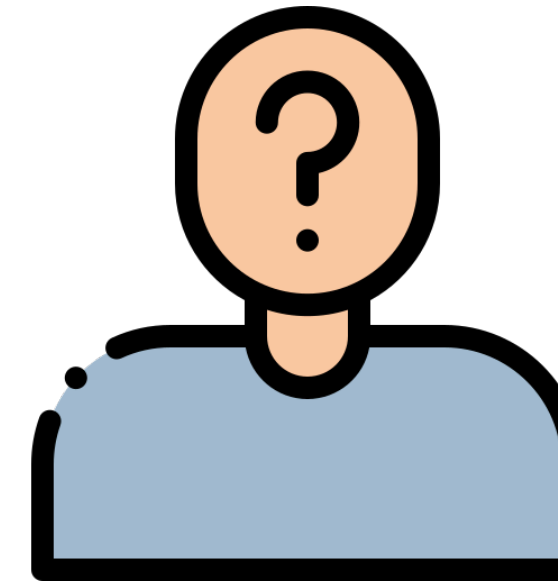


# A Simple Online Problem



... I want to surveil the suspect ...

... while saving the fundings ..

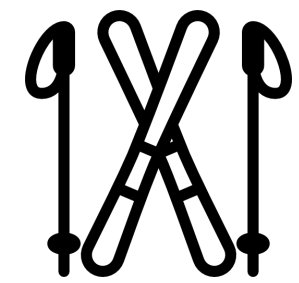


... will ski for an unknown number of days ..

# A Simple Online Problem



... while saving the fundings ..



sale price:           \$100

rental price:         \$10

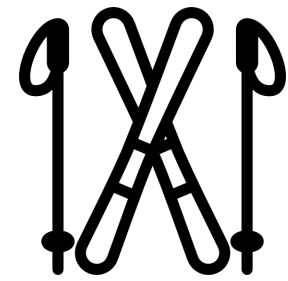


... will ski for  $n$  days ..

↑  
*unknown*

# How to Make Decisions (Buy or Rent) Online?

## The **Ski-Rental** Problem



sale price: \$100  
rental price: \$10

... will ski for  $n$  days ..  
↑  
*unknown*

If  $n$  is known:

If  $n < 10$ , keep renting

Optimal Cost is  $\$10n$

If  $n \geq 10$ , just buy a pair of skis

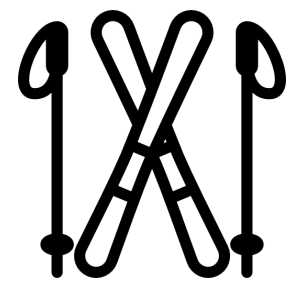
Optimal Cost is \$100

$n$  is unknown:

What should I do?

# 2-Competitive Algorithm

## The Ski-Rental Problem



sale price: \$100

rental price: \$10

... will ski for  $n$  days ..

↑  
*unknown*

If  $n$  is known:

If  $n < 10$ , keep renting

Optimal Cost is  $\$10n$

If  $n \geq 10$ , just buy a pair of skis

Optimal Cost is  $\$100$

$n$  is unknown:

**Buy on day 10 regardless**

If  $n < 10$ , Algorithm Cost is  $\$10n$

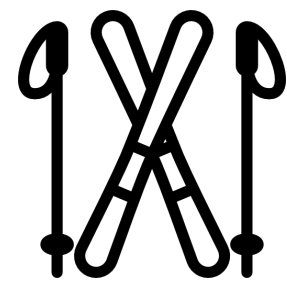
If  $n \geq 10$ , Algorithm Cost is  $\$200$

$$\frac{200}{10}$$

$$\max_{n \geq 1} \frac{\text{Algorithm Cost}(n)}{\text{Optimal Cost}(n)} \leq 2$$

# Competitive Ratio

## The Ski-Rental Problem



sale price: \$100

rental price: \$10

... will ski for  $n$  days ..

↑  
*unknown*

If  $n$  is known:

If  $n < 10$ , keep renting

Optimal Cost is  $\$10n$

If  $n \geq 10$ , just buy a pair of skis

Optimal Cost is  $\$100$

$n$  is unknown:

Buy on day 10 regardless

If  $n < 10$ , Algorithm Cost is  $\$10n$

If  $n \geq 10$ , Algorithm Cost is  $\$200$

$$\text{Competitive Ratio} := \max_{n \geq 1} \frac{\text{Algorithm Cost}(n)}{\text{Optimal Cost}(n)} \leq 2$$

denoted by **CR**

# From Classic to Learning-Augmented World

## The **Ski-Rental** Problem

$n$  is unknown:

Buy on day 10 regardless

If  $n < 10$ , algorithm cost is  $\$10n$

If  $n \geq 10$ , algorithm cost is  $\$200$

$$\text{Competitive Ratio} := \max_{n \geq 1} \frac{\text{Algorithm Cost}(n)}{\text{Optimal Cost}(n)} \leq 2$$

— This is a **simple** and **classic** online problem in computer science

— What if we get a machine learning prediction  $\tilde{n}$  of  $n$ ?

$n$

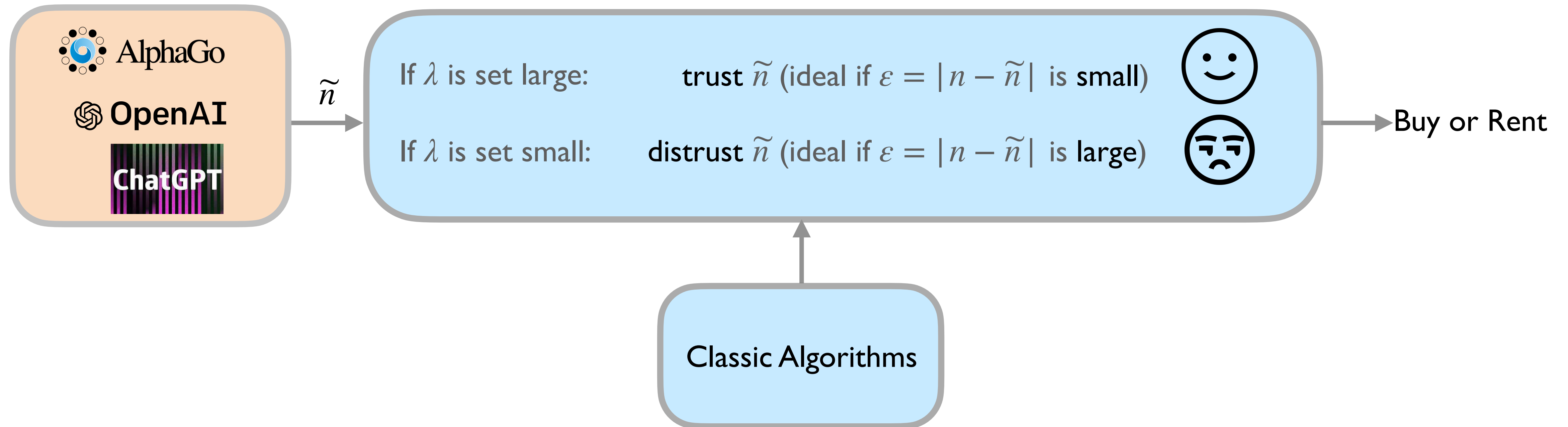
$\tilde{n}$

- $\tilde{n}$  may not be trustworthy
- need to design new **learning-augmented** online algorithms

# Combine Classic Algorithms with Machine Learning Outputs

Learning-Augmented **Ski-Rental** [Kumar et. al. NeurIPS 2018]

Idea: Introduce a trust parameter  $\lambda \in (0,1)$

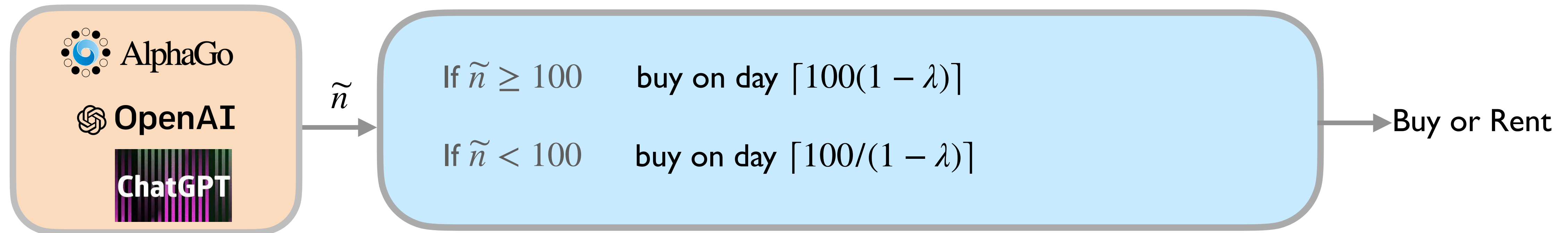




# Combine Classic Algorithms with Machine Learning Outputs

## Learning-Augmented **Ski-Rental** [Kumar et. al. NeurIPS 2018]

Idea: Introduce a trust parameter  $\lambda \in (0,1)$



- Design a *meta-algorithm* based on the classic one with a tuning parameter  $\lambda$
- Varying  $\lambda$  will achieve different results (Robustness and Consistency Trade-off)
- Competitive ratio depends on  $\lambda$

$$\mathbf{CR} \leq \min \left\{ \frac{1 + \lambda}{\lambda}, (1 + \lambda) + \frac{\varepsilon}{(1 - \lambda)\text{OPT}} \right\} \quad \text{OPT} \equiv \min\{100, 10n\}$$

- Both deterministic (above) and stochastic versions
- The bounds are tight [Wei et. al. NeurIPS 2020]



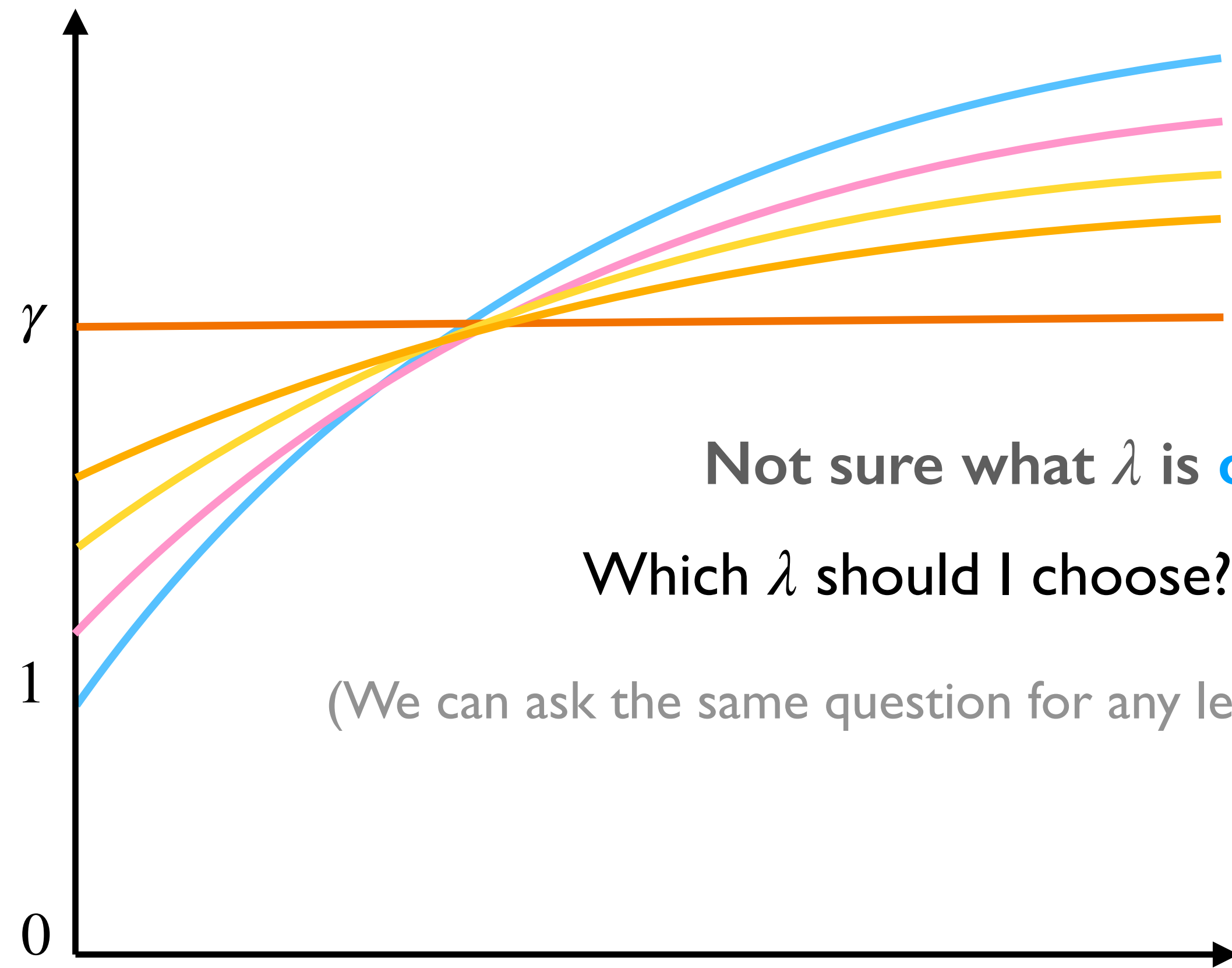


# First Limitation

## General Goal of Learning-Augmented Algorithms

Consistency vs Robustness Trade-off

Competitive ratio  $CR(\varepsilon)$



$\lambda = 1$

$\lambda = 0.7$

$\lambda = 0.5$

$\lambda = 0.2$

$\lambda = 0$

Not sure what  $\lambda$  is **optimal** ...

Which  $\lambda$  should I choose? ( $\varepsilon$  is unknown)

(We can ask the same question for any learning-augmented online algorithms)

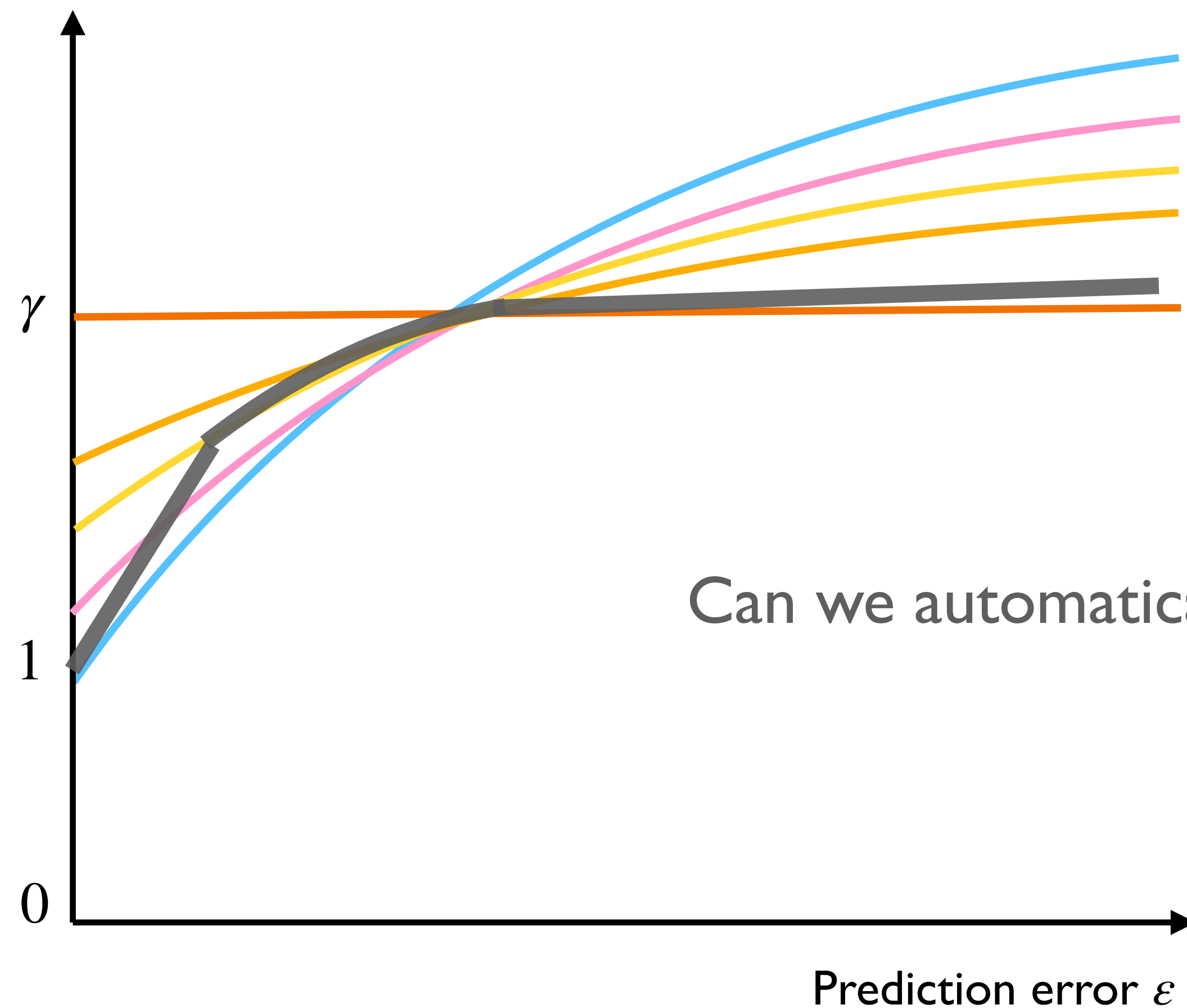
Prediction error  $\varepsilon$

# First Limitation

**Issue:** Prediction error  $\varepsilon$  is not known a priori

**Goal:** Find an online algorithm with good Competitive Ratio **CR** regardless of prediction error  $\varepsilon$

Competitive ratio  $CR(\varepsilon)$



$$\lambda = 1$$

$$\lambda = 0.7$$

$$\lambda = 0.5$$

$$\lambda = 0.2$$

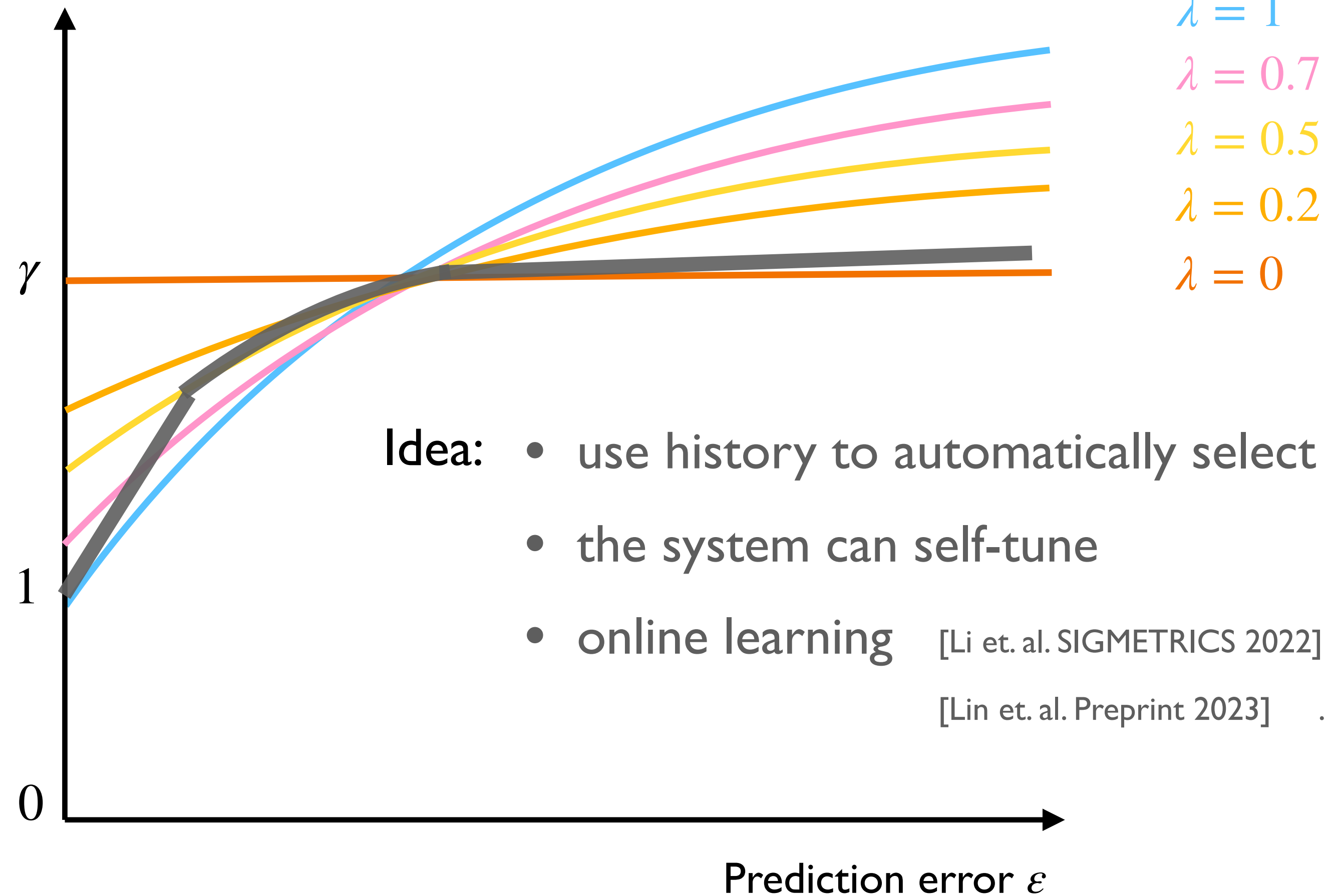
$$\lambda = 0$$

# One Solution: Online Learning

## General Goal of Learning-Augmented Algorithms

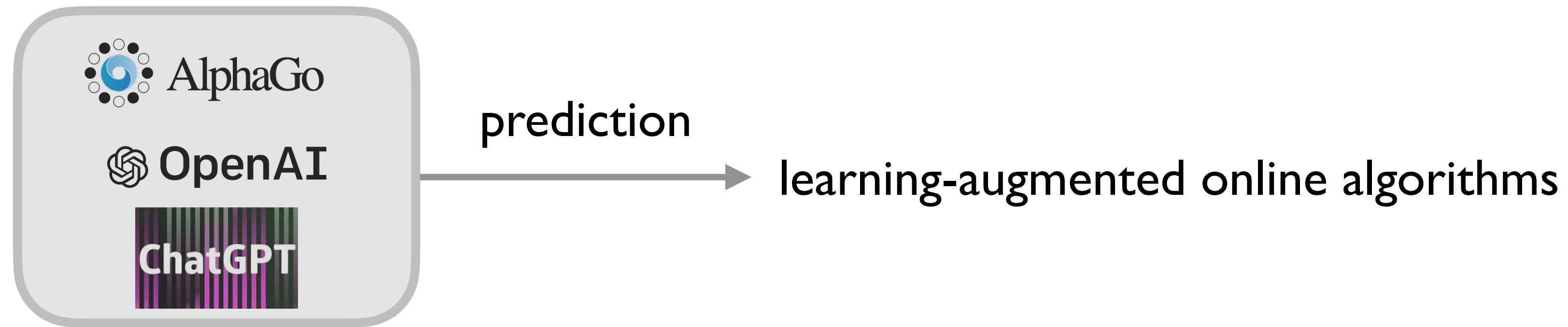
Consistency vs Robustness Trade-off

Competitive ratio  $CR(\varepsilon)$



## Second Limitation

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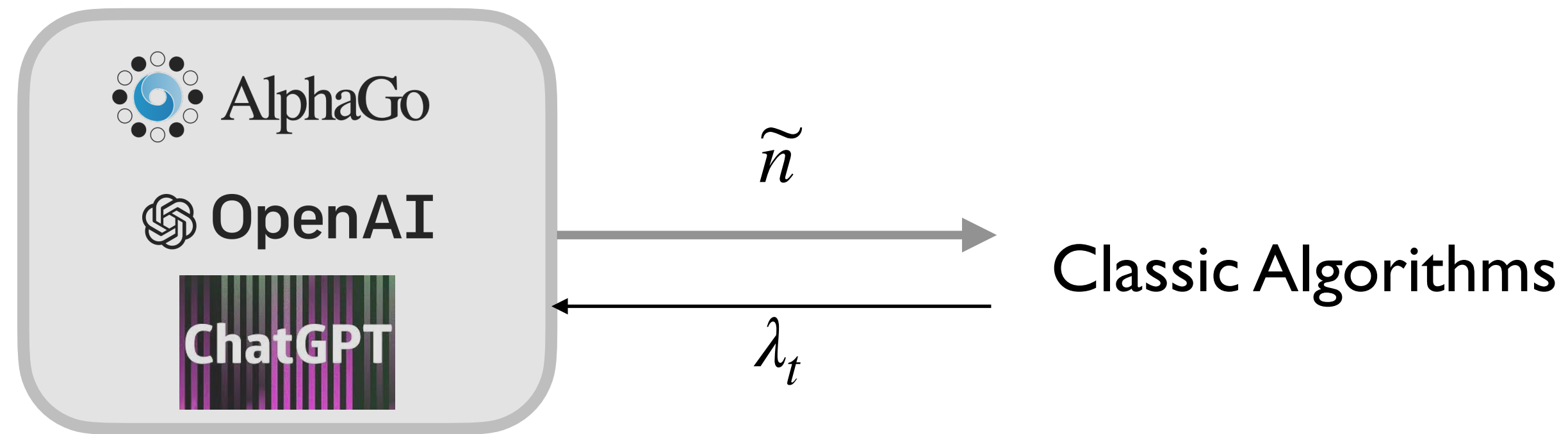
- The machine learning tools are considered as **black-boxes**
- Structural information of the model and ML tools can be helpful
  - specific forms of predictions [Li et. al. SIGMETRICS 2022]
  - grey-box ML models (Q-value functions of value-based policies)

[Li et. al. Preprint 2023]

- can be used to self-tune  $\lambda$  (second solution)

## Second Limitation • Learning-augmented $\rightarrow$ Learning-infused

- Q-learning
- Linear Regression
- Multi-arm bandit



- The machine learning tools are considered as **black-boxes**
- Structural information of the model and ML tools can be helpful
  - specific forms of predictions [Li et. al. SIGMETRICS 2022]
  - grey-box ML models (Q-value functions of value-based policies)

[Li et. al. Preprint 2023]

- can be used to self-tune  $\lambda$  (second solution)

# Learning-Augmented Algorithms

## Online Problems

Ski-rental

Secretary Problem

Online Bipartite Matching

...

## Imperfect Predictions

Number of Skiing Days

Maximum Price

Adjacent Edge-weights

...

[Wei et. al. NeurIPS 2020]

[Purohit et. al. NeurIPS 2018]

[Antoniadis et. al. NeurIPS 2020]

## Black-box AI/ML Advice

Convex Body Chasing

Online Subset Sum

Online Set Cover

Suggested Actions

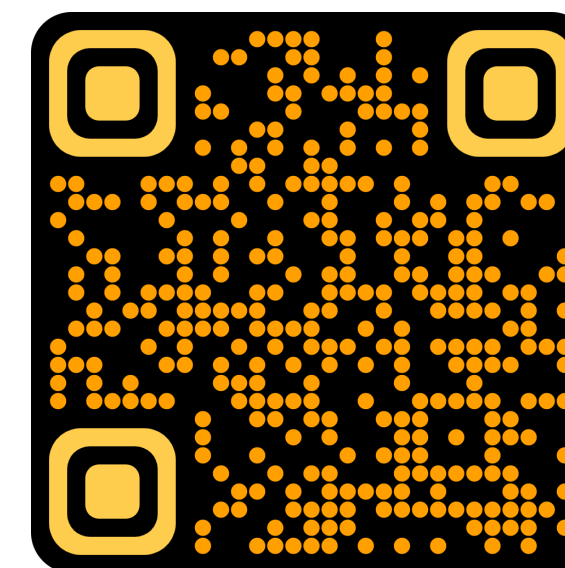
Decision

Predicted Covering

[Christianson et. al. COLT 2022]

[Xu et. al. Journal of Global Optimization 2022]

[Bamas et. al. NeurIPS 2020]



Over 100 topics on this website:

<https://algorithms-with-predictions.github.io/>



# Learning-Augmented Algorithms

## Online Problems

## 不准确预测 Imperfect Predictions

Ski-rental	Number of Skiing Days	[Wei et. al. NeurIPS 2020]	[Purohit et. al. NeurIPS 2018]
Secretary Problem	Maximum Price	[Antoniadis et. al. NeurIPS 2020]	
Online Bipartite Matching	Adjacent Edge-weights		
Linear Quadratic Control	System Perturbations	[Li et. al. SIGMETRICS 2022]	[Li et. al. NeruIPS 2024]

## 不可信AI建议 Black-box AI/ML Advice

Convex Body Chasing	Suggested Actions	[Christianson et. al. COLT 2022]	
Online Subset Sum	Decision	[Xu et. al. Journal of Global Optimization 2022]	
Online Set Cover	Predicted Covering	[Bamas et. al. NeurIPS 2020]	
Q Learning	Q-Value Functions	[Golowich et. al. NeurIPS 2022]	
Value-Based RL	Q-Value Functions (灰盒)/Actions (黑盒)	[Li et. al. NeurIPS 2023]	
Stochastic Game	Type Beliefs	[Li et. al. NeurIPS 2024]	

...

...

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