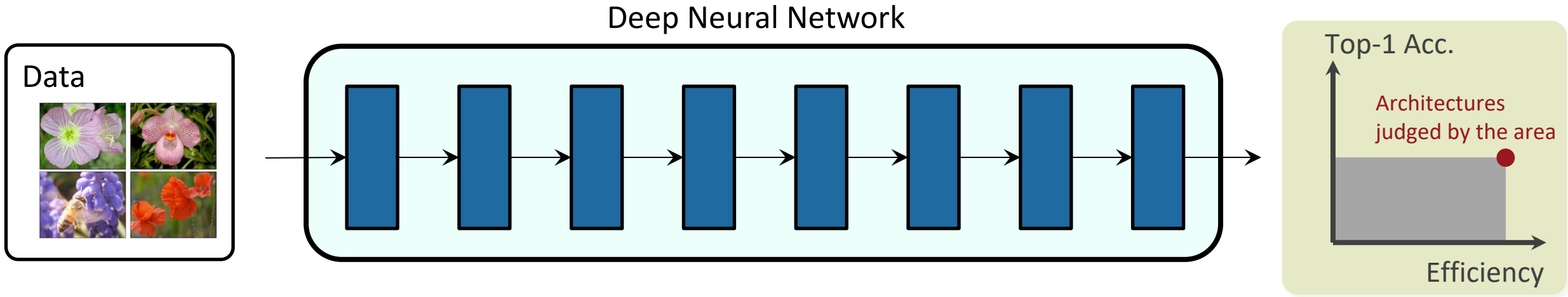
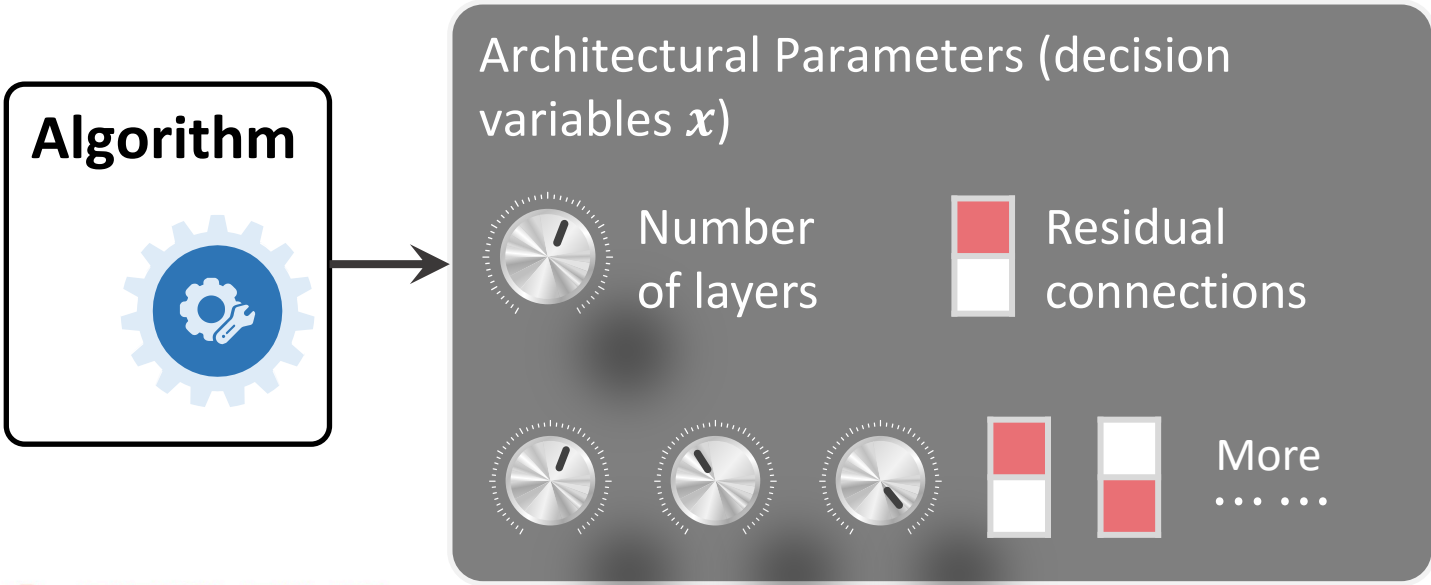


# >>> EvoXBench: Neural Architecture Search as EMO Benchmarks



## Neural Architecture Search (NAS)



$$\min_x F(x) = (f^e(x; \omega^*(x)), f^c(x), f^{\mathcal{H}}(x))$$

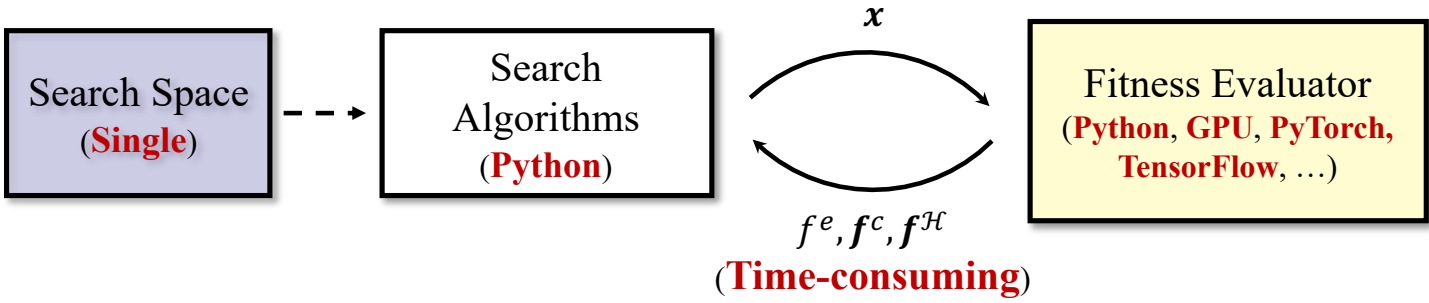
$$\text{s.t. } \omega^* \in \arg \min \mathcal{L}_{trn}(\omega; x), \quad x \in \Omega$$

$$f^c(x): f_1^c(x), f_1^c(x), \dots, f_{M^c}^c(x)$$

$$f^{\mathcal{H}}(x): \begin{cases} f_1^{h_1}(x), \dots, f_{M_1}^{h_1}(x) \\ \vdots \\ f_1^{h_{|\mathcal{H}|}}(x), \dots, f_{M_{|\mathcal{H}|}}^{h_{|\mathcal{H}|}}(x) \end{cases}$$

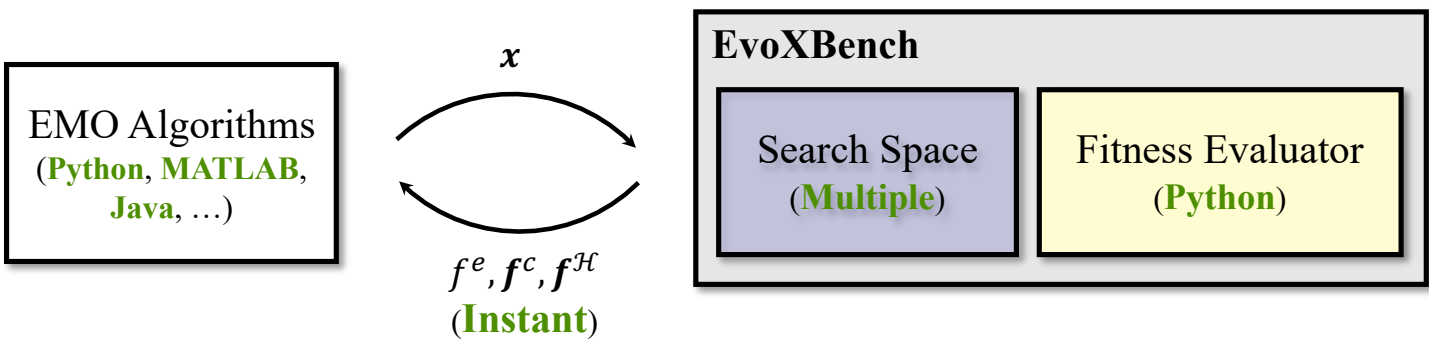
# >>> EvoXBench: A NAS Benchmark Generator Tailored for EMO

## ■ Conventional NAS pipeline:



- Steep computational resources
- One search space at a time
- Python + sophisticated DL software

## ■ EvoXBench pipeline

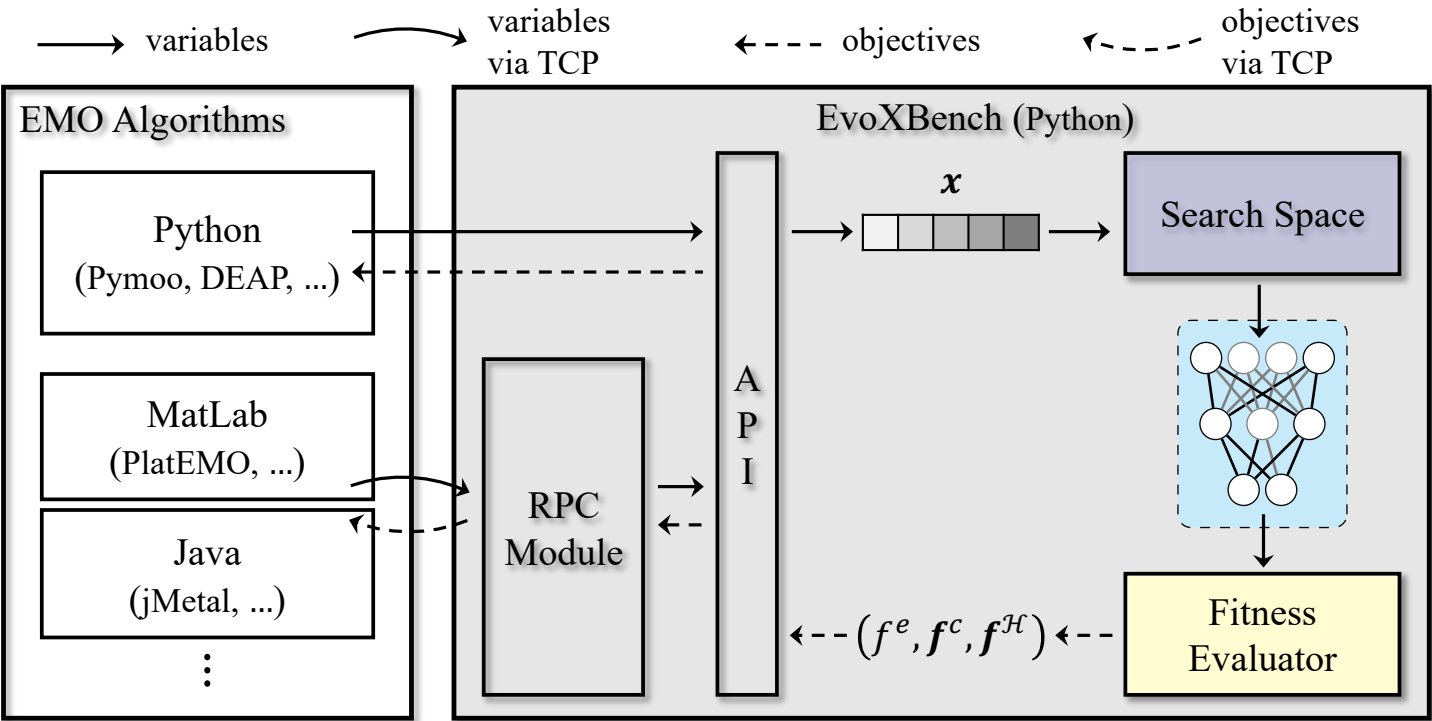


- NO GPU needed
- A unified framework for **multiple search spaces** simultaneously
- Supports **MATLAB, Java, ...**

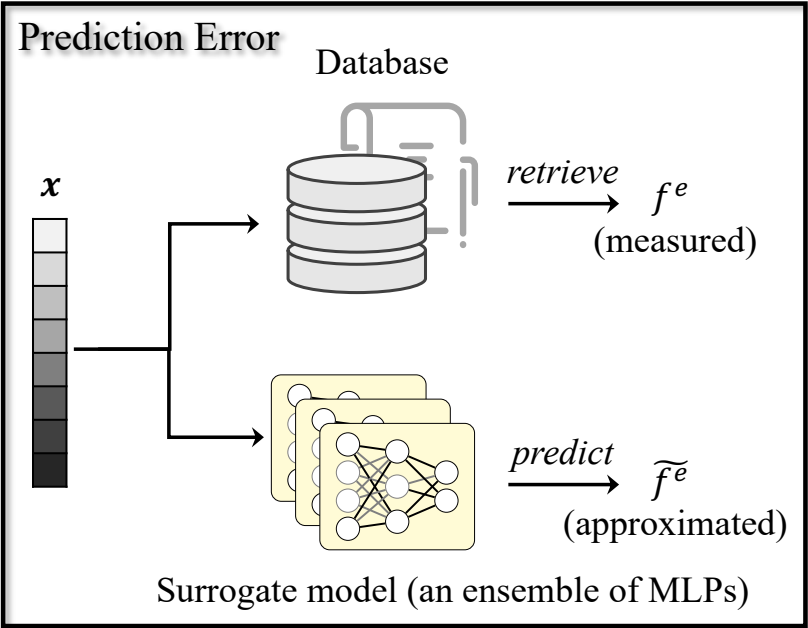
*\* Provides an end-to-end pipeline to generate NAS benchmarks for EMO algorithms to run efficiently*

# >>> EvoXBench: Under the Hood

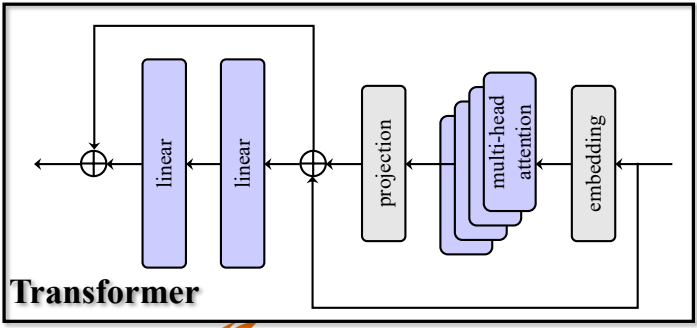
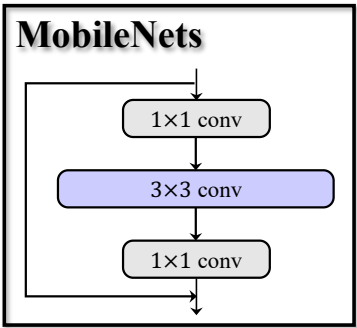
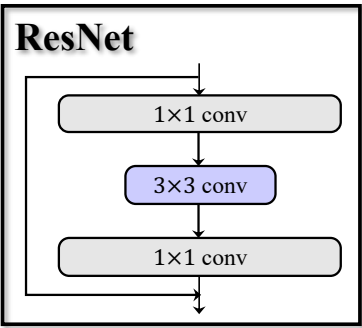
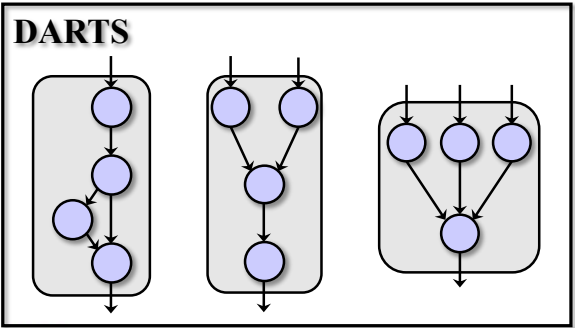
## Overall architecture:



## Fitness Evaluator (Surrogate modeling)

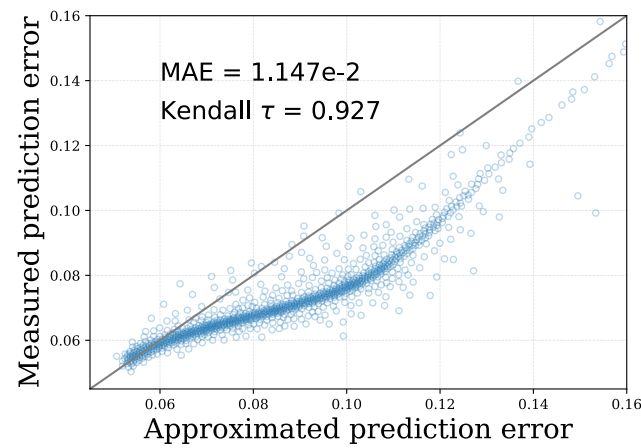


## Search Spaces

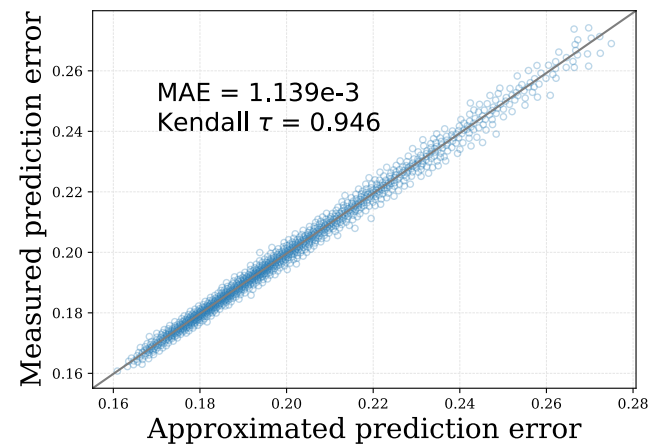


# >>> EvoXBench: Empirical Validation

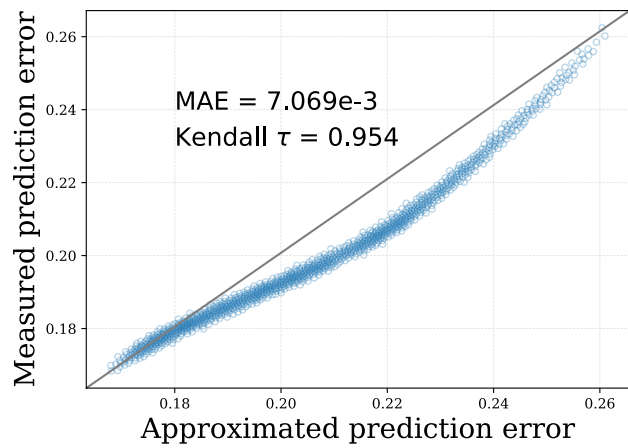
## ■ Accuracy of surrogate modeling:



■ DARTS



■ ResNet



■ MobileNet

## ■ Efficiency (latency in obj. func. evaluation):

Search space	Query method	Python	MATLAB	Java
NASBench101	Database	$0.1139 \pm 0.013$	$0.1716 \pm 0.031$	$0.1970 \pm 0.036$
MobileNets	Surrogate	$0.0380 \pm 0.002$	$0.0528 \pm 0.018$	$0.0574 \pm 0.020$

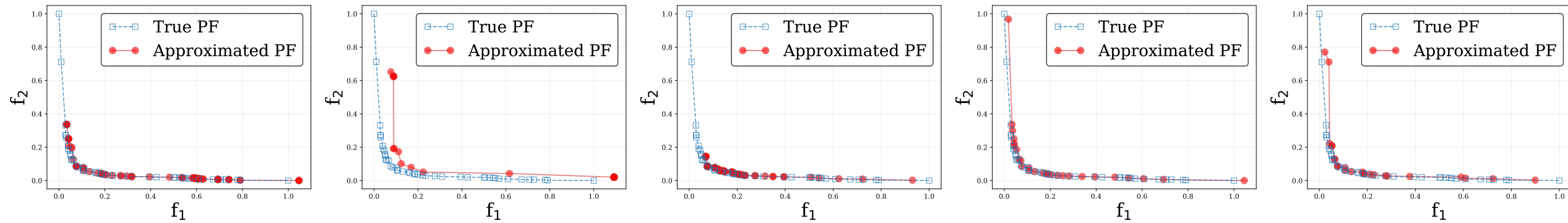
# >>> EvoXBench: Generating NAS Benchmark Test Suites

## ■ C-10/MOP

Problem	$\Omega$	$D$	$M$	Objectives
C-10/MOP1	NB101	26	2	$f^e, f_1^c$
C-10/MOP2	NB101	26	3	$f^e, f_1^c, f_2^c$
C-10/MOP3	NATS	5	3	$f^e, f_1^c, f_2^c$
C-10/MOP4	NATS	5	4	$f^e, f_1^c, f_2^c, f_1^{h1}$
C-10/MOP5	NB201	6	5	$f^e, f_1^c, f_2^c, f_1^{h1}, f_2^{h1}$
C-10/MOP6	NB201	6	6	$f^e, f_1^c, f_2^c, f_1^{h2}, f_2^{h2}, f_3^{h2}$
C-10/MOP7	NB201	6	8	$f^e, f_1^c, f_2^c, f_1^{h1}, f_2^{h1}, f_1^{h2}, f_2^{h2}, f_3^{h2}$
C-10/MOP8	DARTS	32	2	$\dagger f^e, f_1^c$
C-10/MOP9	DARTS	32	3	$\dagger f^e, f_1^c, f_2^c$

## ■ IN-1K/MOP

Problem	$\Omega$	$D$	$M$	Objectives
IN-1K/MOP1	ResNet50	25	2	$f^e, f_1^c$
IN-1K/MOP2	ResNet50	25	2	$f^e, f_2^c$
IN-1K/MOP3	ResNet50	25	3	$f^e, f_1^c, f_2^c$
IN-1K/MOP4	Transformer	34	2	$f^e, f_1^c$
IN-1K/MOP5	Transformer	34	2	$f^e, f_2^c$
IN-1K/MOP6	Transformer	34	3	$f^e, f_1^c, f_2^c$
IN-1K/MOP7	MNV3	21	2	$f^e, f_1^c$
IN-1K/MOP8	MNV3	21	3	$f^e, f_1^c, f_2^c$
IN-1K/MOP9	MNV3	21	4	$f^e, f_1^c, f_2^c, f_1^{h1}$



## ■ C-10/MOP1: NSGA-II, IBEA, MOEA/D, NSGA-III, RVEA

# >>> EvoXBench: How to Get Started

■ Paper



■ Tutorial



■ Code



■ Support

