



# A Large-scale Benchmark for Airfoil Design

Jian Liu<sup>1,2</sup>, **Jianyu Wu**<sup>2,\*</sup>, Hairun Xie<sup>3</sup>, Guoqing Zhang<sup>1,2</sup>, Jing Wang<sup>3</sup>, Wei Liu<sup>3</sup>, Ouyang Wanli<sup>2</sup>, Junjun Jiang<sup>1</sup>, Xianming Liu<sup>1</sup>, Shixiang Tang<sup>2</sup>, Miao Zhang<sup>3</sup>

<sup>1</sup>Harbin Institute of Technology, <sup>2</sup>Shanghai AI Laboratory <sup>3</sup>Shanghai Aircraft Design and Research Institute \***Presenting** 

\*Presenting







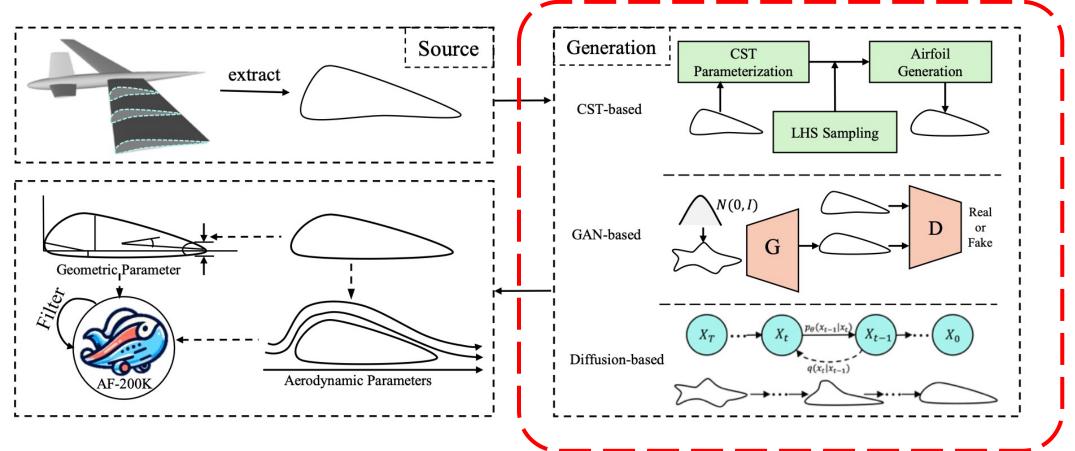
> The existing airfoil datasets are relatively **small-scale**.

- ➤ The current datasets typically provide only a single condition.
  > The current methods do not support progressive aditing
- The current methods do not support progressive editing

existing designs.

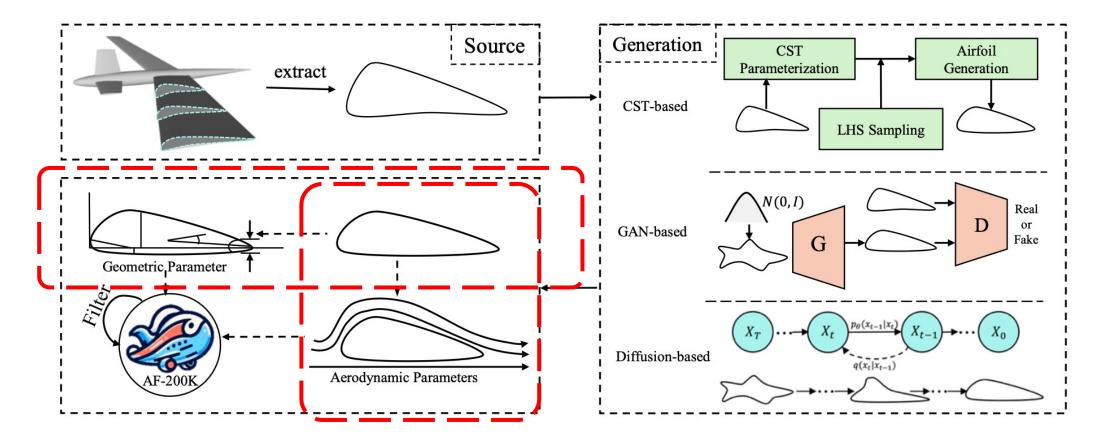
➤Lack of a comprehensive and clear codebase to compare and analyze different approaches.

## **Automatic Data Engine**



- 1. Generation stage:
  - CST-assisted Generation
  - Unconditional Airfoil Generation

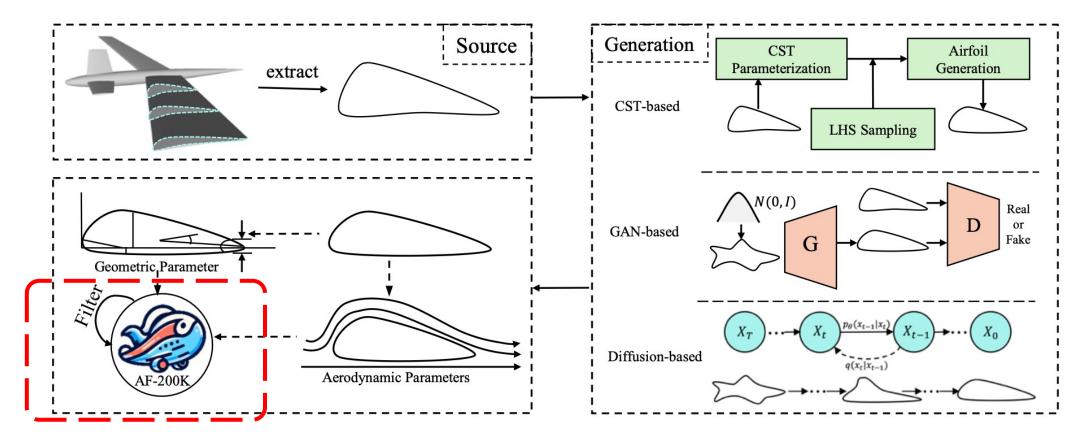
## **Automatic Data Engine**



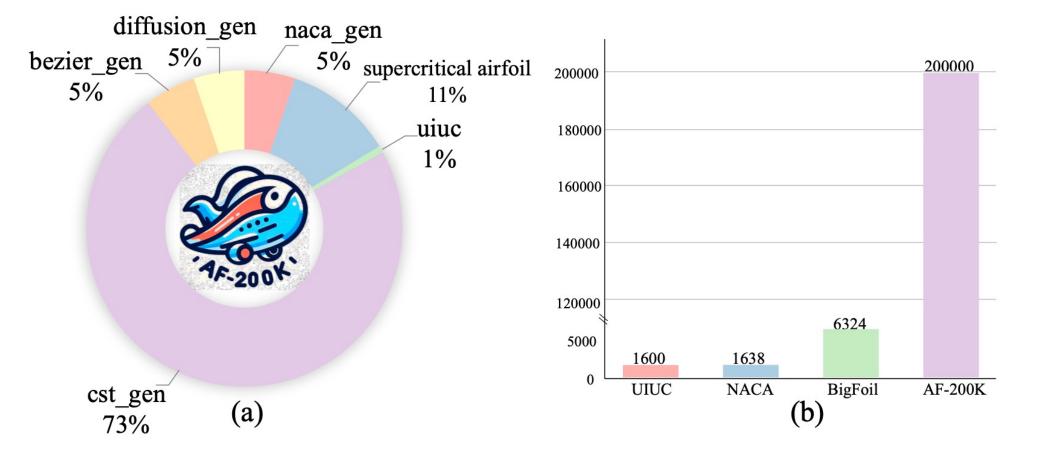
#### 2. Annotation stage:

- Aerodynamic
- Geometric

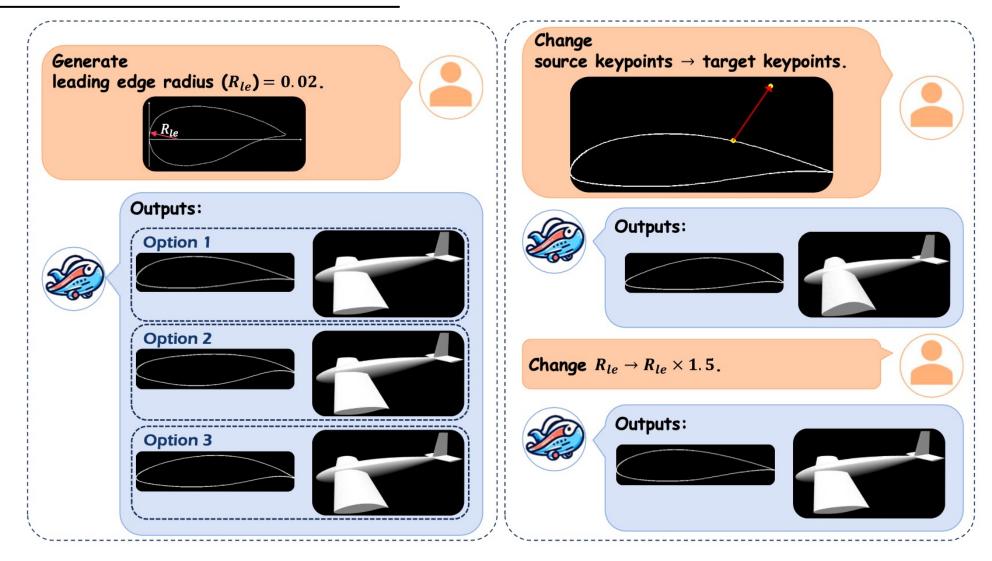
#### **Automatic Data Engine**



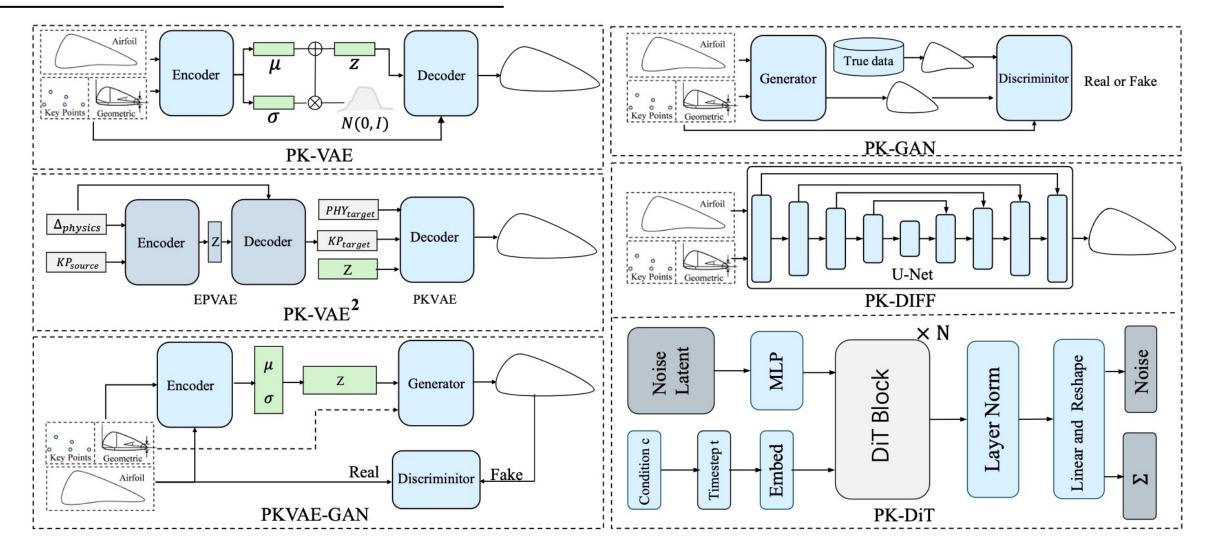
3. Filtering stage



Dataset presentation



Two Airfoil Inverse Design Tasks



The baseline methods for benchmarking the dataset

Comprehensive metrics:

• Label error: 
$$\sigma_i = |\hat{p}_i - p_i|, i = 1, 2, ..., 11$$

• Diversity: 
$$\mathcal{D} = \frac{1}{n} \sum_{i=1}^{n} \log det(\mathcal{L}_{S_i}),$$
  
• Smoothness:  $\mathcal{M} = \sum_{i=1}^{N} \text{Distance}_{Pn \perp |P_{n-1}P_{n+1}|},$   
• Success rate:  $\mathcal{R} = \frac{1}{N} \sum_{i=1}^{N} \mathbb{I}(\frac{\sum_{j=1}^{M} C_j}{M} > 60\%), j = 1, ..., M,$ 

Method	Dataset		Label error $\downarrow \times 0.01$												$\mathcal{M}\downarrow  imes 0.01$
		$\sigma_1$	$\sigma_2$	$\sigma_3$	$\sigma_4$	$\sigma_5$	$\sigma_6$	$\sigma_7$	$\sigma_8$	$\sigma_9$	$\sigma_{10}$	$\sigma_{11}$	$\bar{\sigma}$	$\mathcal{D}\uparrow$	
CVAE [16]	AF-200K	7.29	5.25	3.52	1590	9.9	9.55	2900	1.91	1.53	4.6	10.4	413.1	-155.4	7.09
CGAN [15]	AF-200K	10.7	8.50	5.44	2320	14.3	13.7	5960	2.53	2.23	5.3	12.9	759.6	-120.5	7.31
PK-VAE	AF-200K	6.30	4.79	3.13	862	6.6	6.41	1710	1.35	0.93	3.3	7.8	237.5	-150.1	5.93
PK-GAN	AF-200K	8.18	6.30	4.70	2103	12.0	11.7	3247	2.25	1.96	5.0	12.7	492.3	-112.3	3.98
<b>PKVAE-GAN</b>	AF-200K	5.68	3.17	3.10	565	4.6	4.35	1200	0.91	0.51	2.8	6.3	163.3	-129.6	2.89
PK-DIFF	AF-200K	4.61	3.46	2.15	277	2.2	1.93	1030	0.70	0.11	2.4	3.1	120.6	-101.3	1.52
PK-DIT	UIUC	6.38	5.14	3.36	1183	8.7	8.49	2570	1.69	1.19	3.6	9.8	345.6	-141.7	6.03
PK-DIT	Super	5.20	3.50	2.40	301	2.9	3.32	1050	0.83	0.26	2.7	3.3	125.0	-123.4	1.97
PK-DIT	AF-200K	1.12	3.23	1.54	105	1.3	1.15	979	0.05	0.05	2.3	2.4	<b>99.7</b>	-93.2	1.04

Performance across different datasets

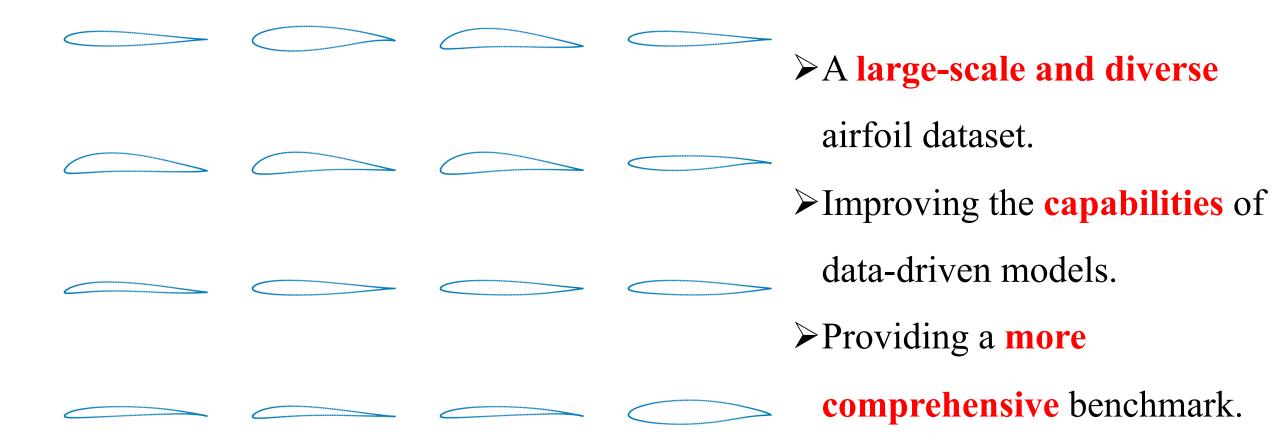
Method	Task	Label error $\downarrow \times 0.01$													$\mathcal{M}\downarrow  imes 0.01$
		$\sigma_1$	$\sigma_2$	$\sigma_3$	$\sigma_4$	$\sigma_5$	$\sigma_6$	$\sigma_7$	$\sigma_8$	$\sigma_9$	$\sigma_{10}$	$\sigma_{11}$	$\bar{\sigma}$		
PK-VAE	EK	9.3	8.33	5.27	2082	12.9	11.1	4620	2.51	2.04	5.1	11.8	615.5	-143.4	7.21
PK-VAE	EP	8.9	6.38	4.94	1780	10.9	9.4	4570	2.05	1.98	4.9	10.3	582.6	-150.8	7.19
PK-VAE <sup>2</sup>	EK	7.1	5.71	4.05	1430	8.0	8.1	3780	1.91	1.52	3.6	8.7	478.1	-133.4	6.20
$PK-VAE^2$	EP	6.5	5.22	3.57	1010	7.8	7.3	2010	1.52	1.03	3.4	7.9	278.5	-135.6	6.36

#### Performance across different design tasks

Method	Label Error $(\times 0.01) \downarrow$												$\mathcal{D}^{+}$	$\mathcal{M}\downarrow  imes 0.01$
	$\sigma_1$	$\sigma_2$	$\sigma_3$	$\sigma_4$	$\sigma_5$	$\sigma_6$	$\sigma_7$	$\sigma_8$	$\sigma_9$	$\sigma_{10}$	$\sigma_{11}$	$\bar{\sigma}$		
NACA-GEN	6.26	5.10	3.29	961	7.69	7.46	2130	1.08	1.038	3.4	8.0	284.9	-136.4	5.09
CST-GEN	5.82	4.09	2.80	572	4.61	4.36	1390	0.94	0.542	3.1	5.9	181.3	-101.5	2.31
BézierGAN-GEN	5.96	4.96	3.07	839	5.64	6.38	1900	0.98	0.817	3.1	7.4	252.5	-125.3	1.21
<b>Diffusion-GEN</b>	5.44	3.83	2.58	353	3.09	3.33	<b>1180</b>	0.89	0.293	2.9	4.2	141.8	-111.9	2.05

Performance across different generated data

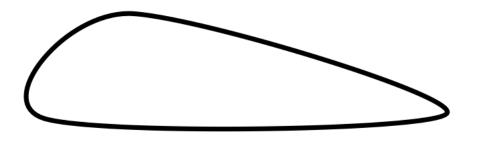
#### Visualization

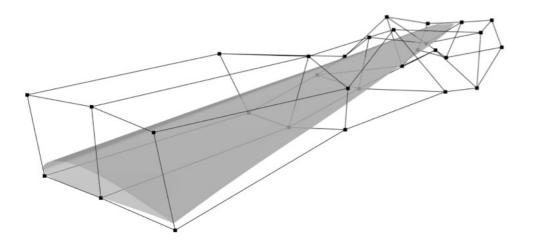


Diverse airfoils generated by PK-DIT

## **Limitation & Future works**

- Dealing with multiple conditions
- Optimization techniques integration
- Dimension extension





#### 2D airfoil inverse design



3D airfoil inverse design

**Deep Generative Model for Efficient 3D Airfoil Parameterization and Generation** Wei Chen and Arun Ramamurthy Arxiv, 2021



# Github Zhihu Arxiv

# Thank you!