
Oral presentation | Fluid-structure interaction
Fluid-structure interaction-III
Mon. Jul 15, 2024 4:30 PM - 6:30 PM Room A

[3-A-04] Data-Driven Fluid-Structure Interaction with Fully-Partitioned Method and Deep Koopman Model

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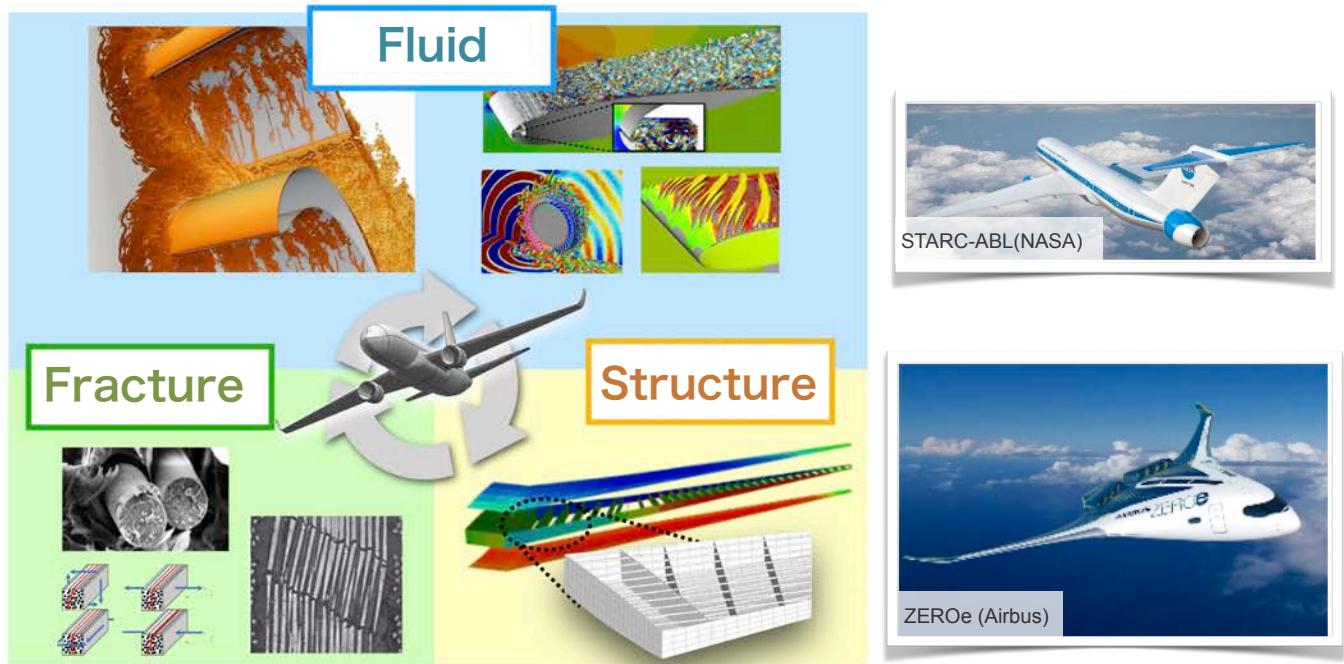
Keywords: Fluid-structure interaction, Koopman, Data-driven approach

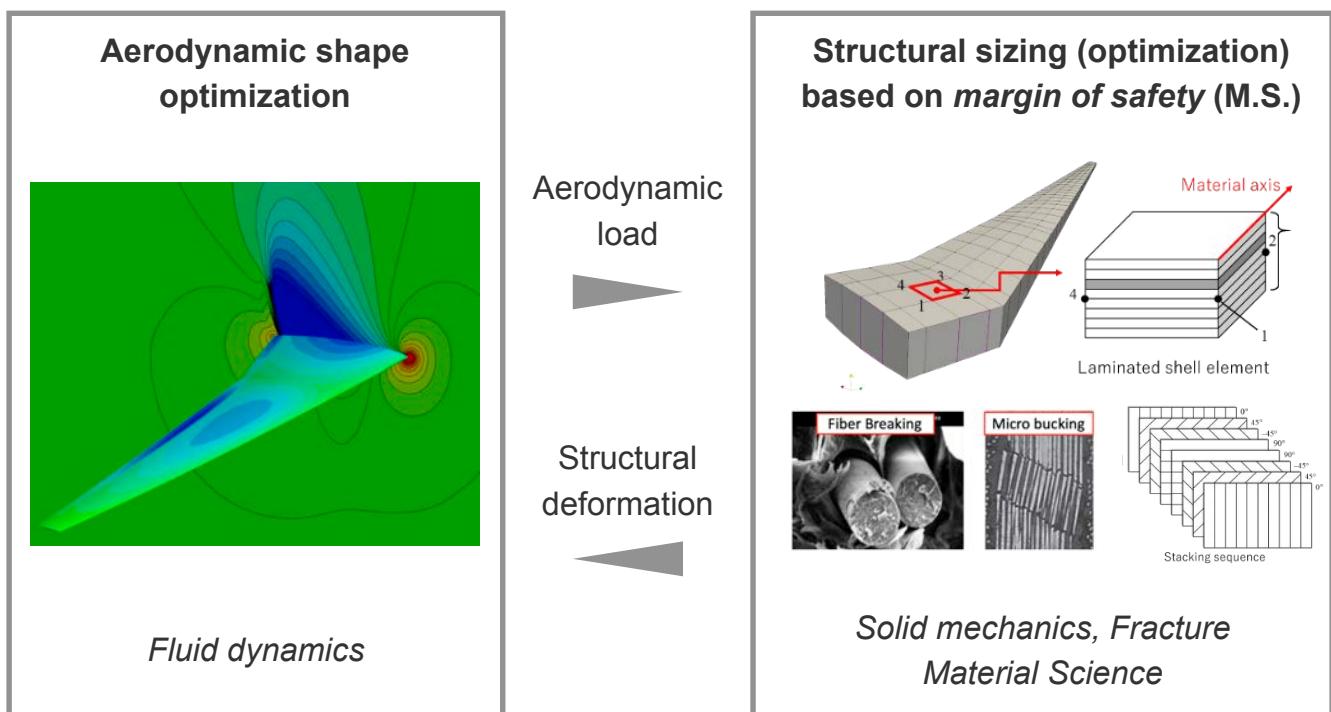
Data-Driven Fluid-Structure Interaction with Fully-Partitioned Method and Deep Koopman Model

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Green Innovation for Global Aviation

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Digitalization of integrated design process with multi-physics simulation

Contents

Residual minimization for static FSI

Abe et al., JSCM, 48, 6 (2022)

The graph shows the residual force decreasing as the deformation parameter q_m increases, with green circles representing parallel simulations by FFP. A red triangle marks the minimum point. The flowchart details the iterative process: Assume deformation → Evaluate residual force (three parallel steps) → Find deformation to minimize residual.

Data-driven dynamic FSI

Yamazaki et al., APS-DFD 2023

This diagram illustrates a data-driven dynamic FSI model. It shows an encoder taking data $D = X \cup Y$ and producing hidden states \tilde{Y} and \tilde{X} . The encoder also generates an equation $A = (\tilde{Y} - B\Gamma)\tilde{X}^\dagger$. An external matrix Γ provides inputs $[u_1, u_2, \dots, u_T]$. The hidden states \tilde{Y} and \tilde{X} are processed by a decoder to produce predicted outputs \hat{Y}_{prd} and \hat{X} .

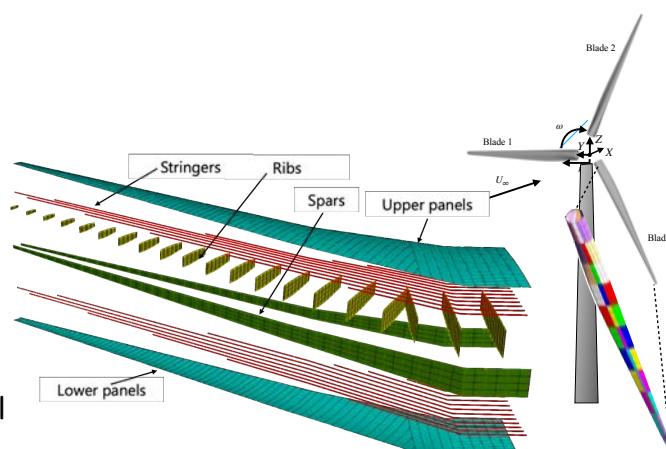
Aircraft design simulator

DASH

Digital Aircraft Design tool of Tohoku University

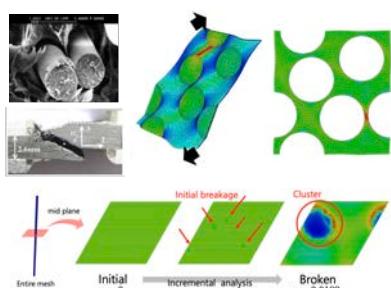
Date, Abe, et al., J. Aero. Sci. Tech. 2022

Abe, et al., CM3-ECCOMAS (invited, 2023)

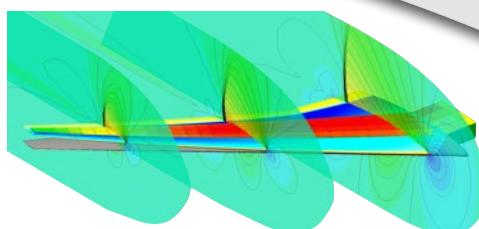


Microscale analysis for composite material

[$\mu\text{m} - \text{mm}$]

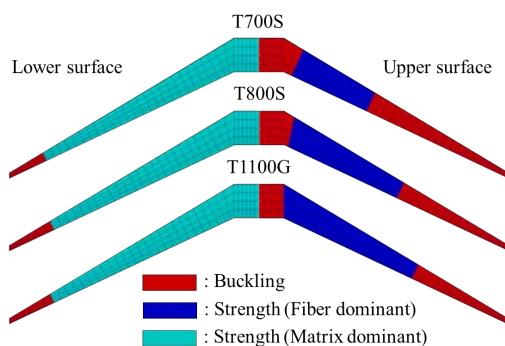


Macroscale aeroelastic / design analysis [m]

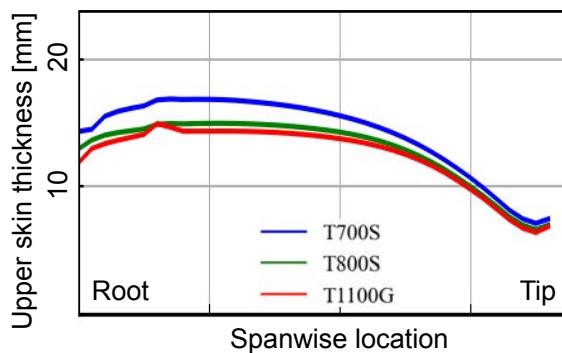
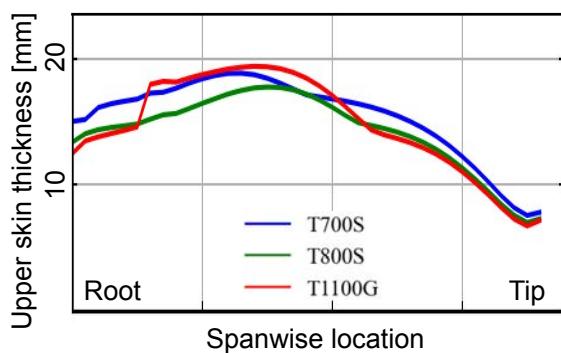
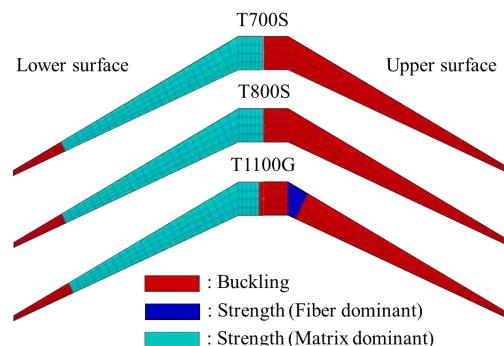


Aerostructural design in equilibrium condition 6

One-way (CFD → CSD)



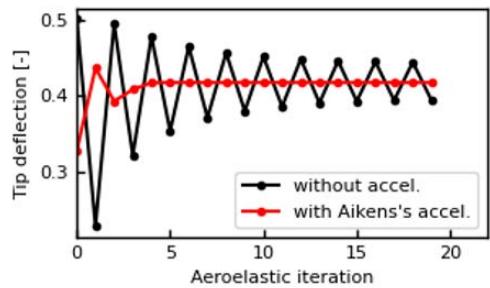
Two-way (CFD ⇌ CSD)



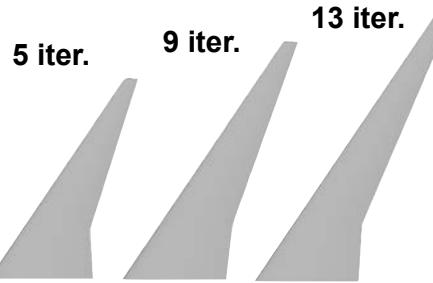
- Failure mode cannot be accurately captured by one-way coupled analysis

Sequential iteration

- ✓ 5 - 20 times aeroelastic iteration
- ✓ 50 - 200 times structural-sizing iter.

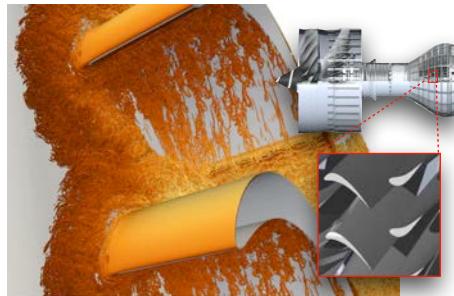


- ✓ High-aspect-ratio wing requires larger iter.



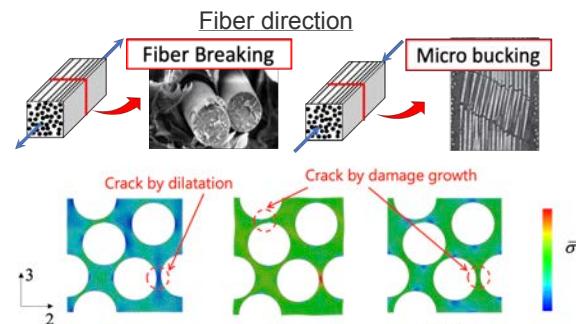
High-fidelity models

- ✓ CFD at real high-load conditions



A.S. Iyer, Y. Abe et al., *C&F*, 226, 104989, (2021)

- ✓ CSD with damage propagation



Conventional iterative method for static FSI

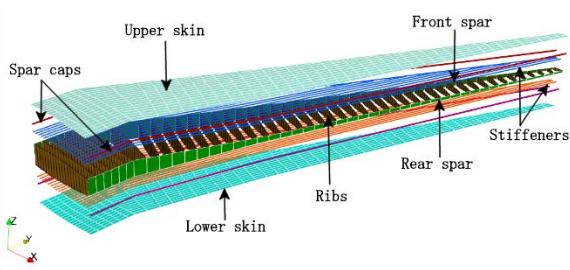
- Structural deformation in equilibrium condition

$$\lambda_S(\mathbf{u}_S) - f_S(\mathbf{u}_S) = 0 \quad (\mathbf{u}_S : \text{displacement})$$

Internal force

$$\lambda_S = \underline{\mathcal{S}^*(\mathbf{u}_S)}$$

Structural solver

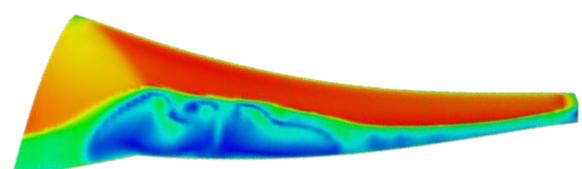


External force

$$f_S = \underline{T^T \mathcal{A}(\mathbf{x}_{Av}^{(0)}, \underline{T u}_S)}$$

Flow solver

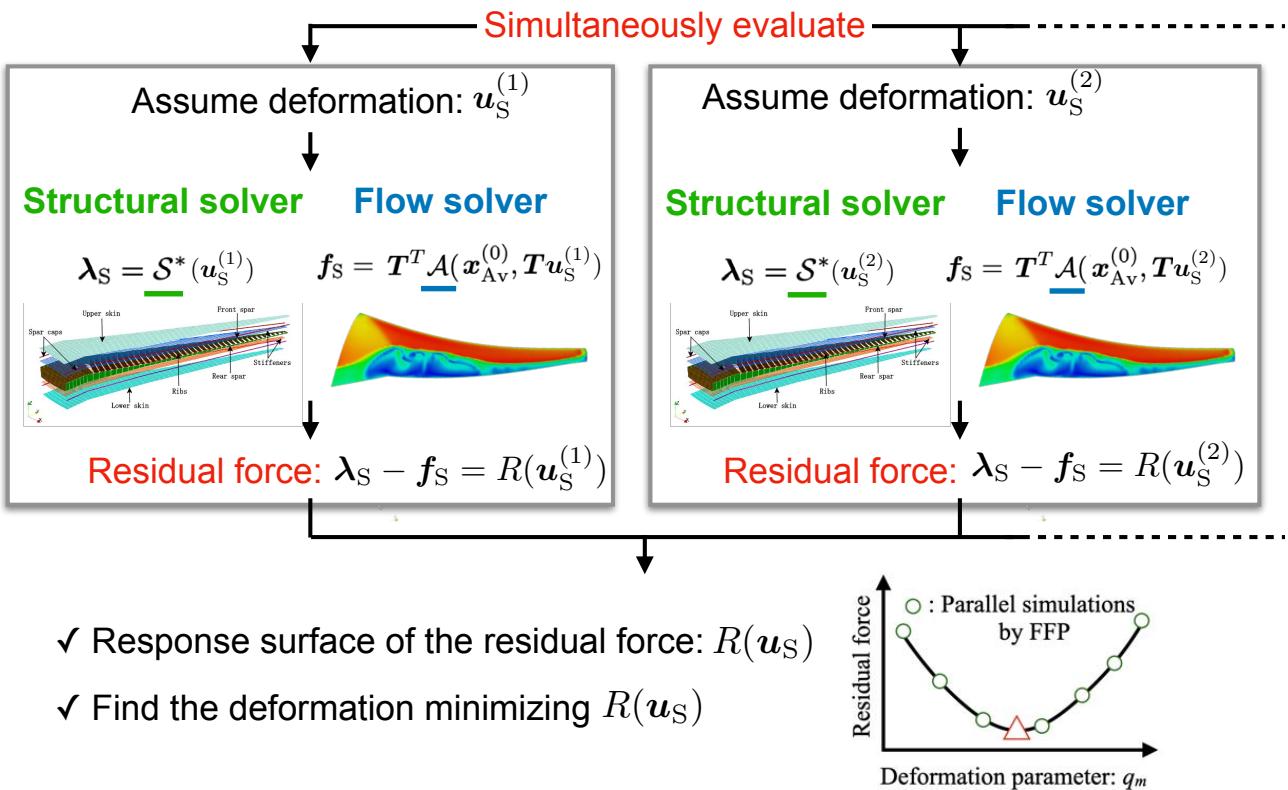
Transformation matrix



Residual minimization approach in static FSI

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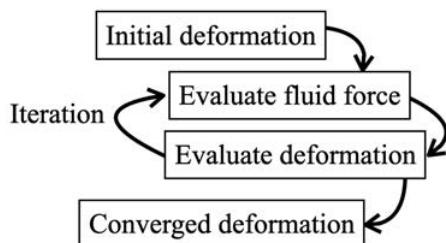
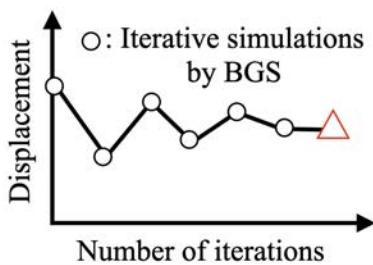
► Force-based fully partitioned (FFP)



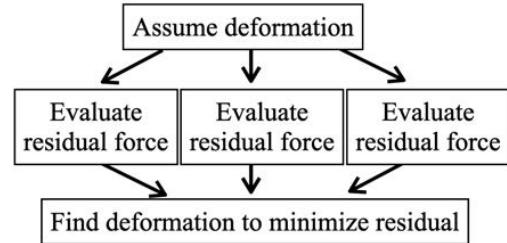
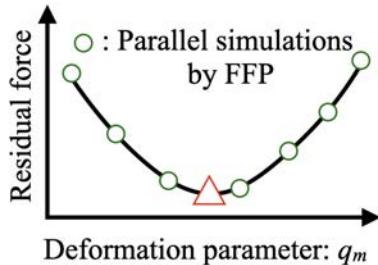
Residual minimization approach in static FSI

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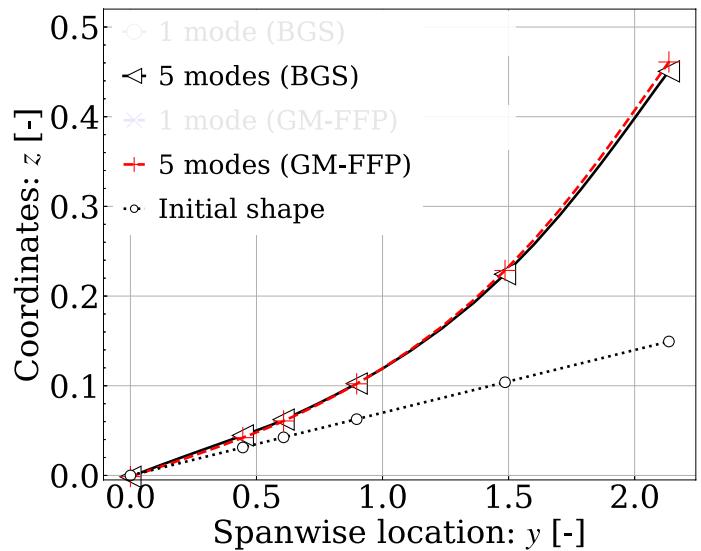
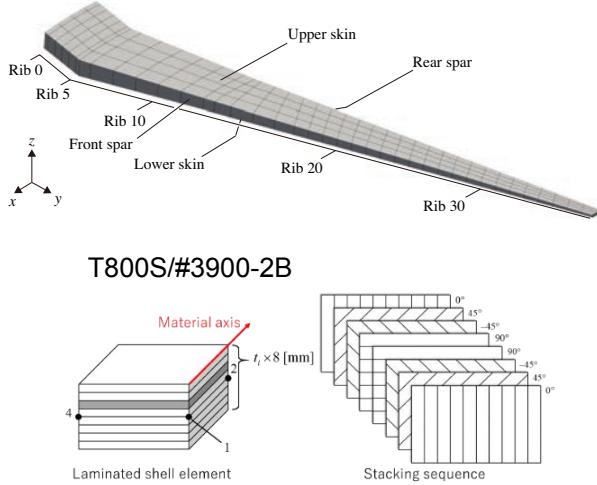
► Nonlinear Block Gauss-Seidel (BGS)



► Force-based fully partitioned (FFP)



► Leading edge deformation of CFRP wing



✓ GM-FFP is **highly parallelizable** in iterative direction.

=> It becomes faster if the iteration of BGS is large.

BGS v.s. GM-FFP (AoA = 0 deg.)

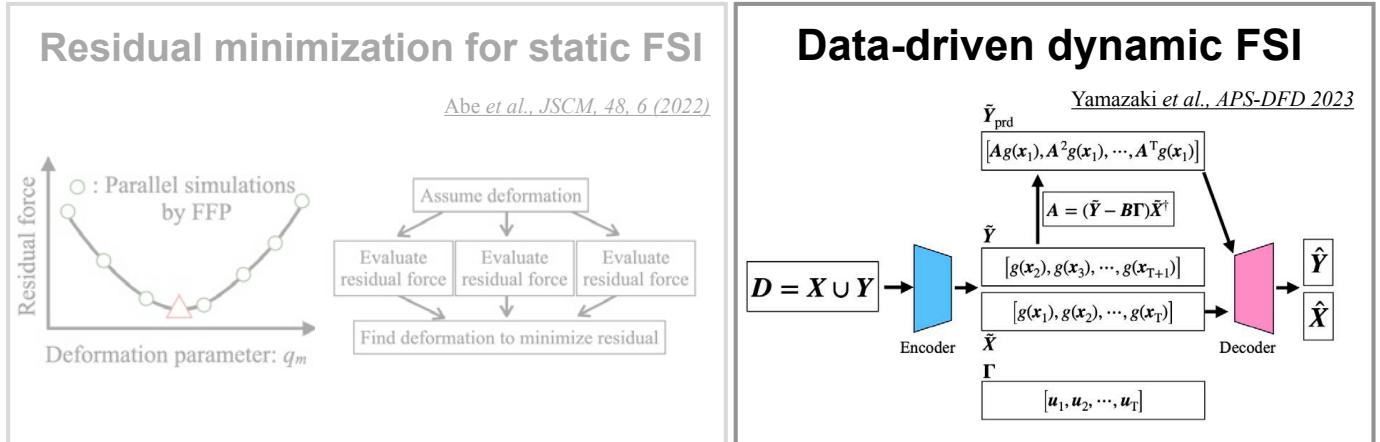
► Elapsed time with CPU (4 cores for 1 simulation)

Method	Cores	Elapsed time
BGS (1 mode, 4 iter.)	4	302.9 sec.
BGS (5 modes, 6 iter.)	4	459.1 sec.
GM-FFP (1 mode)	20	74.69 sec.
GM-FFP (5 modes)	20	373.44 sec.

approx. 4 times speed-up

✓ GM-FFP is parallelizable in iterative direction.

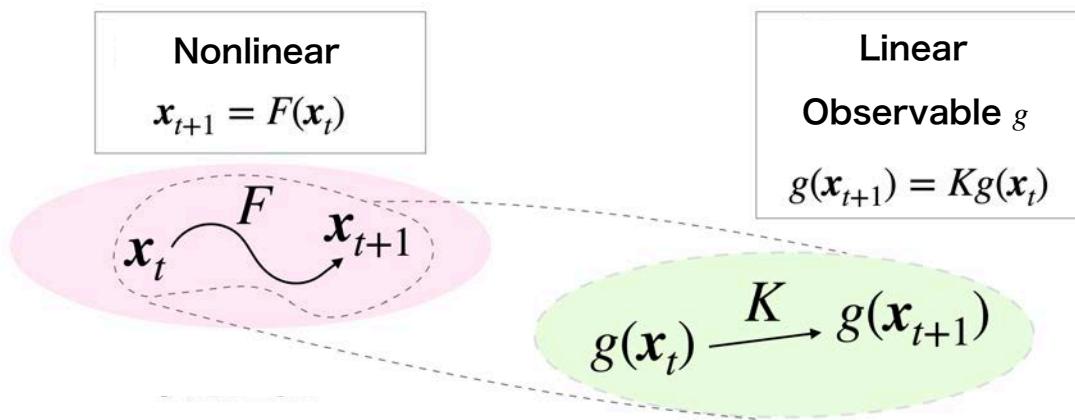
=> It becomes faster if the iteration of BGS is large.



Deep Koopman dynamical model

Morton et al., NeurIPS 2018; Takeishi et al., NeurIPS 2017

$$x_{t+1} = F(x_t) \longrightarrow g(x_{t+1}) = g(F(x_t)) = \mathcal{K}g(x_t)$$



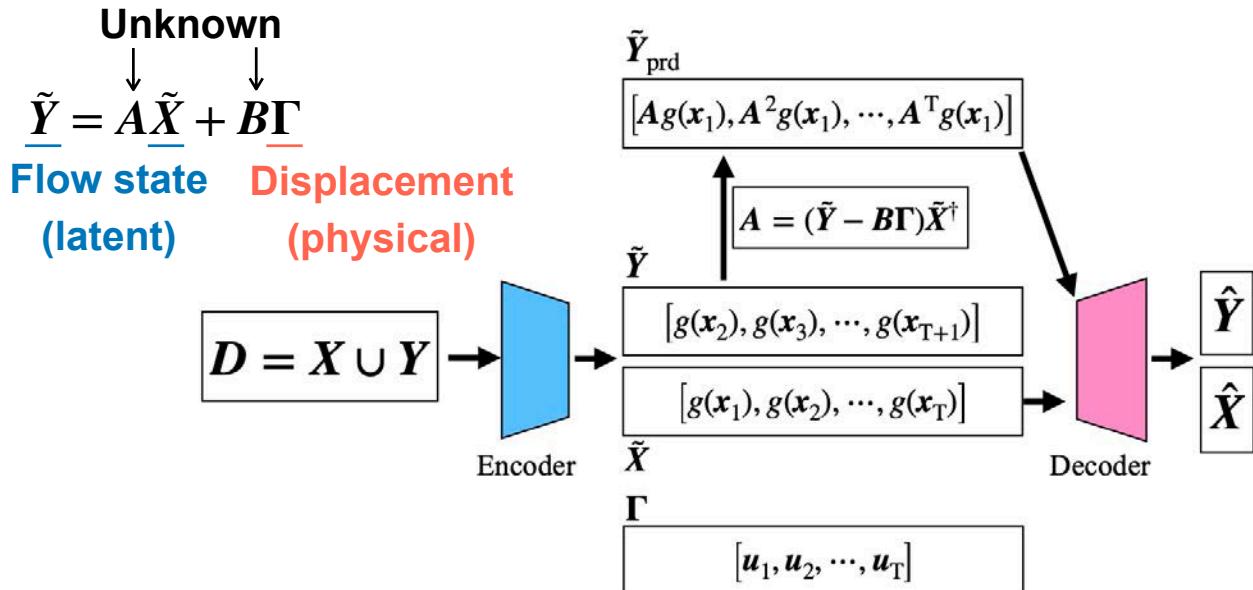
G
(Koopman Invariant Subspace)

Deep Koopman dynamical model

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Morton et al., NeurIPS 2018; Takeishi et al., NeurIPS 2017

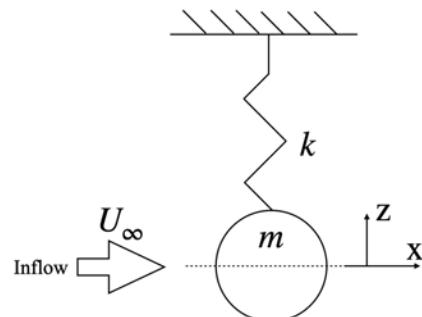
$$x_{t+1} = F(x_t) \longrightarrow g(x_{t+1}) = g(F(x_t)) = \mathcal{K}g(x_t)$$



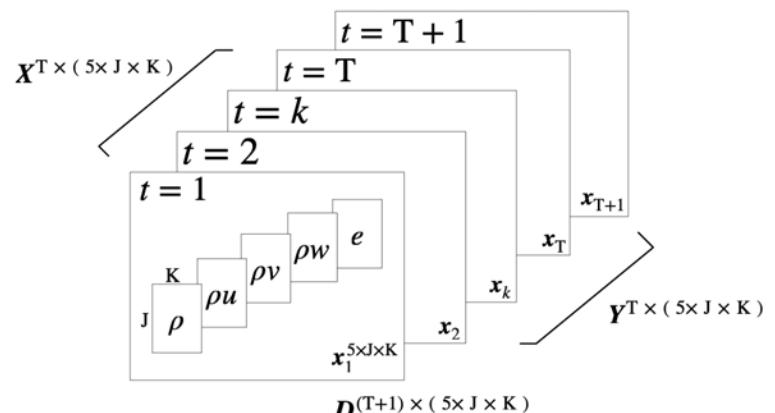
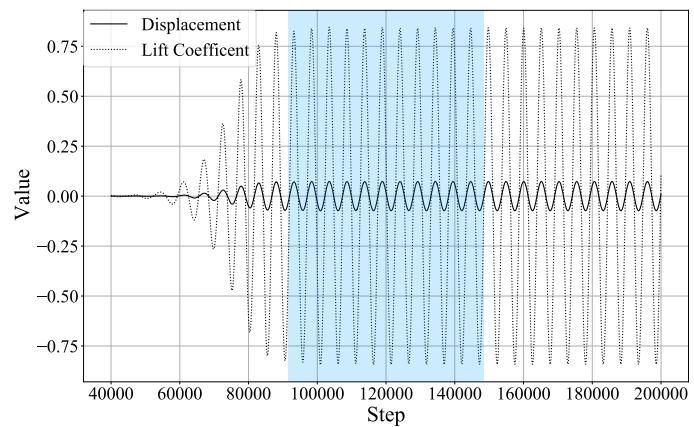
Deep Koopman dynamical model

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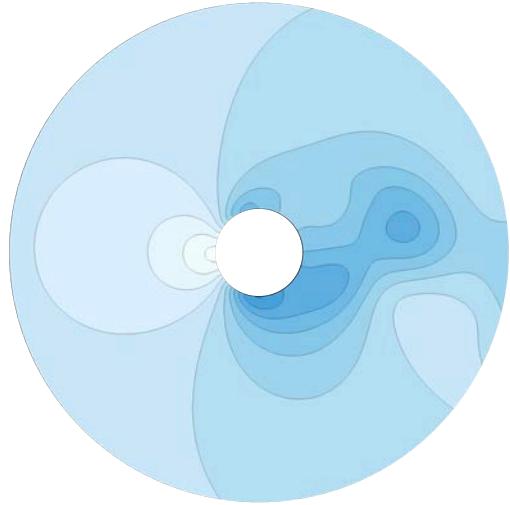
- Flow around cylinder



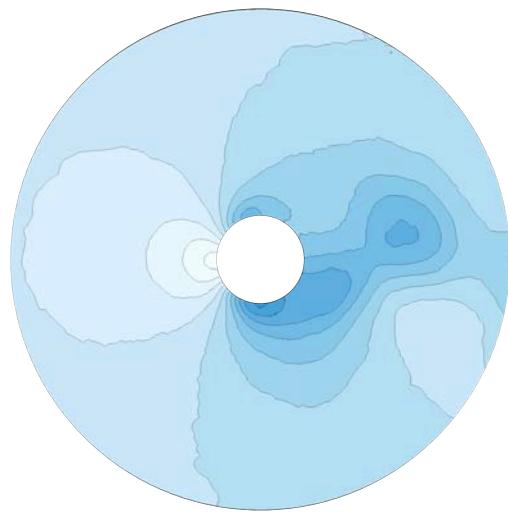
- Flow variables: X, Y
input: $u = [z, \dot{z}]^T$



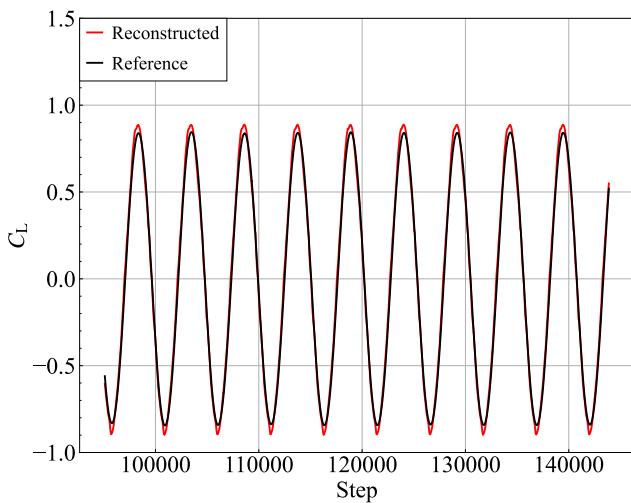
Original (pressure field)



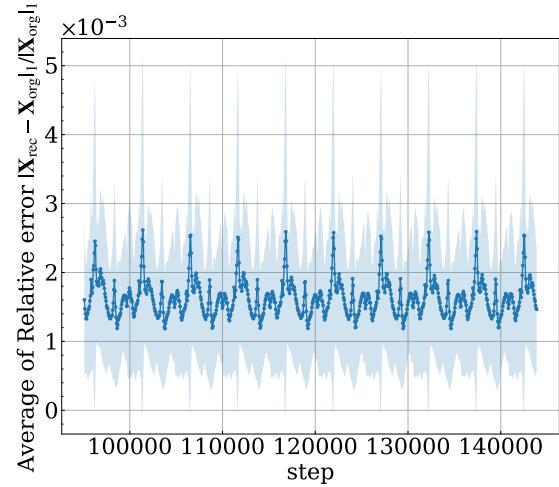
Reconstructed



- Lift force history



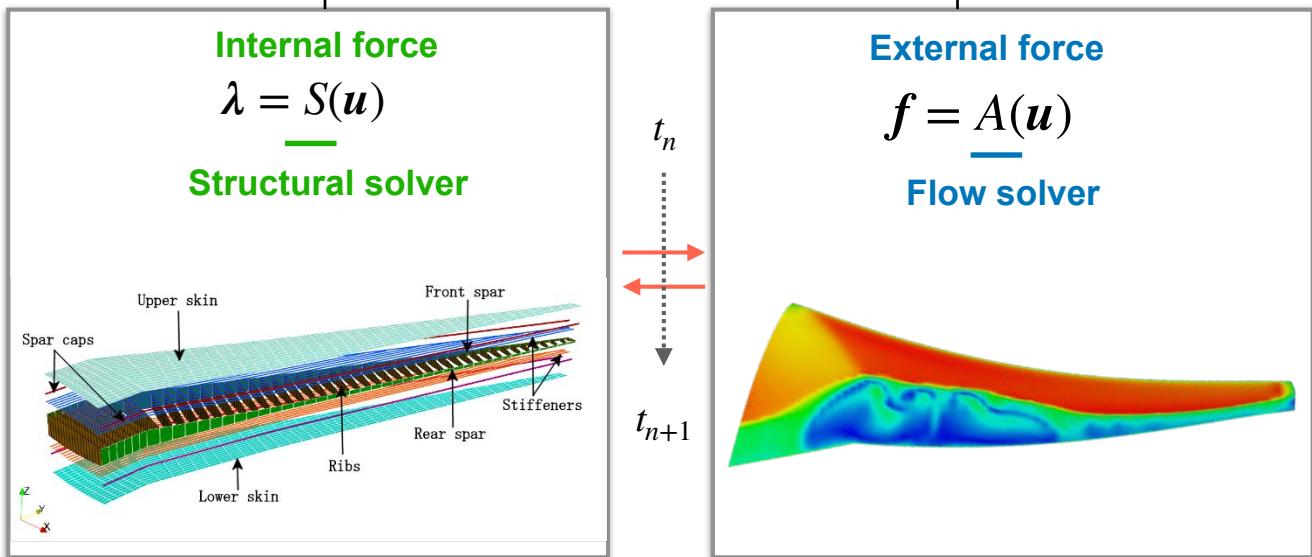
- Relative error (loss function)



► Iteration at fixed time

Farhat and Lesoinne *CMAME* (2000)
Barcelos *et al.*, *CMAME* (2006)

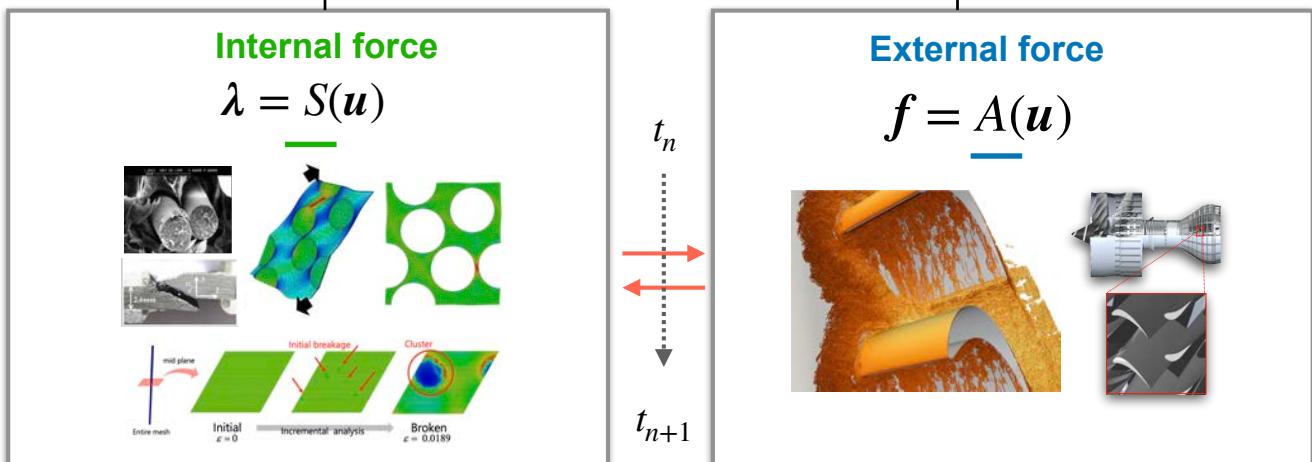
$$\lambda(\mathbf{u}) - \mathbf{f}(\mathbf{u}) = 0 \quad (\mathbf{u}: \text{displacement})$$



► Iteration at fixed time

Farhat and Lesoinne *CMAME* (2000)
Barcelos *et al.*, *CMAME* (2006)

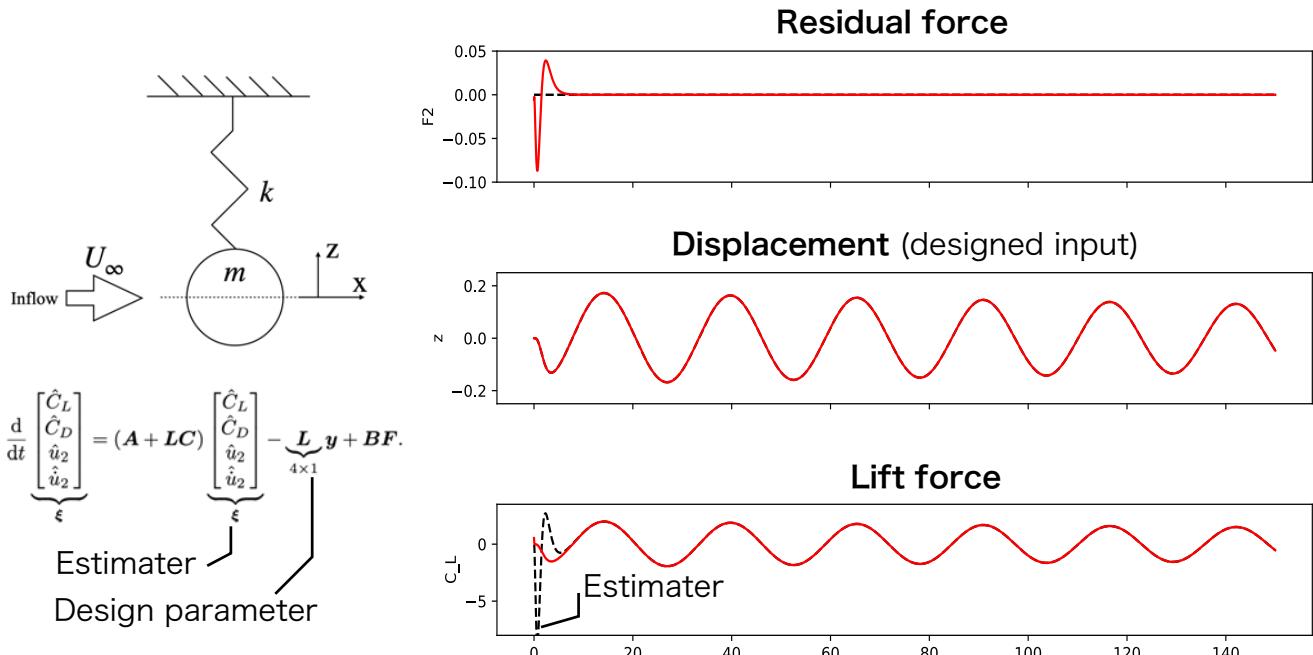
$$\lambda(\mathbf{u}) - \mathbf{f}(\mathbf{u}) = 0 \quad (\mathbf{u}: \text{displacement})$$



High-fidelity simulations in each physics?

- Large number of degrees of freedom in space
- Communications between “big” solvers at each time step

- **Observer-based control design (DMDc-based flow dynamics)**



Residual force can be replaced by simulation results

Conclusions

Data-driven FSI approach was proposed with

- Residual force minimization
- Deep Koopman dynamical model

- ▶ Parallelization of computing residuals for FSI solution was introduced
- ▶ Deep autoencoder was applied in the Koopman dynamical model

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