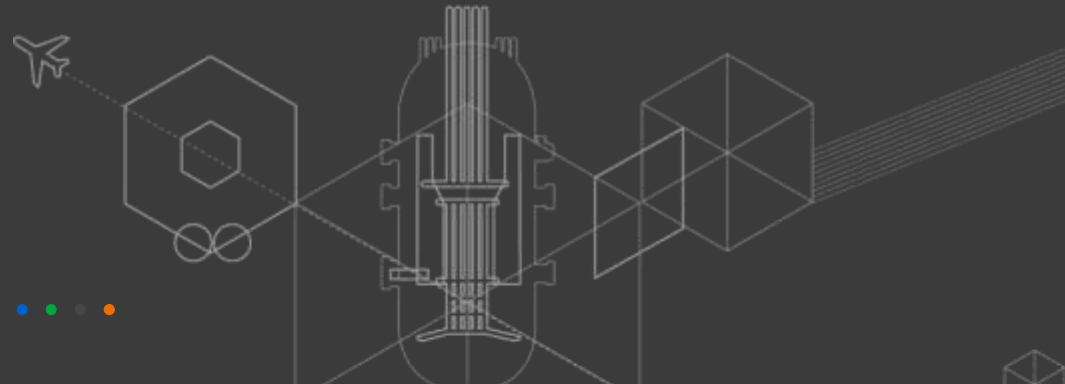


KAERI



딥러닝을 이용한 최적설계 및 시뮬레이션 소개

한국원자력연구원 유용균
(ygyu@kaeri.re.kr)

2018.03.19

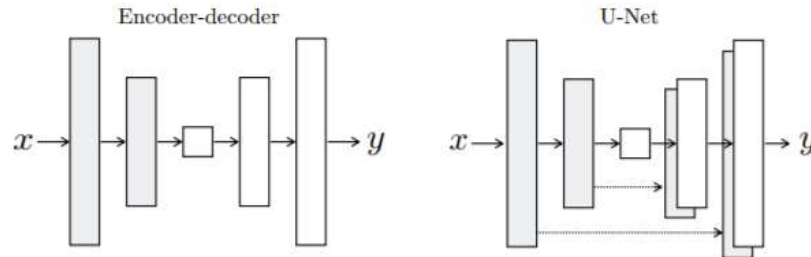
Image-to-Image Translation with Conditional Adversarial Networks



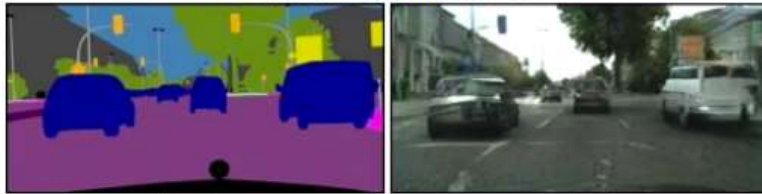
Image-to-Image Translation with Conditional Adversarial Networks



Image-to-Image Translation with Conditional Adversarial Networks



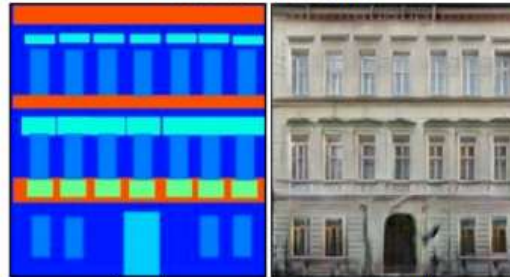
Labels to Street Scene



input

output

Labels to Facade



input

output

BW to Color



input

output

Aerial to Map



input

output

Day to Night



input

output

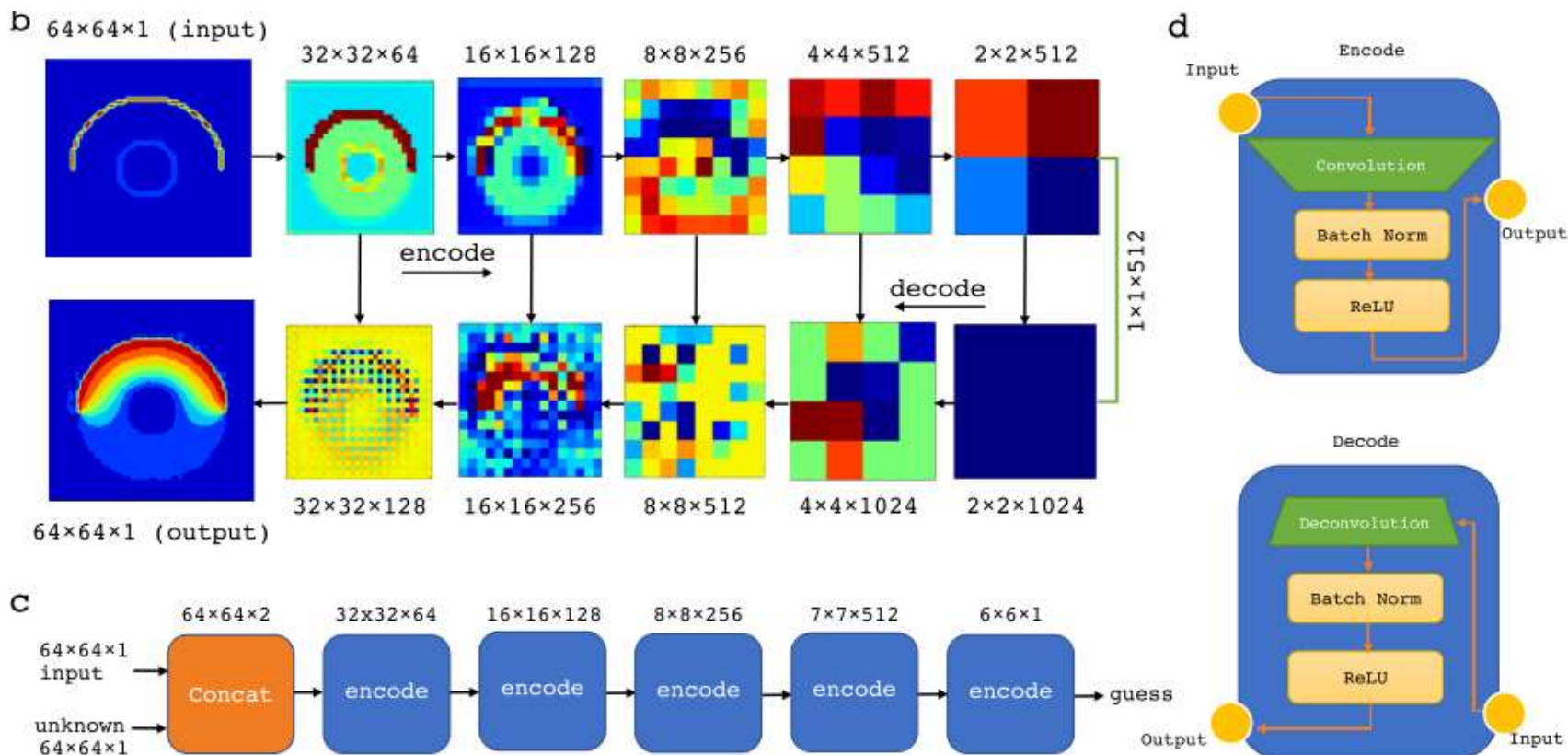
Edges to Photo



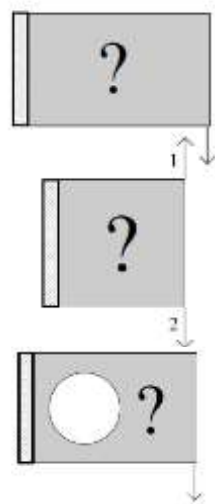
input

output

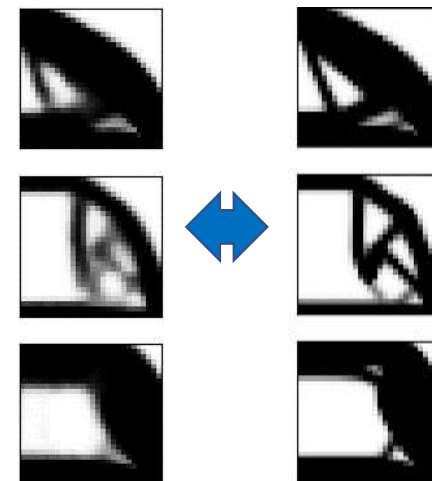
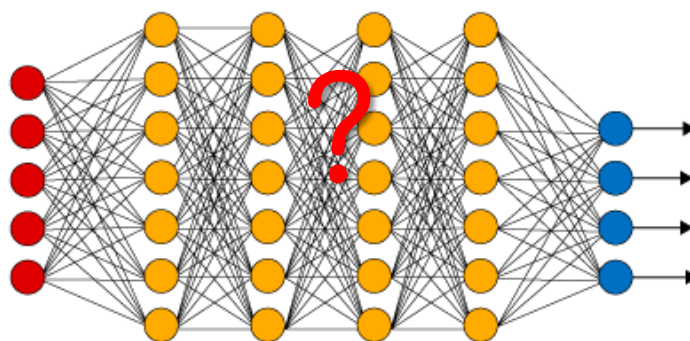
Deep Learning the Physics of Transport Phenomena



Deep Learning for Topology Optimization Design



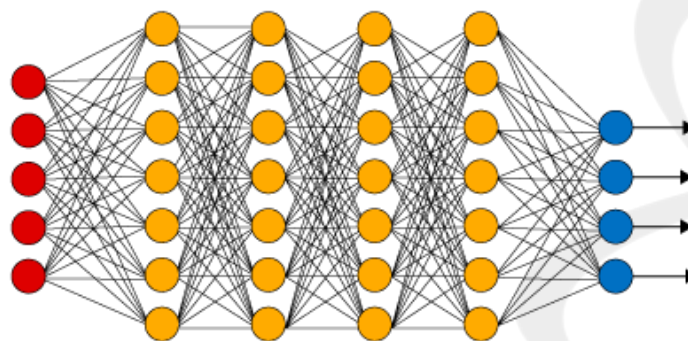
Deep Learning Neural Network



predicted

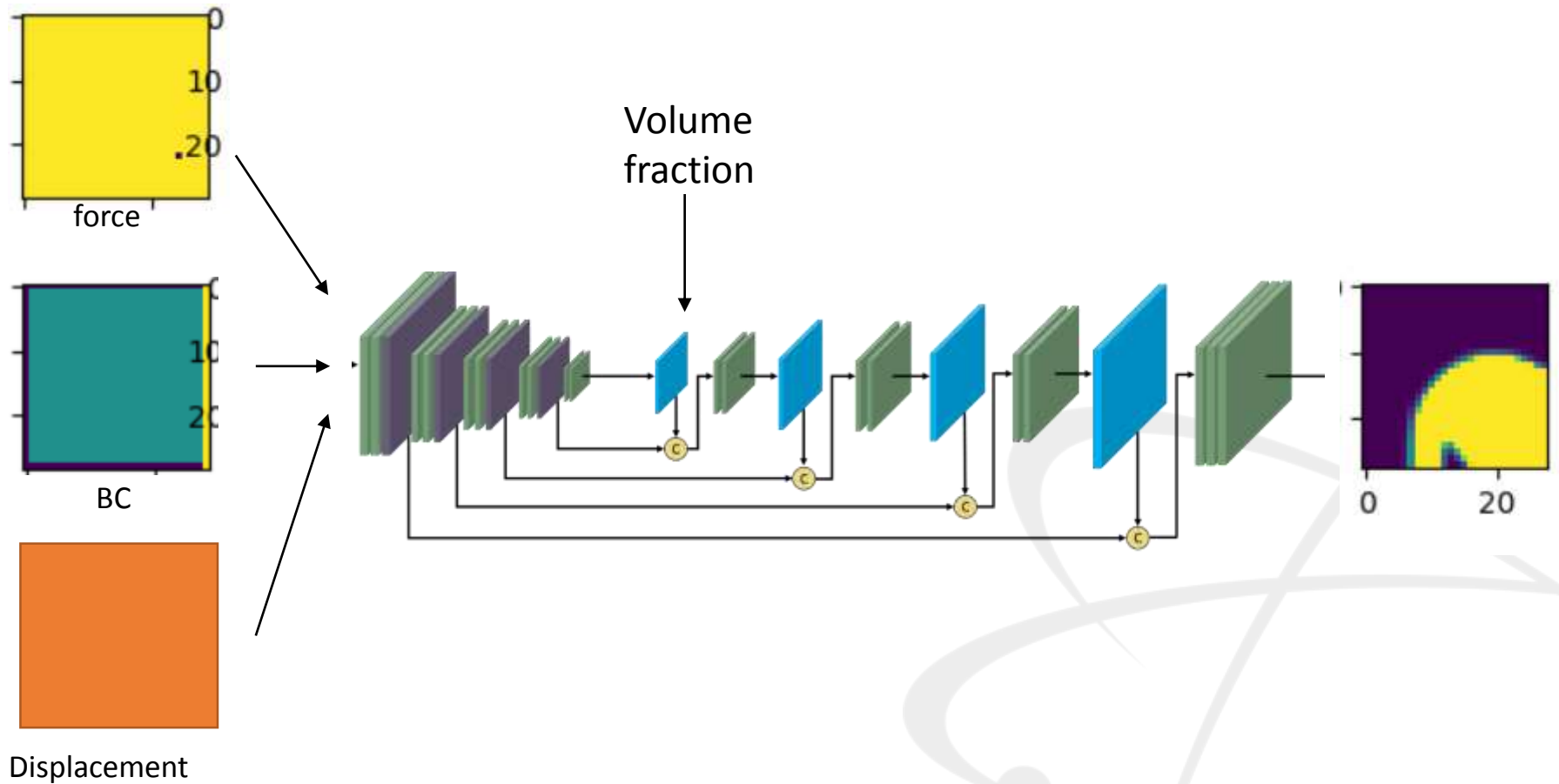


Deep Learning Neural Network

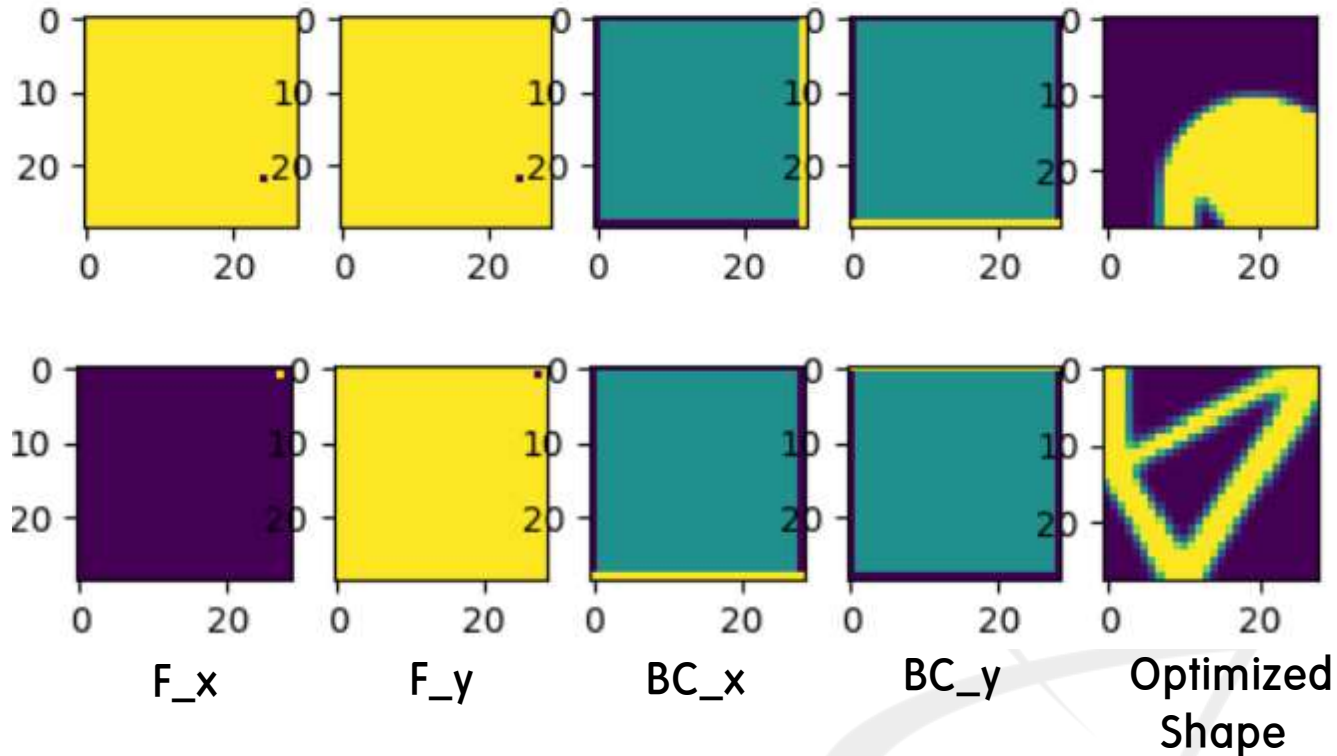


Inference

Deep Learning for Topology Optimization Design

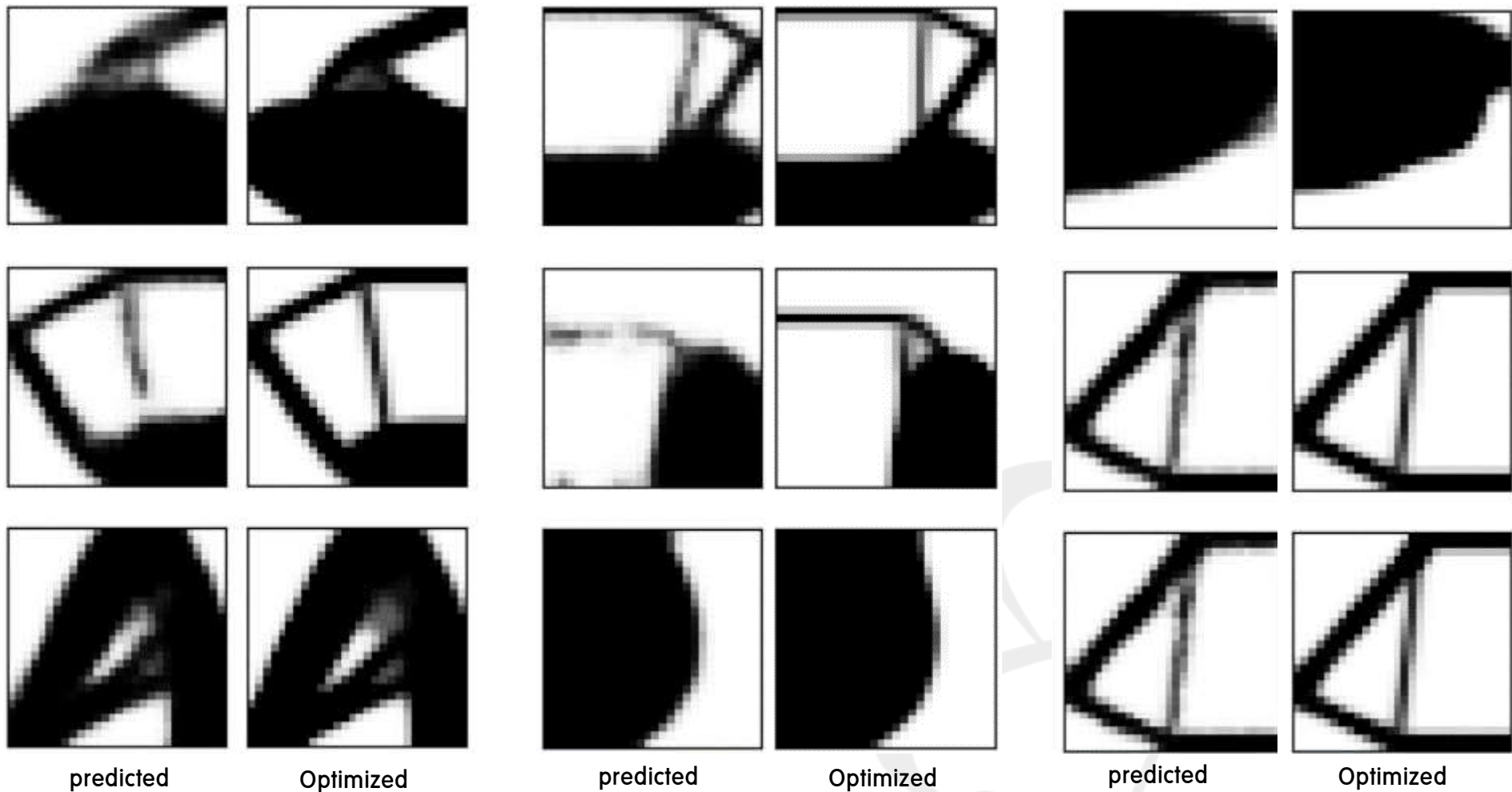


Deep Learning for Topology Optimization Design



입력 및 출력 데이터의 시각화

Deep Learning for Topology Optimization Design



predicted

Optimized

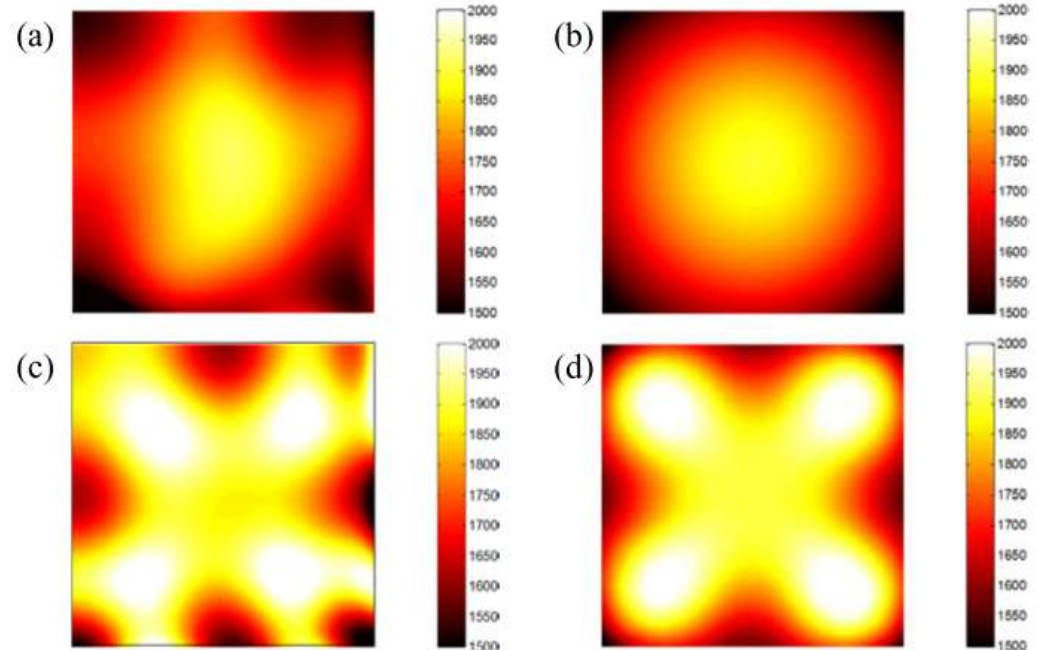
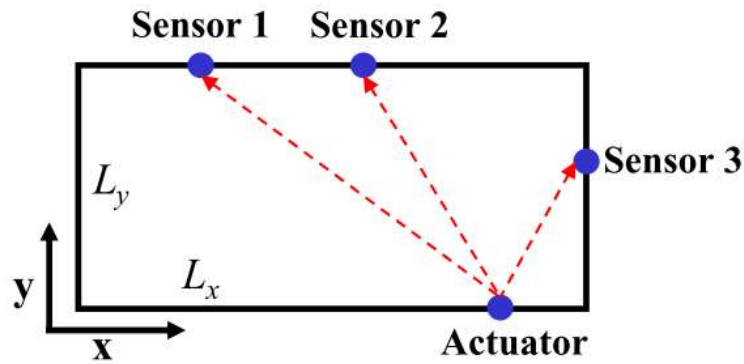
predicted

Optimized

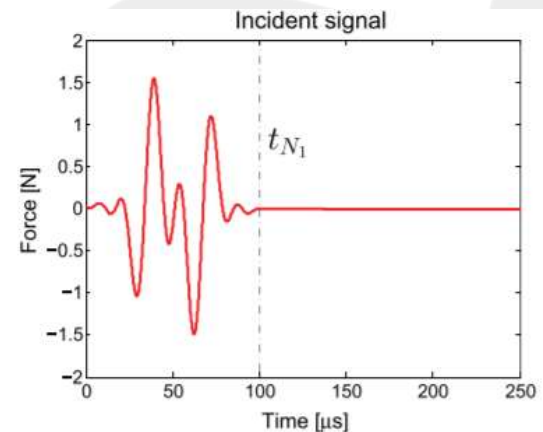
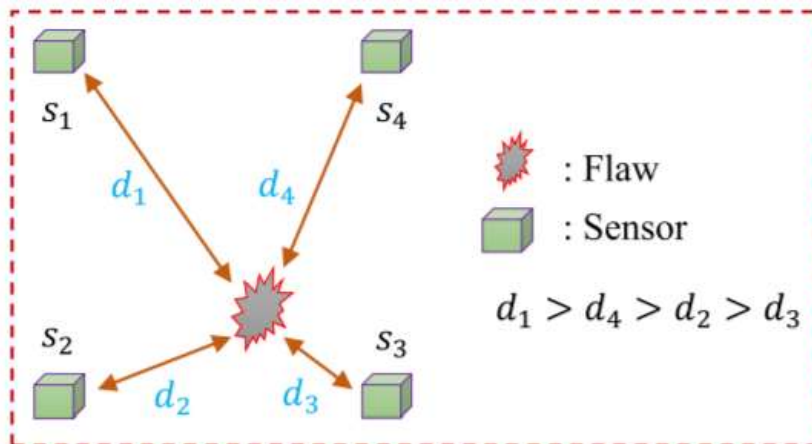
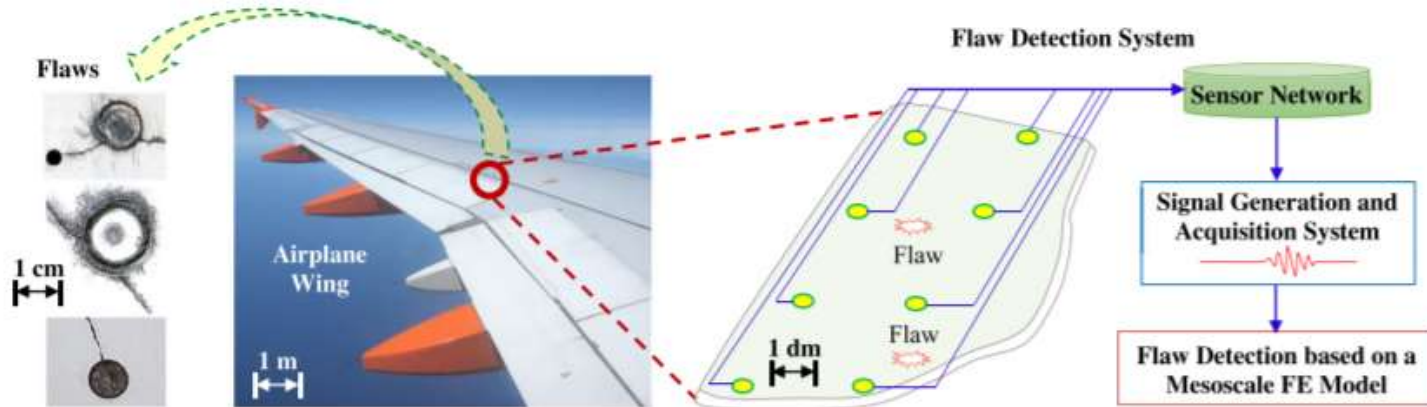
predicted

Optimized

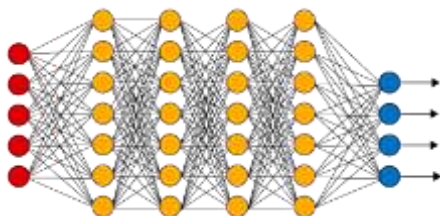
Inverse estimation of the temperature field within a gas-filled duct section by use of acoustic data



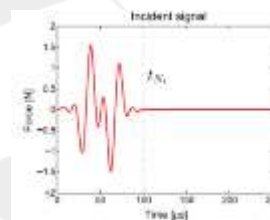
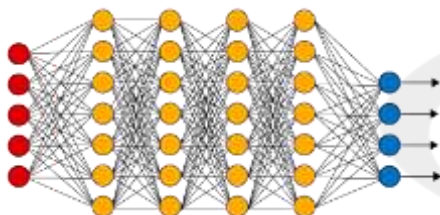
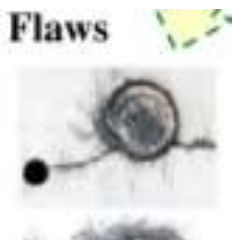
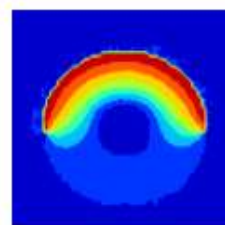
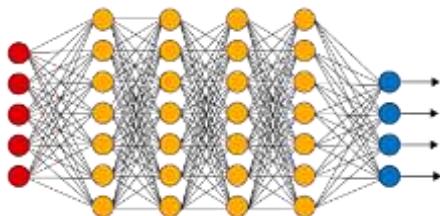
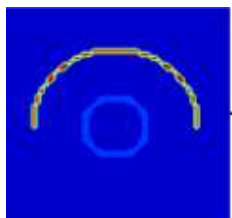
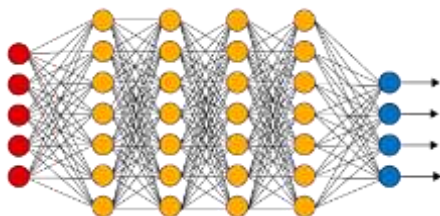
비파괴 검사



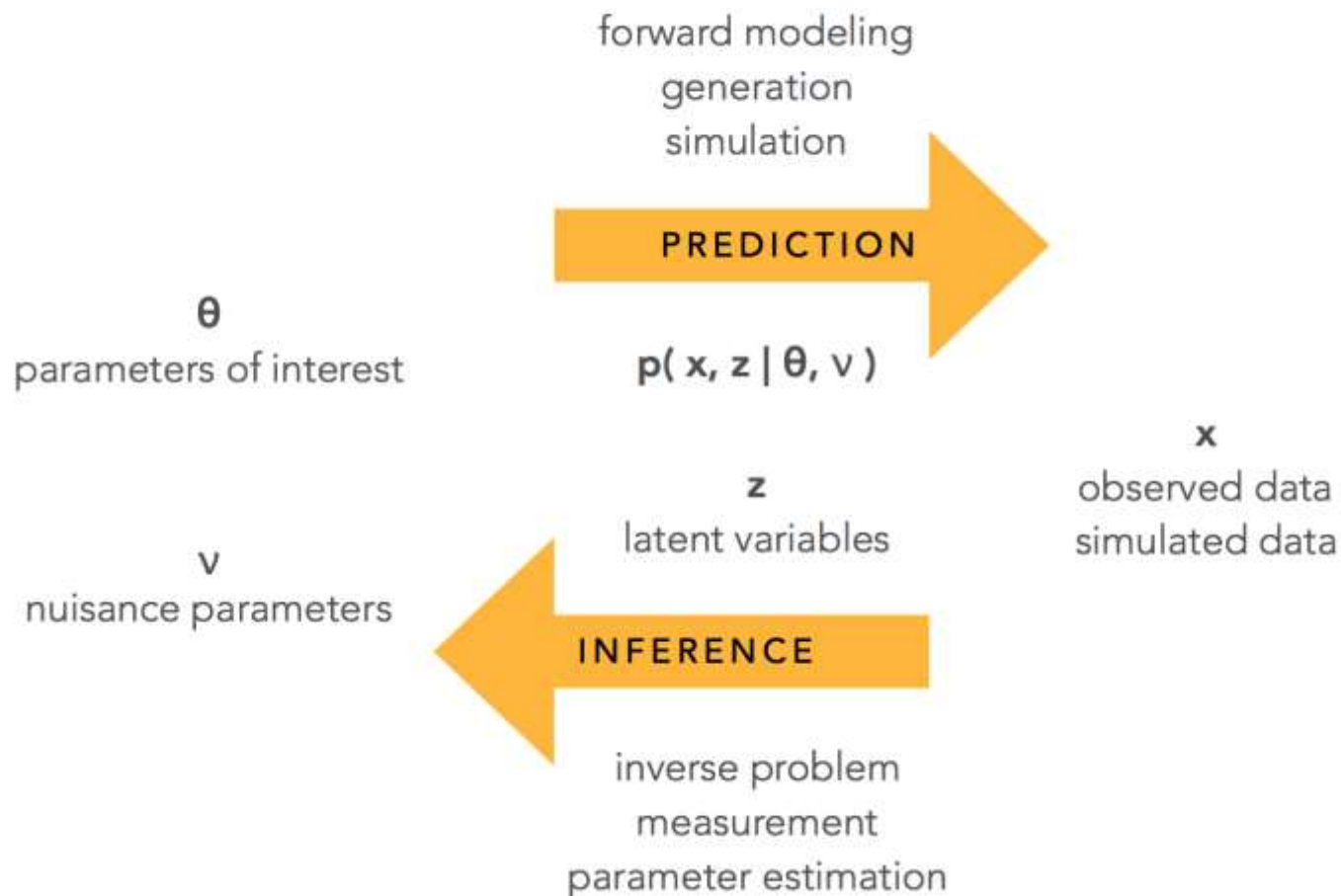
딥러닝이란?



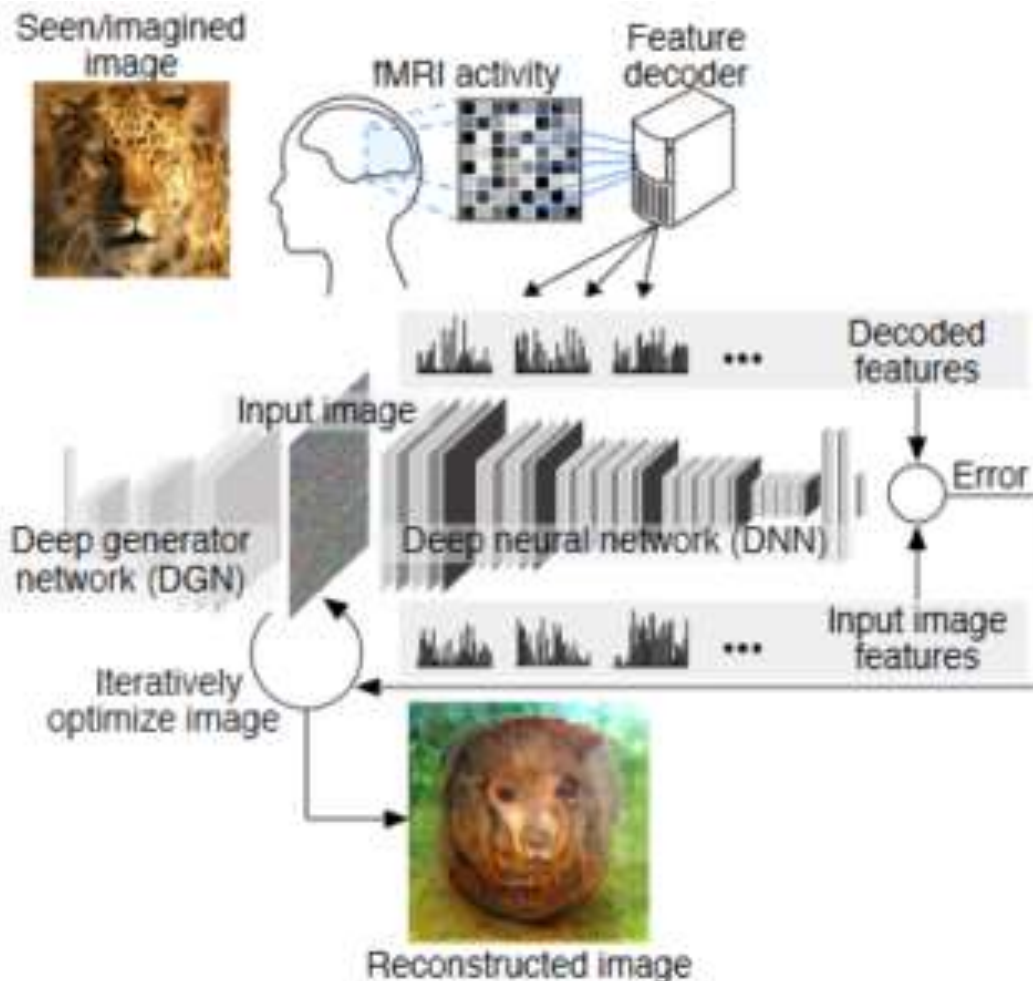
사람



딥러닝을 이용한 물리현상 분석



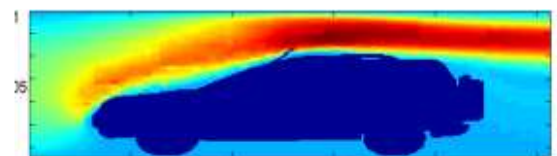
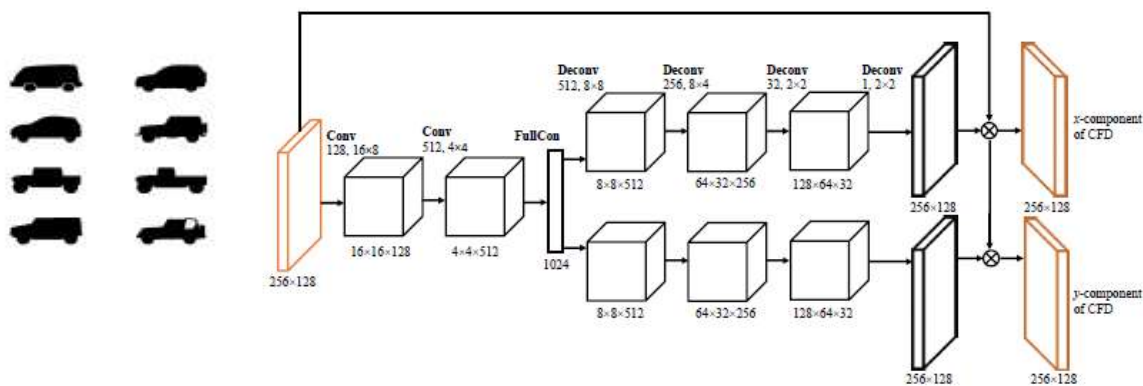
fMRI to Image



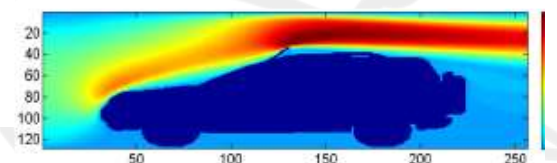
뇌의 동작
원리를 꼭
알아야 할까요?

Convolutional Neural Networks for Steady Flow Approximation

$$y=f'(x)$$



CNN Prediction

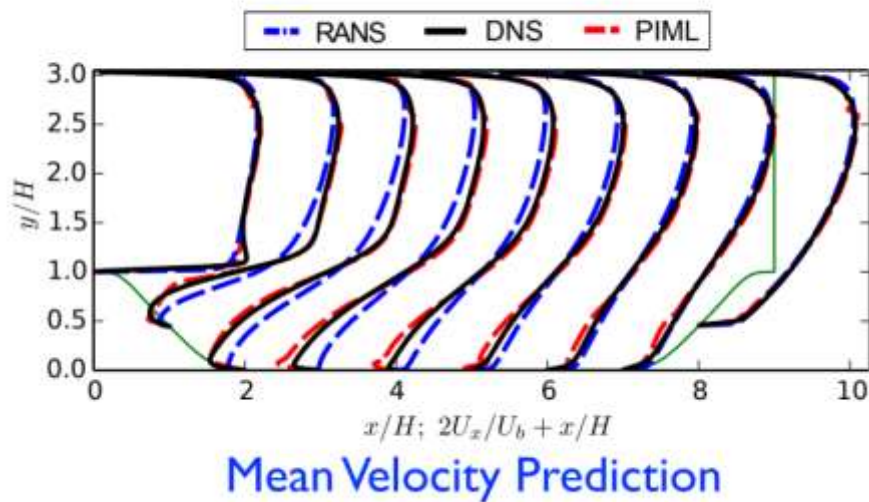


LBM

A comprehensive physics-informed machine learning framework for predictive turbulence modeling

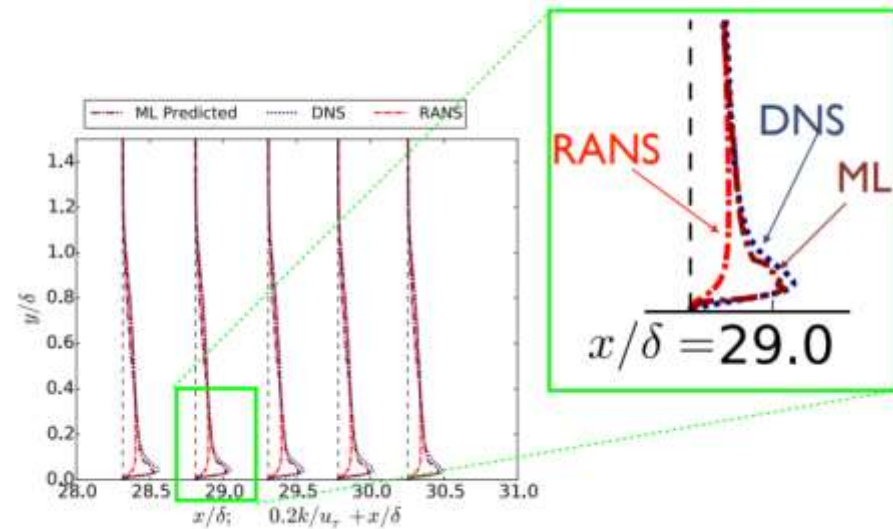
$$y=f(x)+f'(x)$$

Learning Both Reynolds Stress & Eddy Viscosity



24

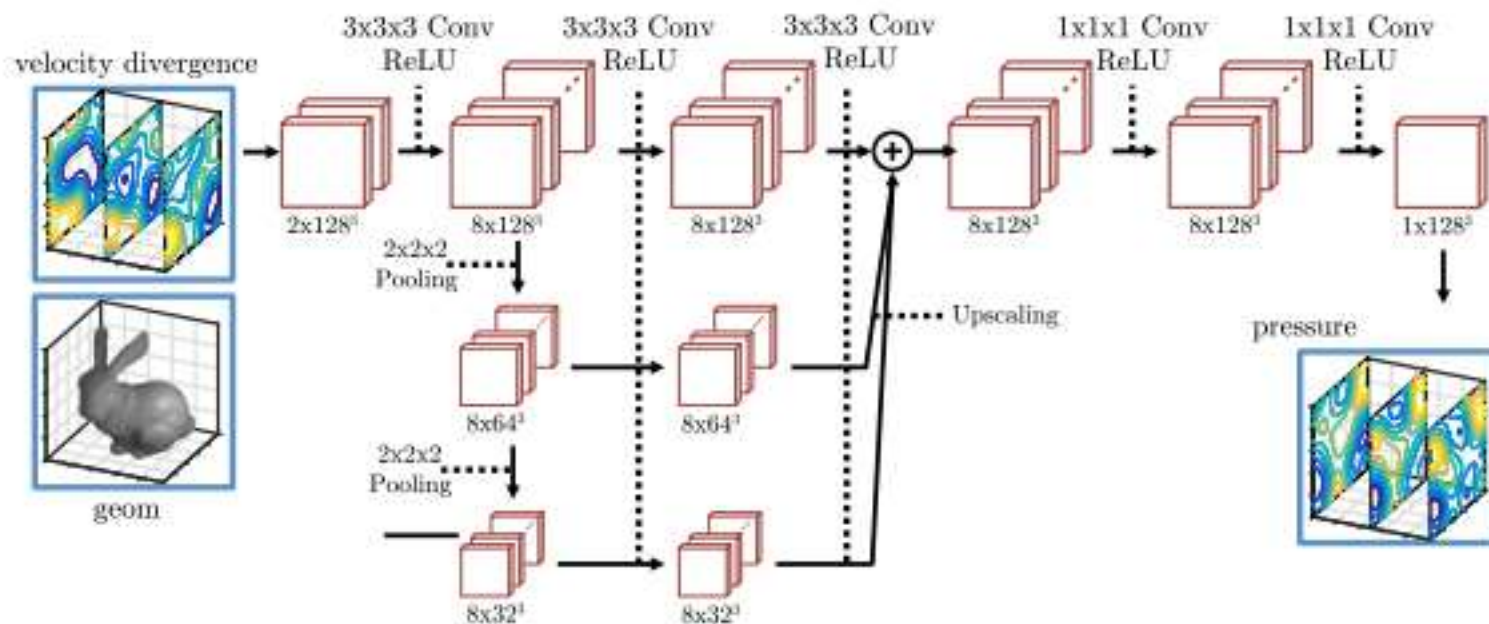
Turbulent Kinetic Energy



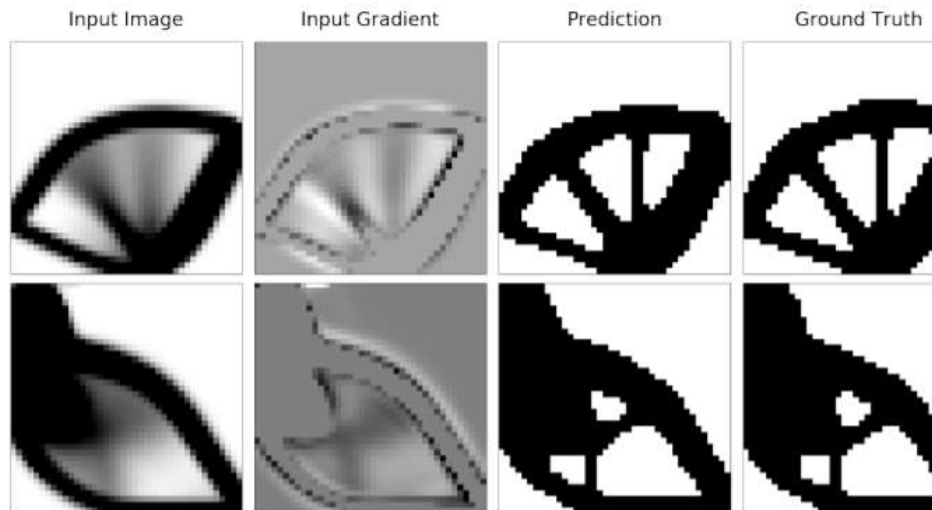
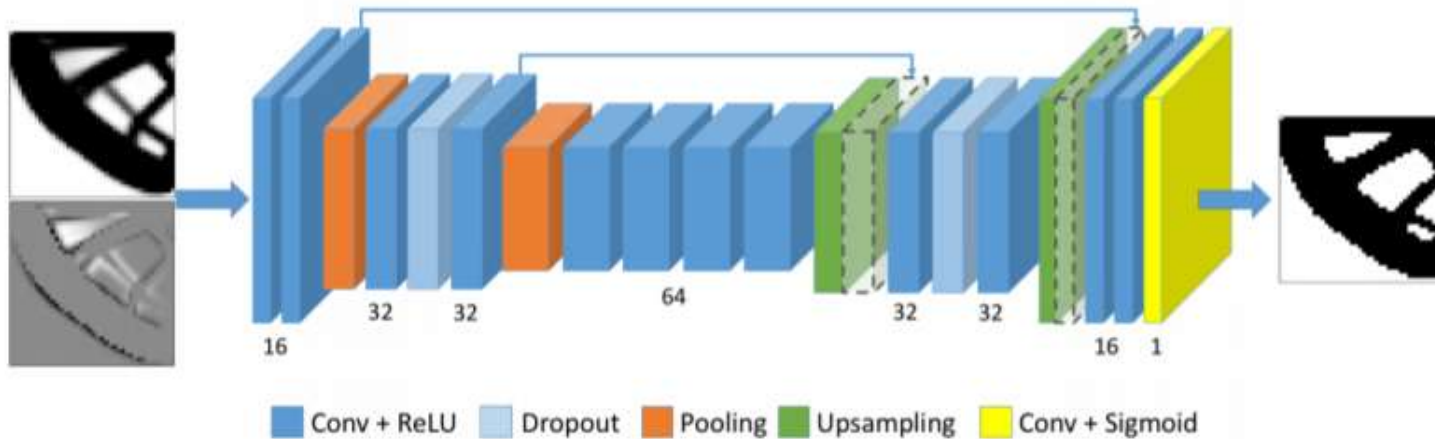
38

Accelerating Eulerian Fluid Simulation with Convolutional Networks

$$y = g(f'(x))$$



Neural networks for topology optimization

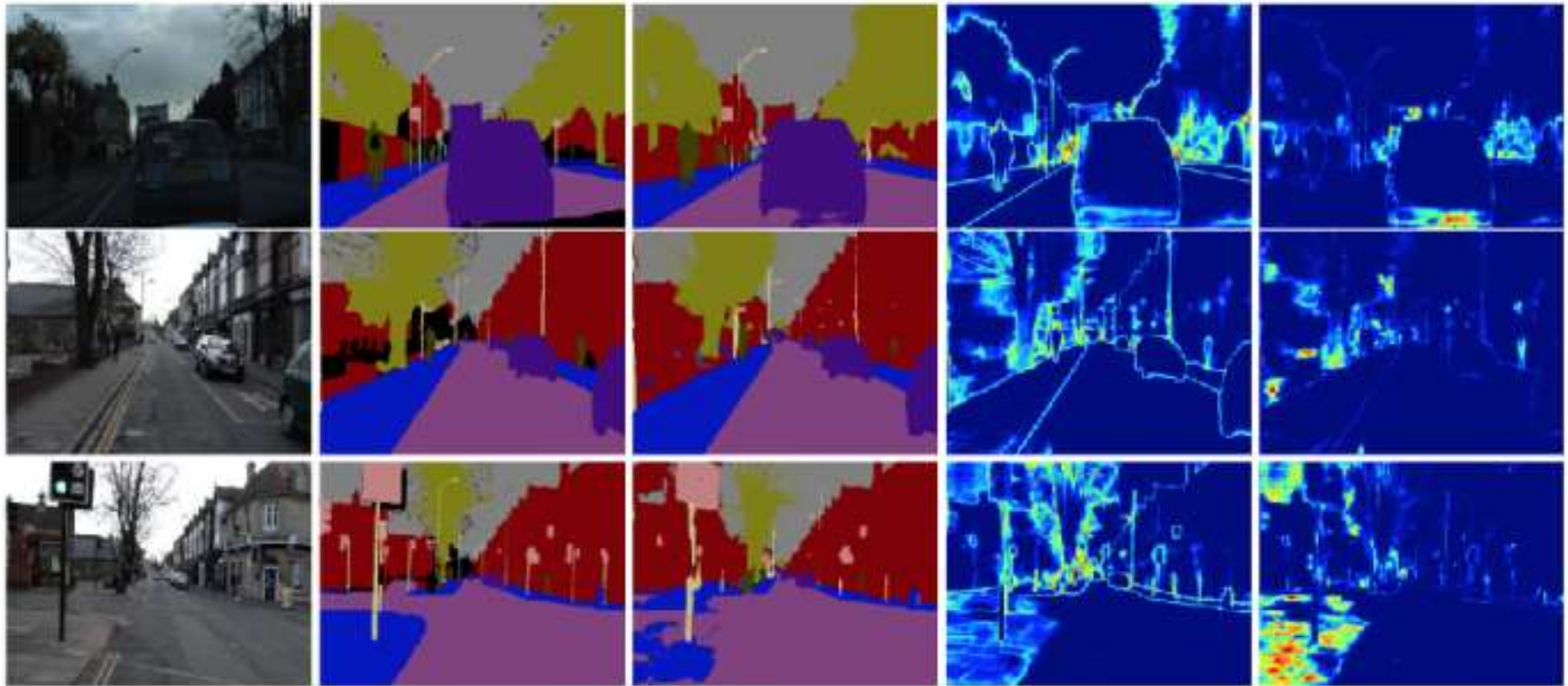


<https://arxiv.org/abs/1709.09578>

문제점....

얼마나 예측한 값을
믿을 수 있는가?

Aleatoric & Epistemic Uncertainty



(a) Input Image

(b) Ground Truth

(c) Semantic Segmentation

(d) Aleatoric Uncertainty

(e) Epistemic Uncertainty

<https://arxiv.org/abs/1703.04977>



Hybrid Approach?

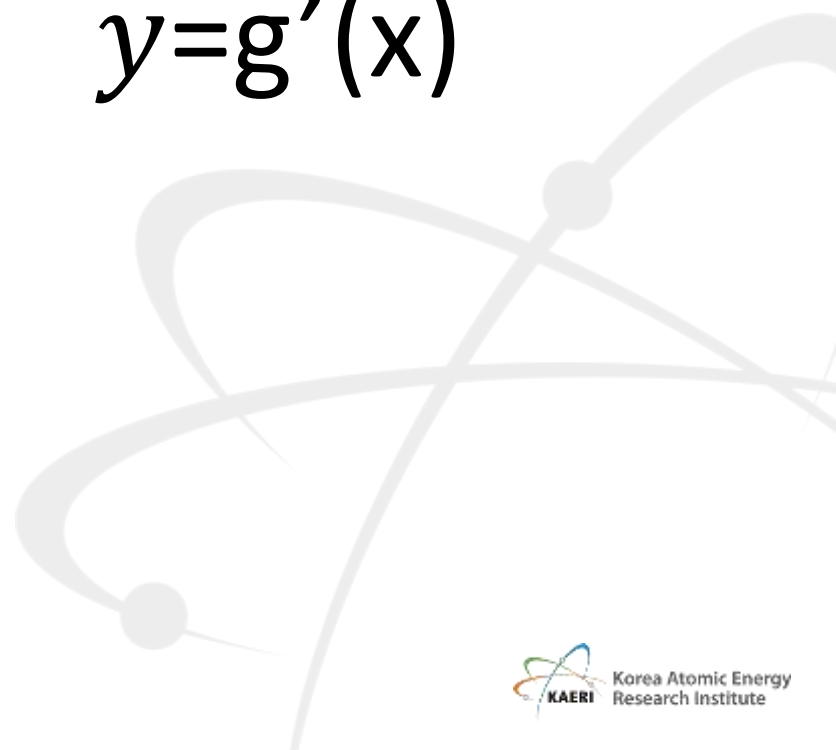
PDE

$$y=g(x)$$



Machine Learning

$$y=g'(x)$$

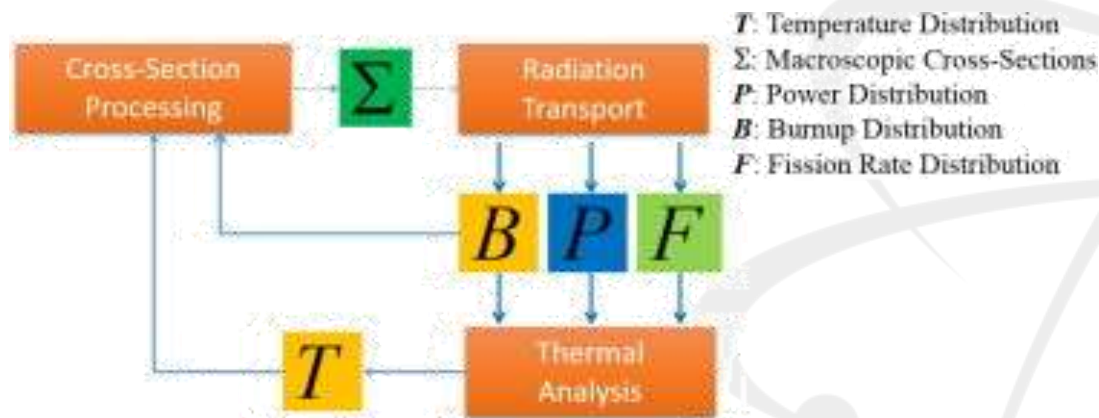


Dimensionality reducibility for multi-physics reduced order modeling

The **final goal** of this study is to construct a **surrogate model for the coupled Rattlesnake-BISON models**

The **computational cost** needed for the construction of surrogate models for a multi-physics model can be **significantly reduced** if one employs dimensionality reduction to identify the effective DOF.

Another important conclusion of this study is that while fine mesh simulation is highly needed to accurately describe the multi-physics nature of system behavior, it comes at a great cost.

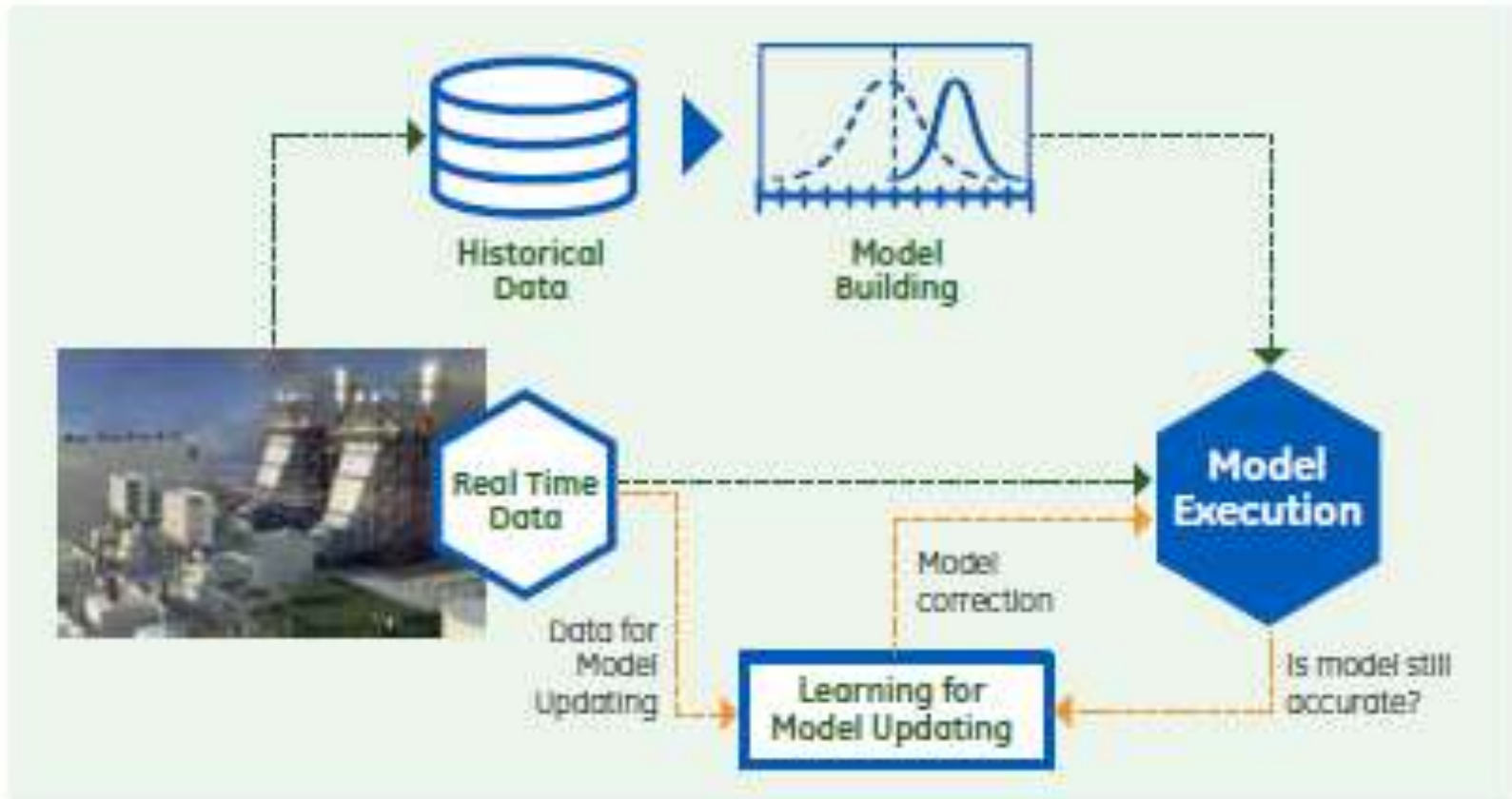


Combustion modeling using principal component analysis

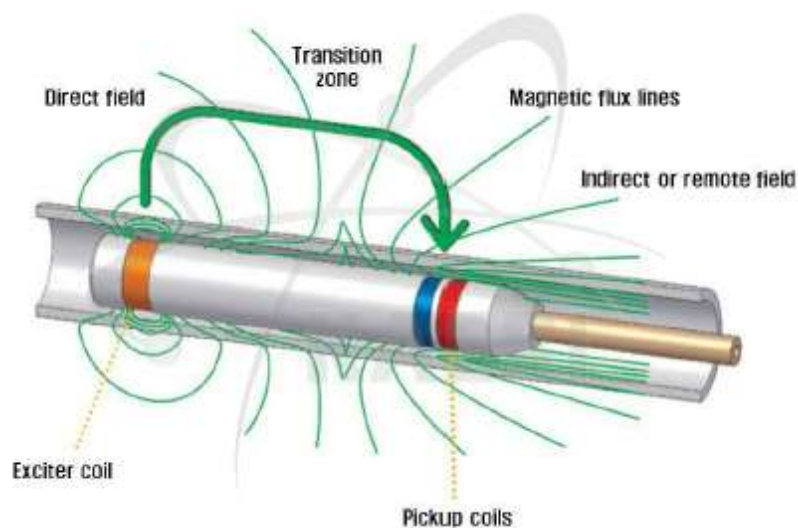
- Direct numerical simulation of combustion systems is impossible
 - Resolution requirement
 - Number of equations to be solved
 - Ex) 53 species and 325 reactions
 - 57 strongly coupled PDE
- PCA offers the potential to automate the selection of an optimal basis for representing the manifolds

$$\mathbf{X} \approx \boldsymbol{\eta} \mathbf{A}^T \quad \rho \frac{D(\boldsymbol{\Phi})}{Dt} = -\nabla \cdot (\mathbf{j}_{\boldsymbol{\Phi}}) + (s_{\boldsymbol{\Phi}}) \quad \rho \frac{D}{Dt}(\boldsymbol{\eta}) = -\nabla \cdot (\mathbf{j}_{\boldsymbol{\eta}}) + (s_{\boldsymbol{\eta}});$$

Digital Twin

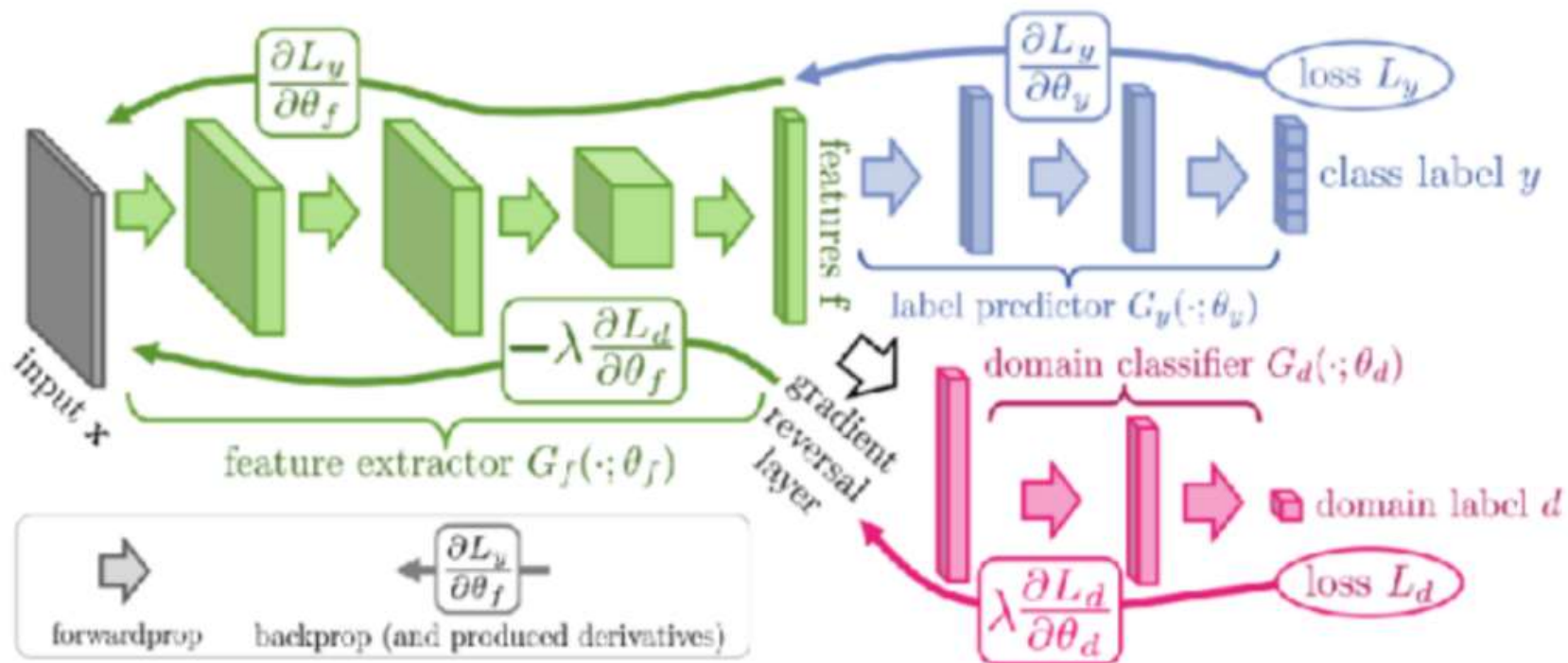


Simulation vs. Real Data



- (필요한) 데이터는 부족
- 시뮬레이터가 필요.
- 시뮬레이터와 실제 데이터 사이의 차이는?

Domain Adaptation



<https://arxiv.org/abs/1505.07818>

연구 그룹

- Duraisamy, Univ. of Michigan
4.9~4.10 Workshop in KISTI
- Virginia Tech
- Nam Dinh

**Principles and Applications
of Data-Driven Fluid Dynamics Modeling**

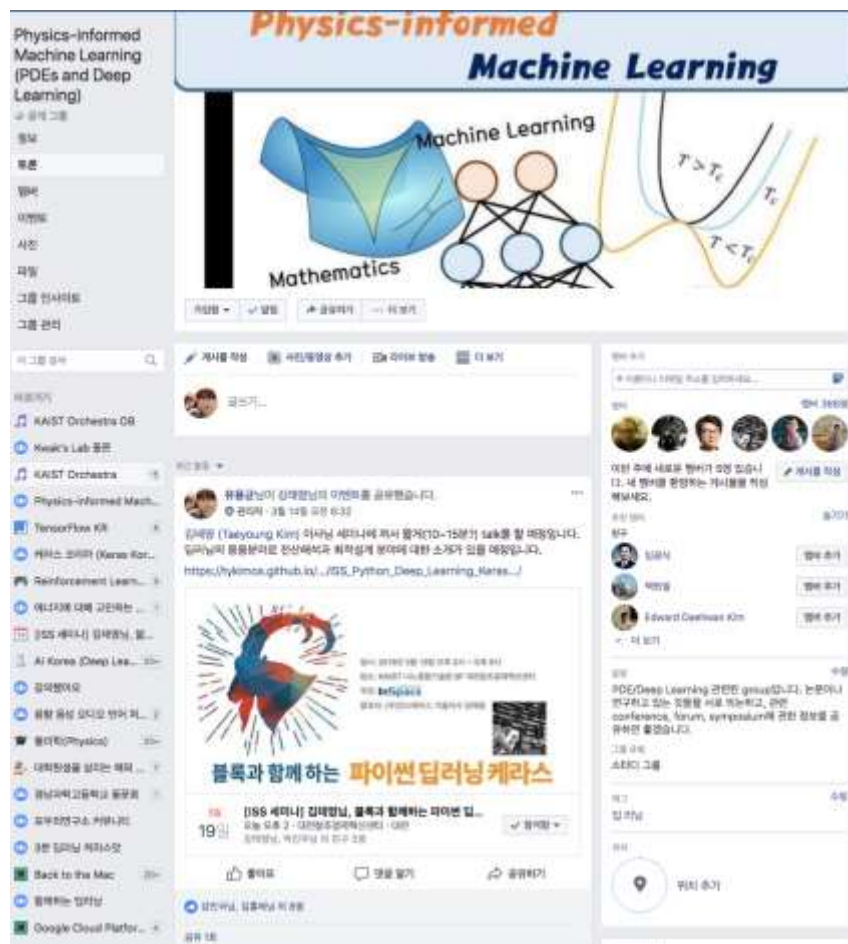
일시: 4/9(월)~10(화) Tutor: Prof. Karthik Duraisamy
 장소: KISTI(대전) 빌라동 슈퍼컴교육실 Dep. of Aerospace Eng., Univ. of Michigan

Programme

	· Registration (10:00~10:30) 1. Background in data-driven modeling: (10:30~) · Linear and non-linear regression · Introduction to statistical inference · Introduction to machine learning
Day 1 (4/9, Mon)	Lunch time
	2. Projection-based reduced order modeling for fluid dynamics
	3. Data-driven discovery of closure models for dynamical systems
	Dinner time
Day 2 (4/10, Tue)	4. Data-driven turbulence modeling (1/2) (10:30~) · Nature and form of uncertainties in turbulence modeling · Techniques for solution of inverse problems
	Lunch time
	5. Data-driven turbulence modeling (2/2) · Field Inversion and Machine learning · Applications in data-driven turbulence modeling
	6. Discussion (16:00~)
참가신청 및 문의	- 조금원 박사(ckw@kisti.re.kr) - 신정훈 박사(shandy77@kisti.re.kr, 042-869-0602)

주최: 한국과학기술정보연구원

Facebook group: Physics-informed Machine Learning



덕업일치를 꿈꾸며....



출처:네이버 웹툰 '호랭총각'

THANK YOU

