



Unintentional Bias and AI-Enabled Medical Devices

(Real World Domain Shift and Domain Adaptation Examples)

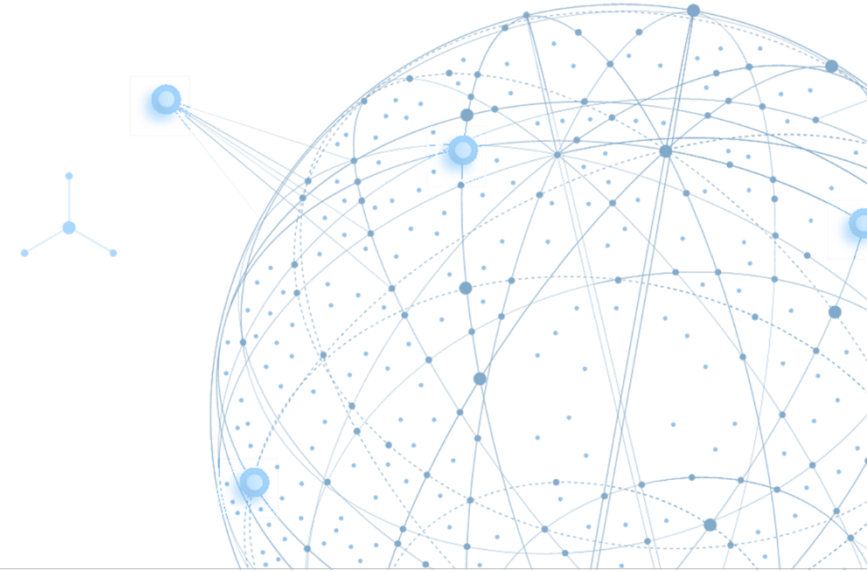
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CEO, JLK Inc.



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01

Real World Domain Shift Problem

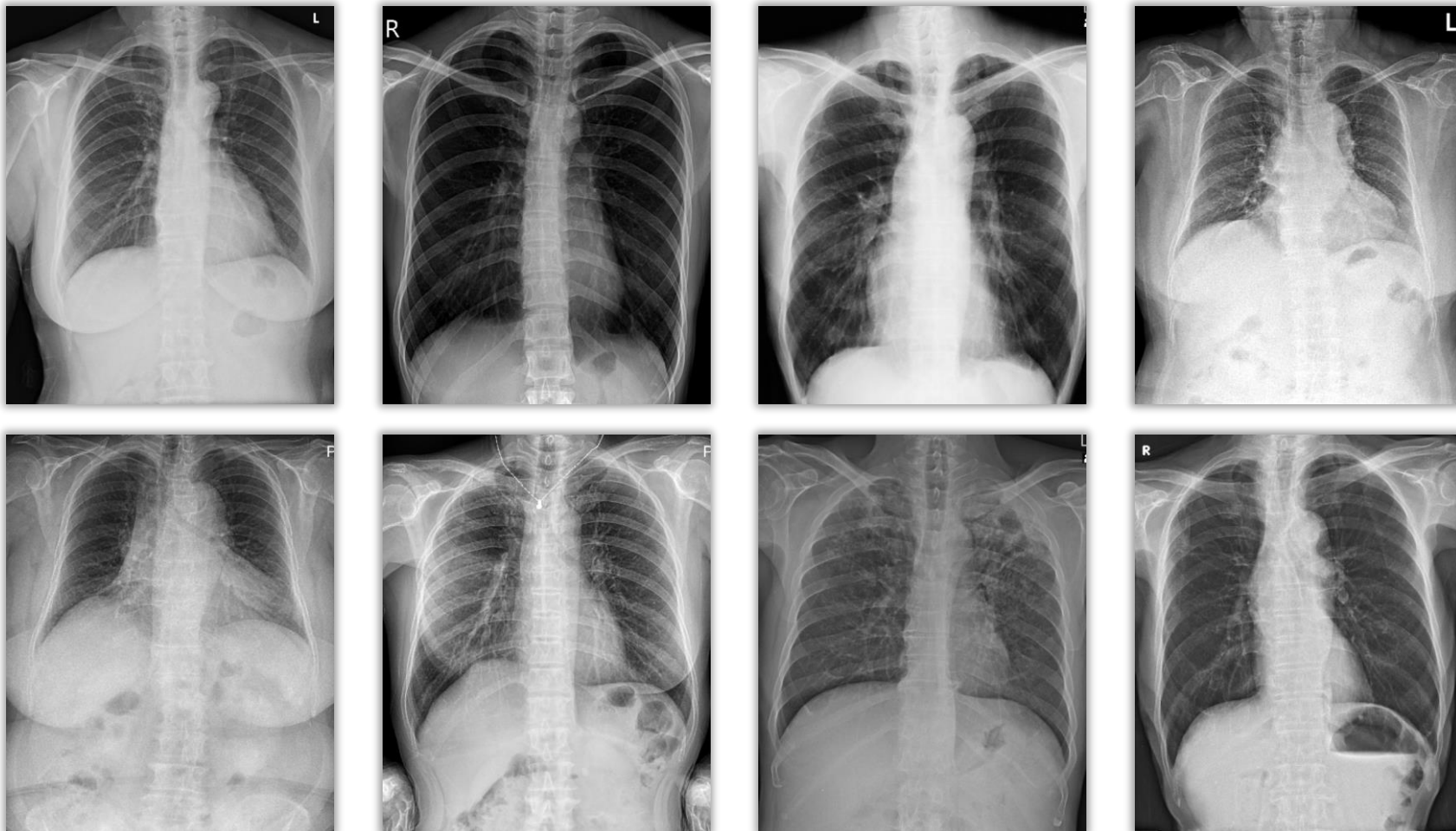


Real World Problem

Undefined image texture differences and unexpected image conditions
because of various equipment vendors and clinical situations

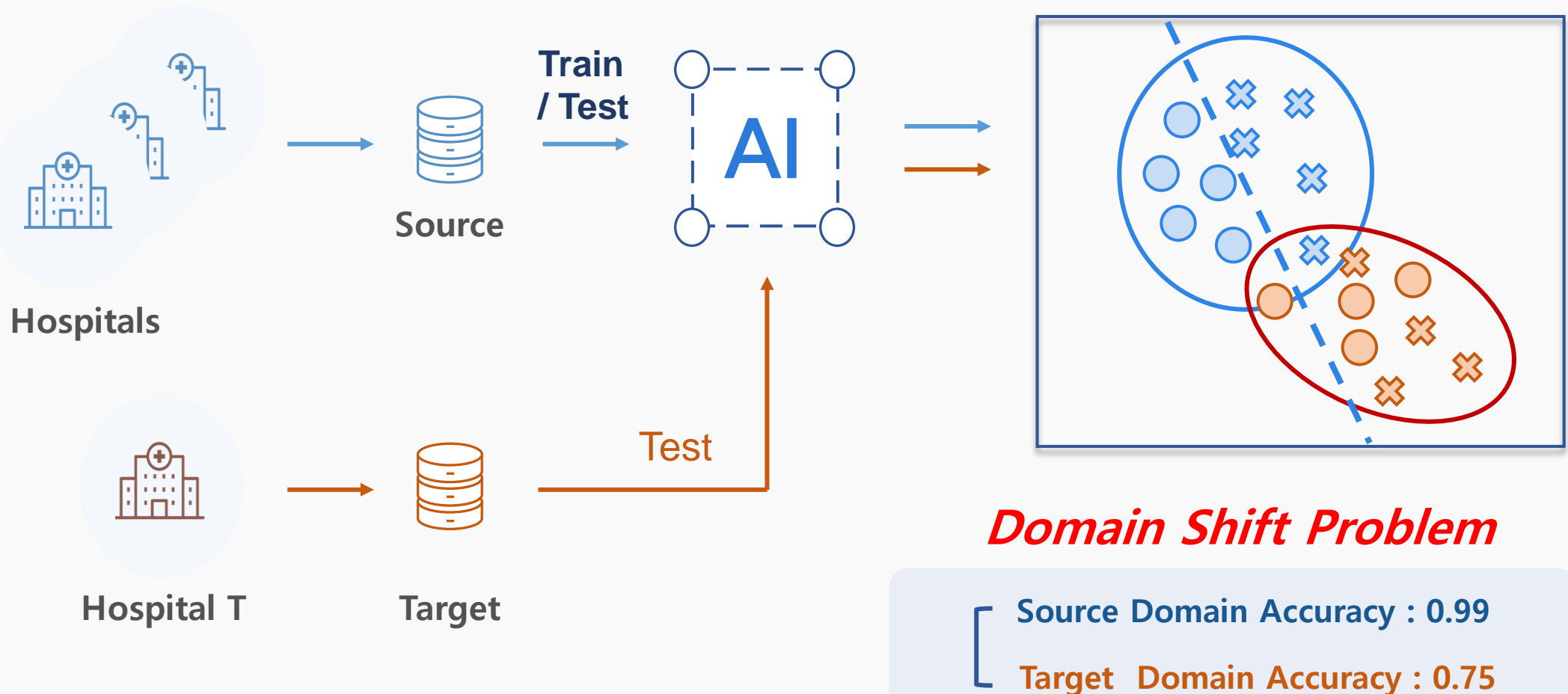


Unexpected performance degradation of trained AI



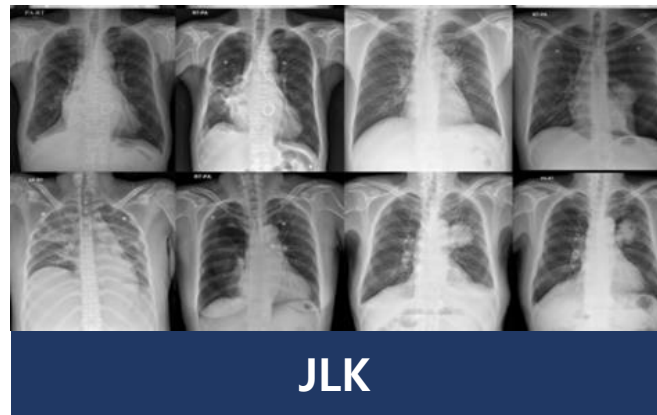
Domain Shift Problem

- ✓ Caused by **dataset bias**, such as gender, age, race, equipment and so on.
- ✓ One of the main reasons that **degrade** robustness of the trained network.

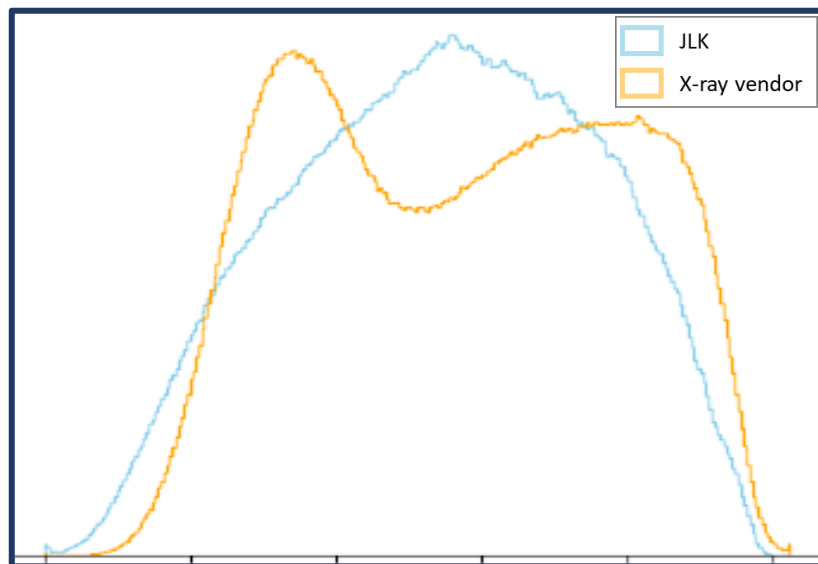


Real World Domain Shift Problem

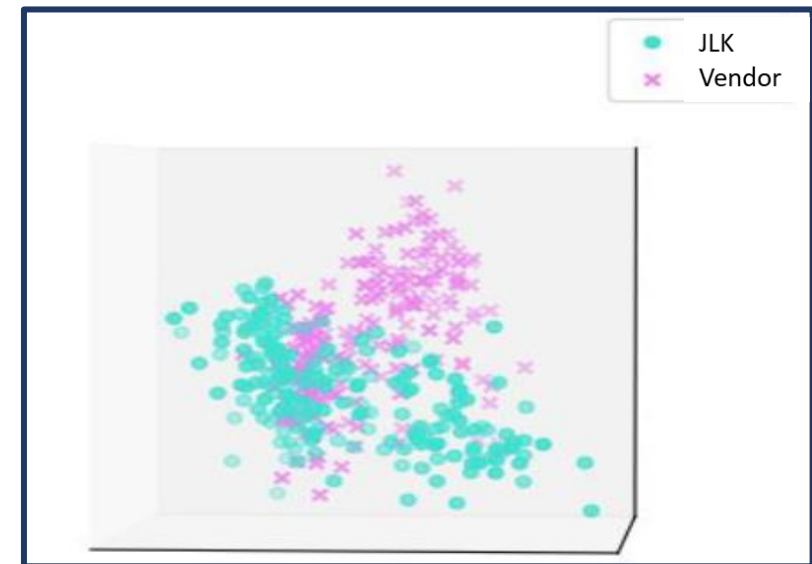
Texture difference exists between source and target domain dataset resulted from different X-Ray equipment vendors



Pixel-intensity analysis



Histogram



PCA

Real World Domain Shift Problem

Performance degradation in **low-quality(normal)** and **early pneumonia** images



Normal



Normal, low quality



Early Pneumonia



Pneumonia

Real World Domain Shift Problem

Practical Example -

Unexpected Cases for Pneumonia Analysis

Grad : Mid, Probability : 40.66%



Grad : Mid, Probability : 59.46%





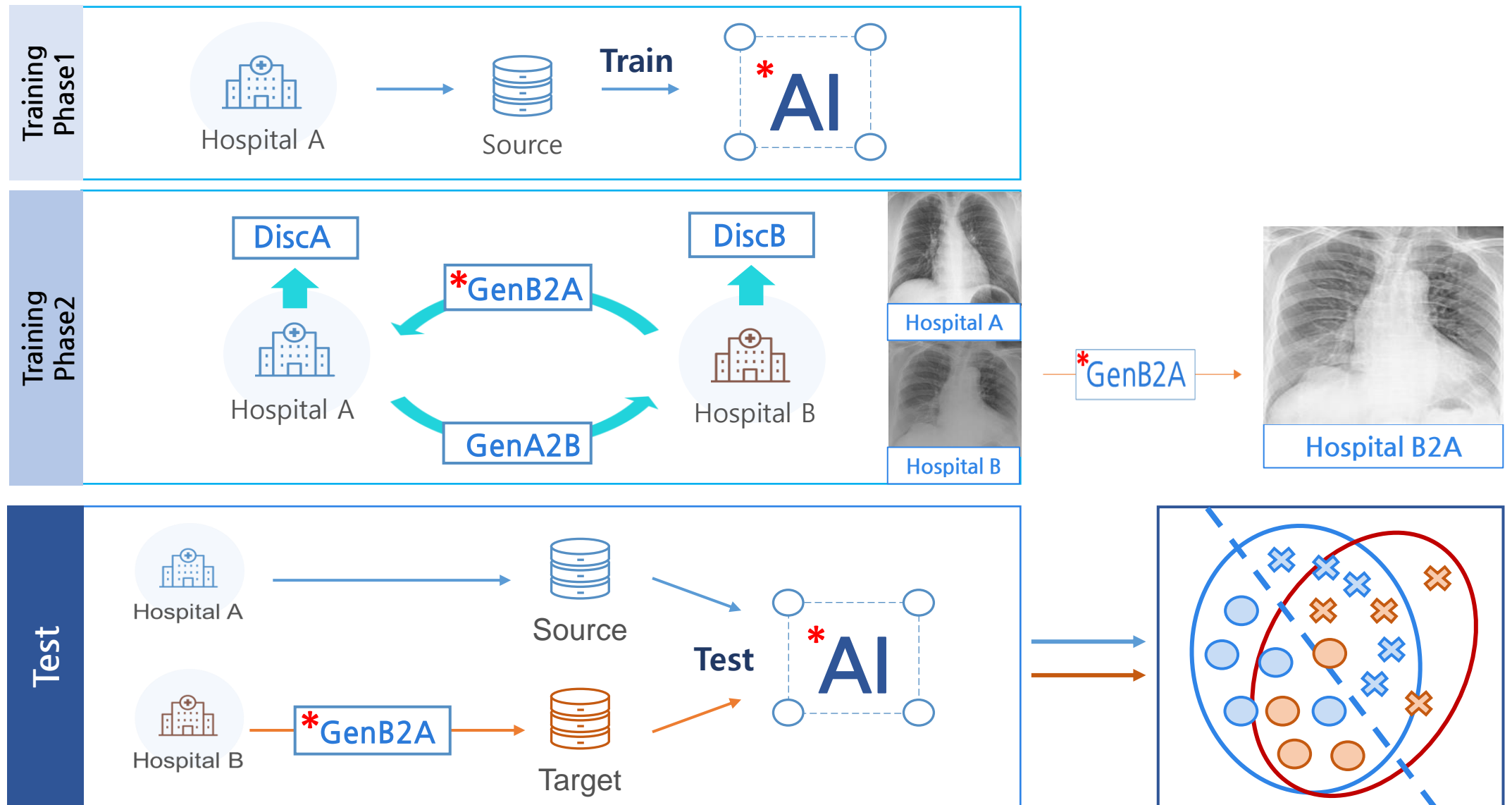
02

Domain Adaptation Examples



Generalize vendor differences using a generative neural network

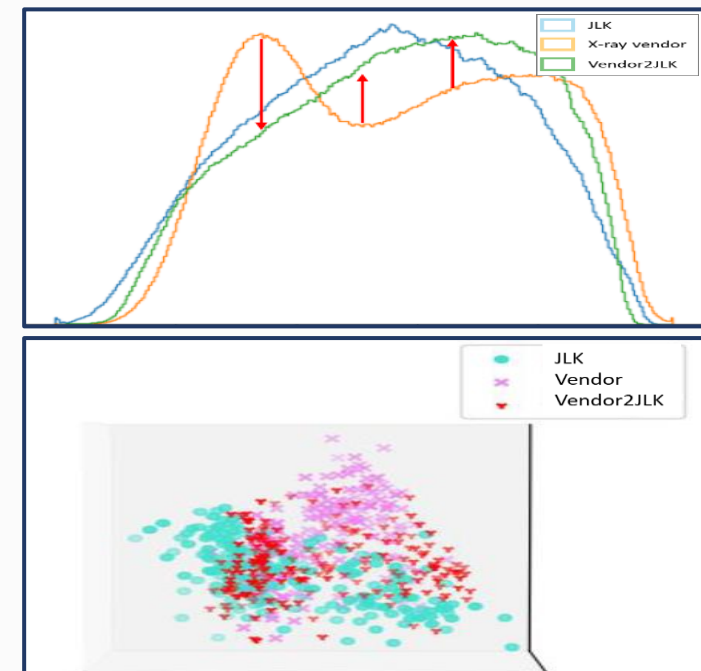
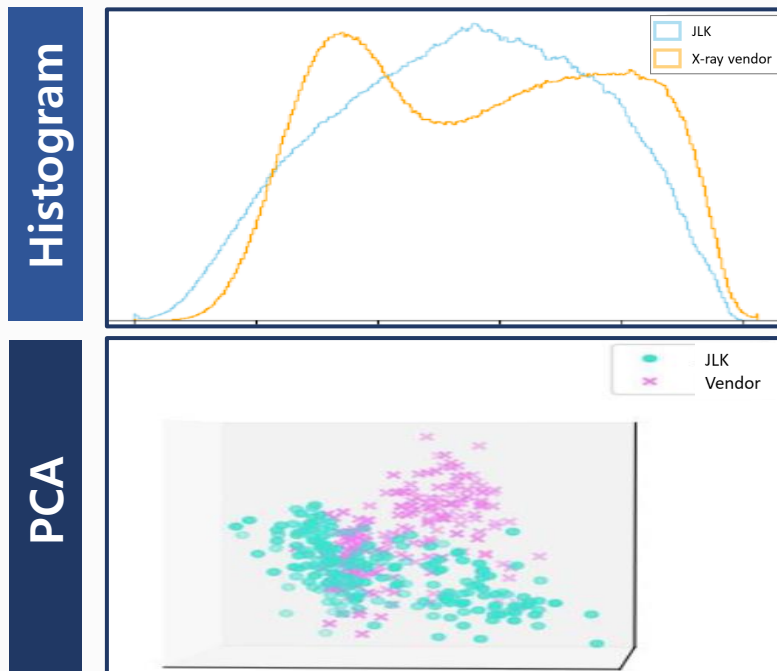
- ✓ It is important to match **image textures** between source and target dataset
- ✓ Generator trained based on CycleGAN is used in **preprocessing step** to handle the domain shift problem



Generalize vendor differences using a generative neural network

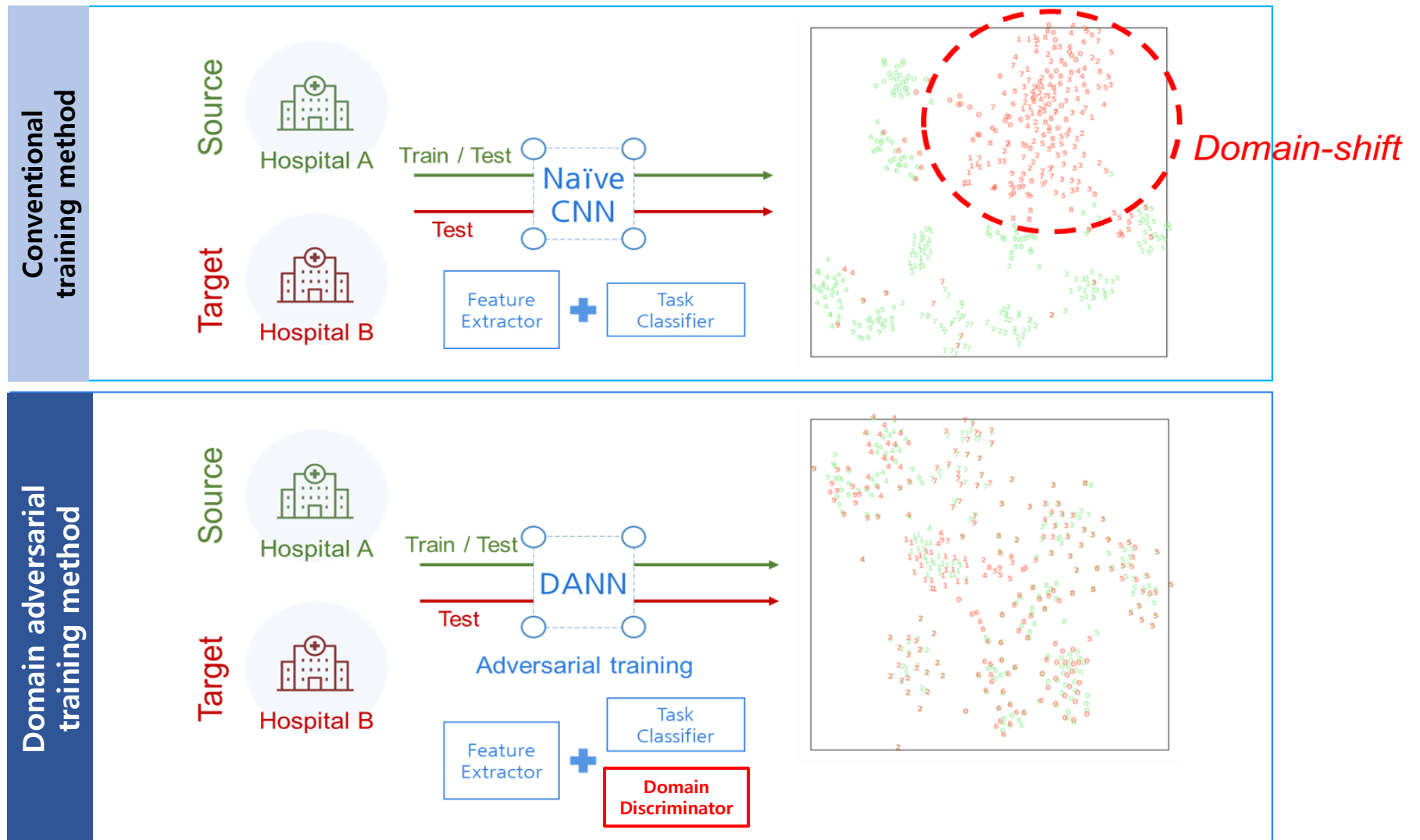
Practical Example -

- ✓ The Histogram of target dataset becomes similar to source dataset
- ✓ As a result of PCA, the distribution of the source and target becomes similar



Domain Adversarial Training

- ✓ Training a feature extractor to generate domain-independent features
- ✓ New network called '**domain discriminator**' is added to use both class information and domain information





03 Conclusion



1. If we have distribution in real-world, we get unintended AI analysis results depending on race, type of equipment, clinic protocol, age and body type.
2. Ultimately, it is important to construct representative learning data which could predict these all exceptional cases and remove bias of learning data.
3. However, in an exceptional case such as this, relevant database does not exist and only a few of case studies are found in journals. Thus, it is difficult to make predictions on the situation.
4. Regulatory Affairs Guidelines stipulated by Korea's Ministry of Food and Drug Safety in 2019 include standards for performance enhancement through additional training dataset.
5. If we were to take aforementioned such traits as Active Implantable Medical Device's inevitable characteristic, and say that unexpected degradation in performance was found on the approved AIMD articles, the manufacturer will proceed with an update and report details to build a database for unexpected circumstances, further reinforcement on the Regulatory Affairs for Medical Devices will take place.



04 Appendix



Company Overview (Medical AI Solution Coverage)

JLK Inc - Leading the AI Medical Industry with Innovative Technology

CEO

Dongmin Kim

IPO

IPO in Dec. 2019

Foundation

Feb 2014

Locations

South Korea, US, Japan

Employees

88 (as of September, 2022)

Website

www.jlkgroup.com

MEDIHUB



BRAIN

Multiple Neurological Disorder Analysis

MRI / CT / MRA

Hyper-acute/acute stroke, brain aging & dementia,
brain aneurysm



CANCER

Multiple Cancer Type Analysis

Pathology / MRI / Endoscopy / X-Ray / Mammography

12 cancer types including prostate cancer,
breast cancer, stomach cancer, colon cancer



Business Platforms

AI Telemedicine



AI Genetic
Analysis



AI Medical Data

Hello Data



Thank You
