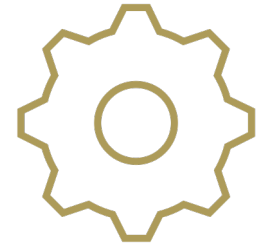




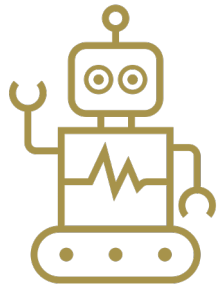
Smart



Fast



Systems



Machine Learning



Computer Vision

# Building Smart and Fast Systems using Machine Learning and Computer Vision.

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Thaleia Dimitra Doudali

Assistant Research Professor @IMDEA Software Institute

# About Me



Born and raised  
in Greece.



2015



Undergrad in ECE at  
NTUA, Athens, Greece.



2021



PhD in CS at  
Georgia Tech, Atlanta, USA.

Advised by Ada Gavrilovska.

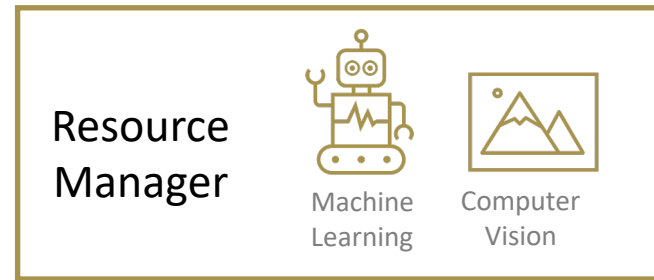
Start: October 2021



Assistant Professor at  
IMDEA, Madrid, Spain.

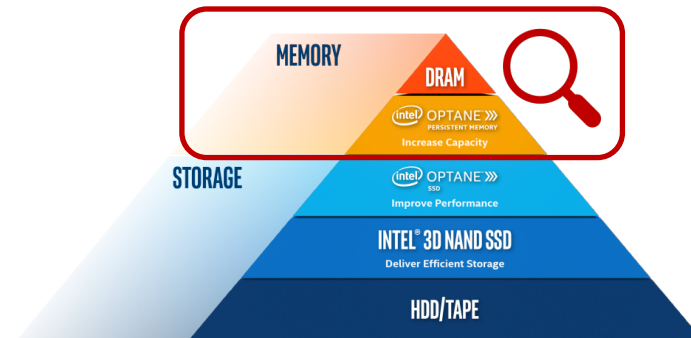
# About My Research

I build software systems that manage heterogeneous hardware resources using machine learning and computer vision.



System Level

Heterogeneous Hardware



# The Era of Data

“More than **65 ZB** of data will be created, captured, copied, and consumed in the world this year.”

Source: International Data Corporation, March 2021.

Exploded  
Data Sizes



Scientific Simulations



Big Data



Artificial Intelligence



Video Analytics

*Data Analytics Pipeline*

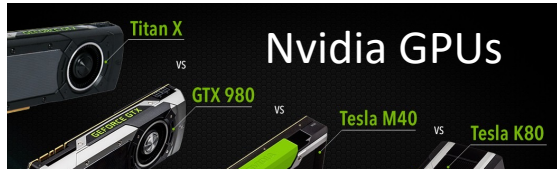
**ZBs of data**



Need for speed and massive storage capacities!

# The Era of Heterogeneous Hardware

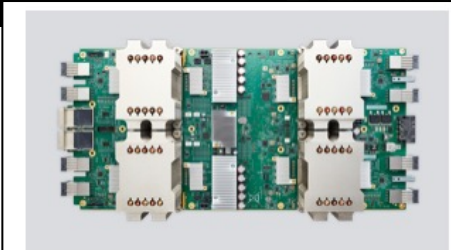
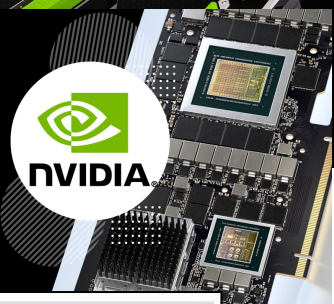

## Compute Acceleration



Titan X  
vs  
GTX 980  
vs  
Tesla M40  
vs  
Tesla K80

Nvidia GPUs

ARE DPUS THE NEXT DATA CENTER REVOLUTION?




Cloud TPU v2

180 teraflops

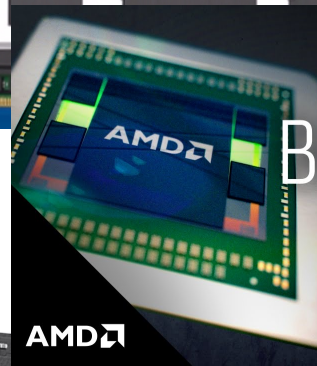
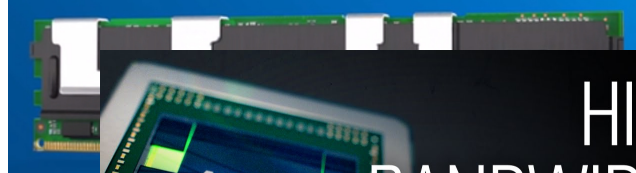
64 GB High Bandwidth Memory (HBM)



## Data Storage Acceleration



intel OPTANE  
PERSISTENT MEMORY



AMD HIGH BANDWIDTH MEMORY



## Network Acceleration

Mellanox Innova™ -2 Flex  
Open Programmable SmartNIC



## Interconnection Standards



Gen-Z Consortium

GEN Z



# Heterogeneity Across Computing Platforms

## Supercomputers



HPCwire

Since 1987 - Covering the Fastest Computers in the World and the People Who Run Them

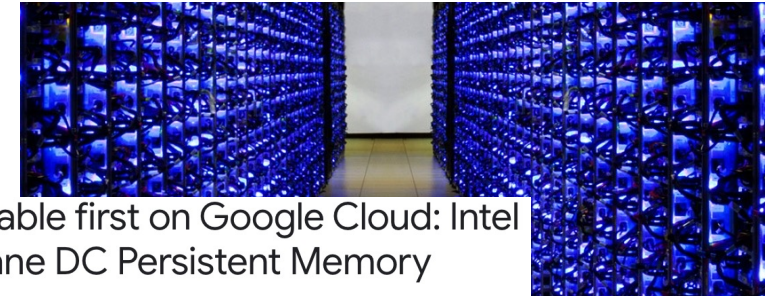
- Home
- Technologies
- Sectors



Application Performance	200 PF
Number of Nodes	4,608
Node performance	42 TF
Memory per Node	512 GB DDR4 + 96 GB HBM2
NV memory per Node	1600 GB
Total System Memory	>10 PB DDR4 + HBM2 + Non-volatile
Processors	2 IBM POWER9™ 9,216 CPUs 6 NVIDIA Volta™ 27,648 GPUs
File System	250 PB, 2.5 TB/s, GPFS™
Power Consumption	13 MW
Interconnect	Mellanox EDR 100G InfiniBand
Operating System	Red Hat Enterprise Linux (RHEL) version 7.4

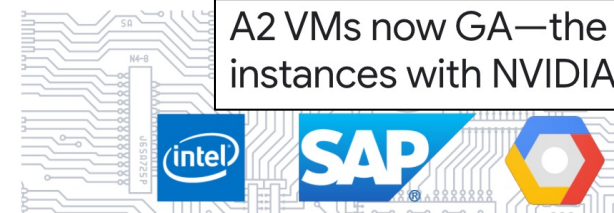


## Datacenters



Available first on Google Cloud: Intel Optane DC Persistent Memory

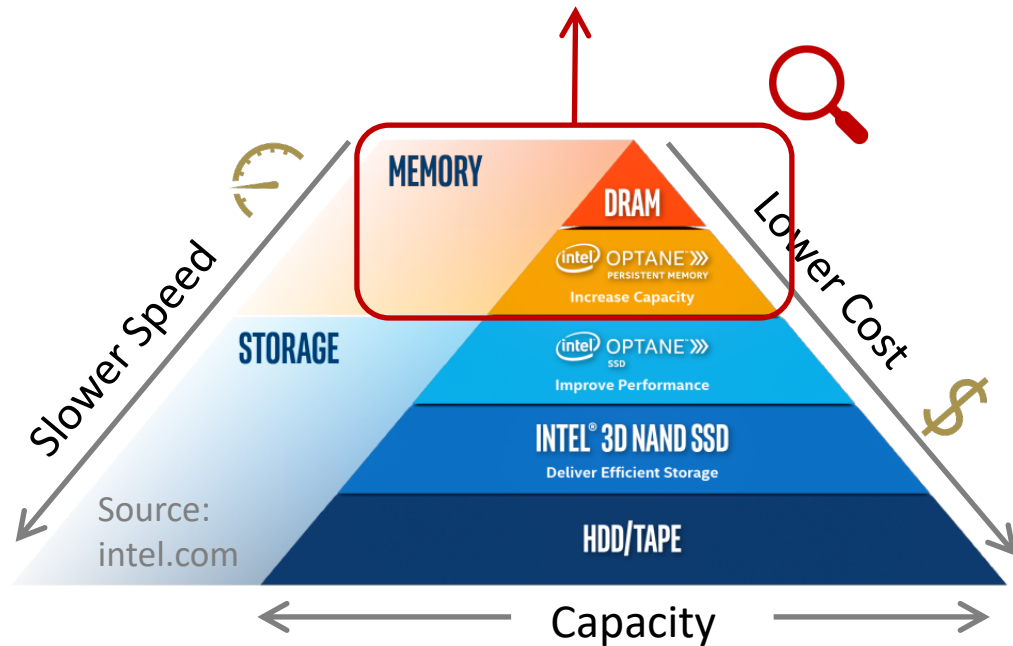
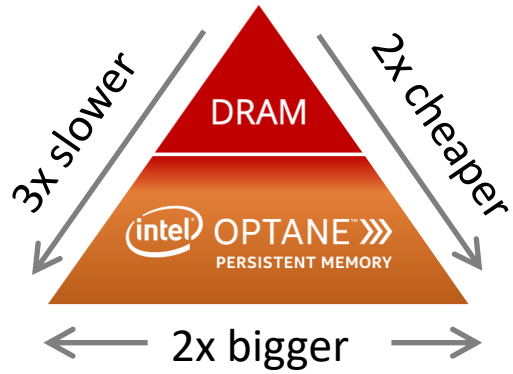
A2 VMs now GA—the largest GPU cloud instances with NVIDIA A100 GPUs

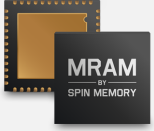



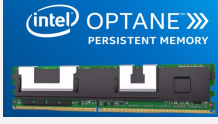


## Personal Devices

70% faster ML accelerators  
New image signal processor  
50% faster CPU  
Faster than any other smartphone chip  
A14 in a smartphone  
16-core Neural Engine  
Machine learning controller  
Best machine learning platform in a smartphone  
6-core CPU  
11 trillion operations per second on the Neural Engine  
Improved memory compression  
11.8 billion transistors  
Secure Enclave  
50% faster GPU  
Faster than any other smartphone chip  
4-core GPU

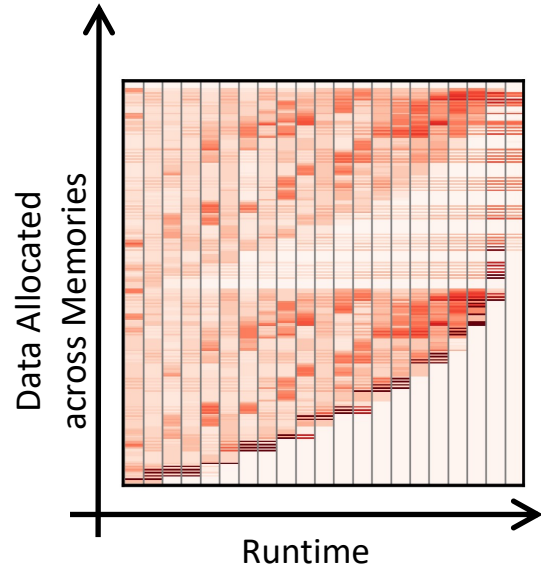
# Heterogeneity Trade-offs



Characteristic	Technology	Hardware Vendors
Low Latency	MRAM	 
High Bandwidth	HBM	 
Persistence	PMEM	

Examples of other heterogeneous memory technologies.

# Building Software to Maximize the Hardware Efficiency



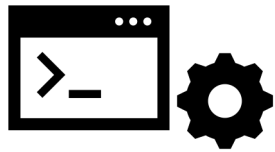
**Dynamic Data Movements!**



It is a **complex decision mix** to manage the data allocated across memories.

E.g., Which / How much / Where / When to move data?

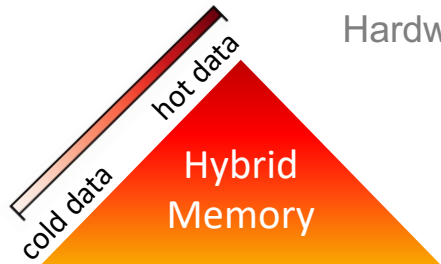
Application Level



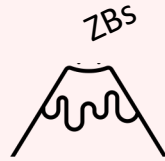
Resource Management

Software System Level

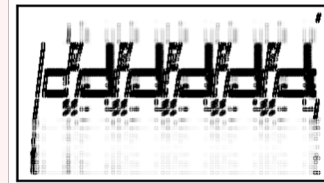
Hardware Level



**Why do we need smarter and faster systems?**



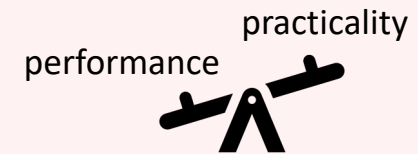
Application data sizes



Complex data access patterns



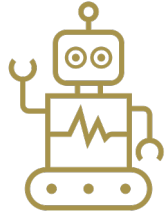
Exploded system parameter space



Hard to balance



# Talk Outline



## **Building *Smart* Systems**

Foundations for practical Machine Learning (ML)-based Management



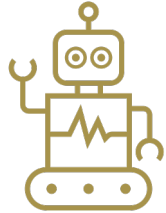
## **Building *Fast* Systems**

Reducing ML-based Management Overheads with Visualization



Open Research Questions

# Talk Outline



## **Building *Smart* Systems**

Foundations for practical Machine Learning (ML)-based Management



## **Building *Fast* Systems**

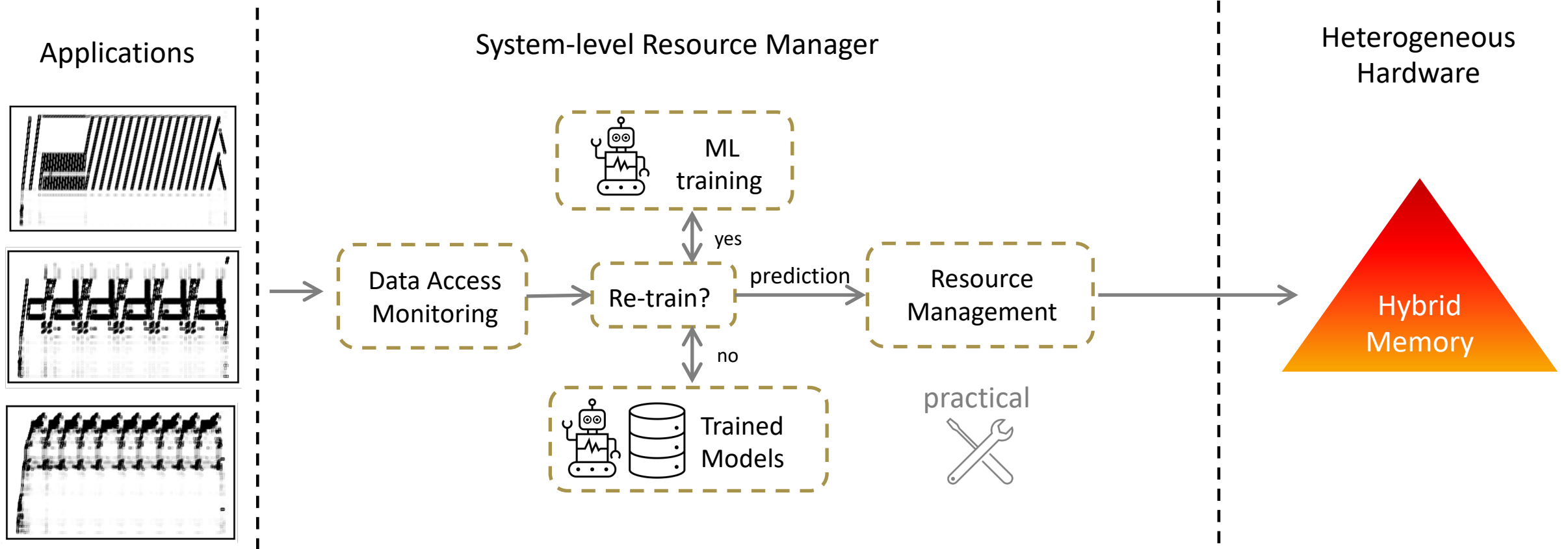
Reducing ML-based Management Overheads with Visualization



Open Research Questions

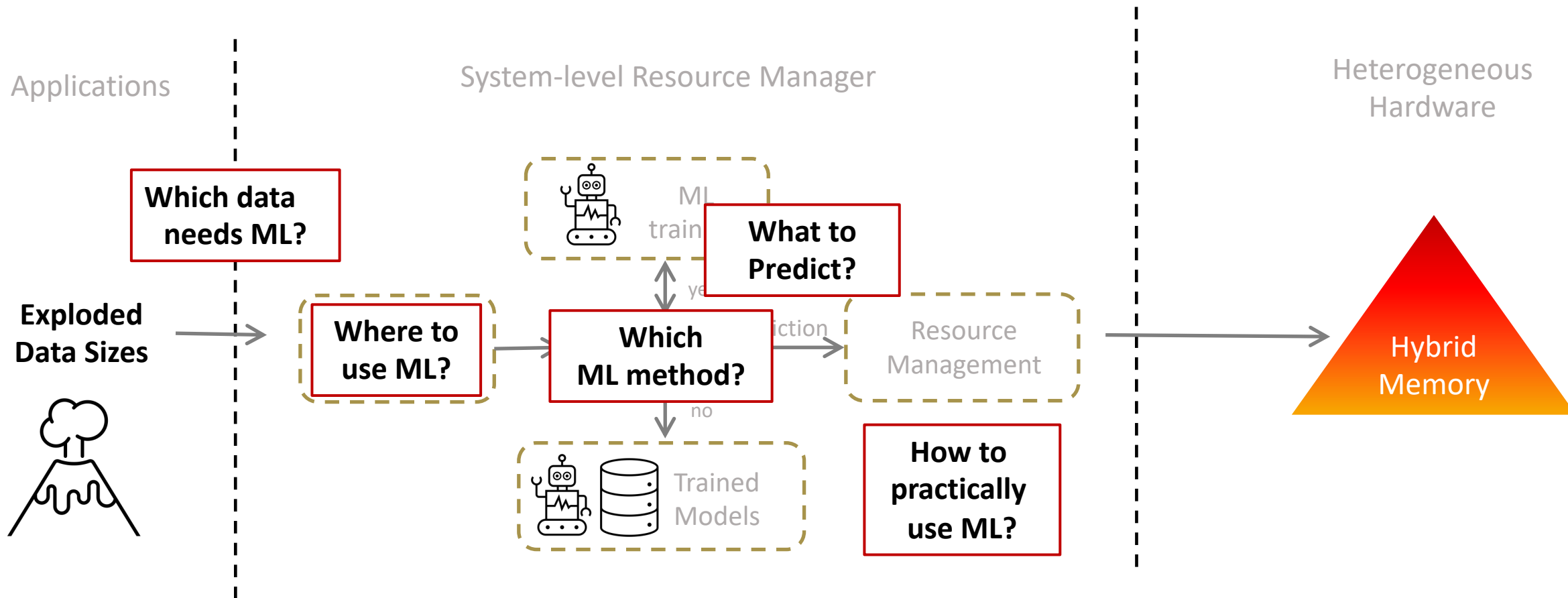
# The Vision

ML-augmented heterogeneous resource manager.



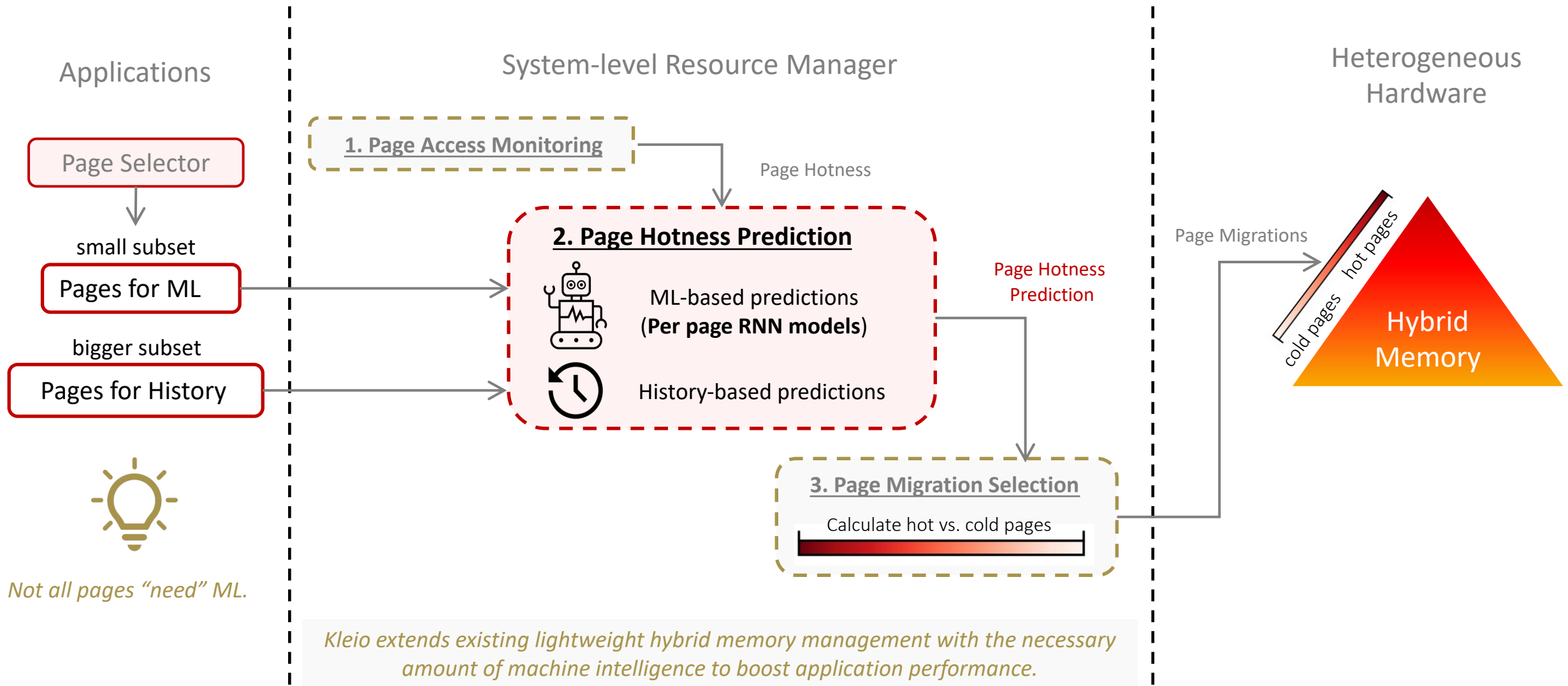
# Contributions Towards the Vision

Laying the grounds for the *practical* integration of ML.



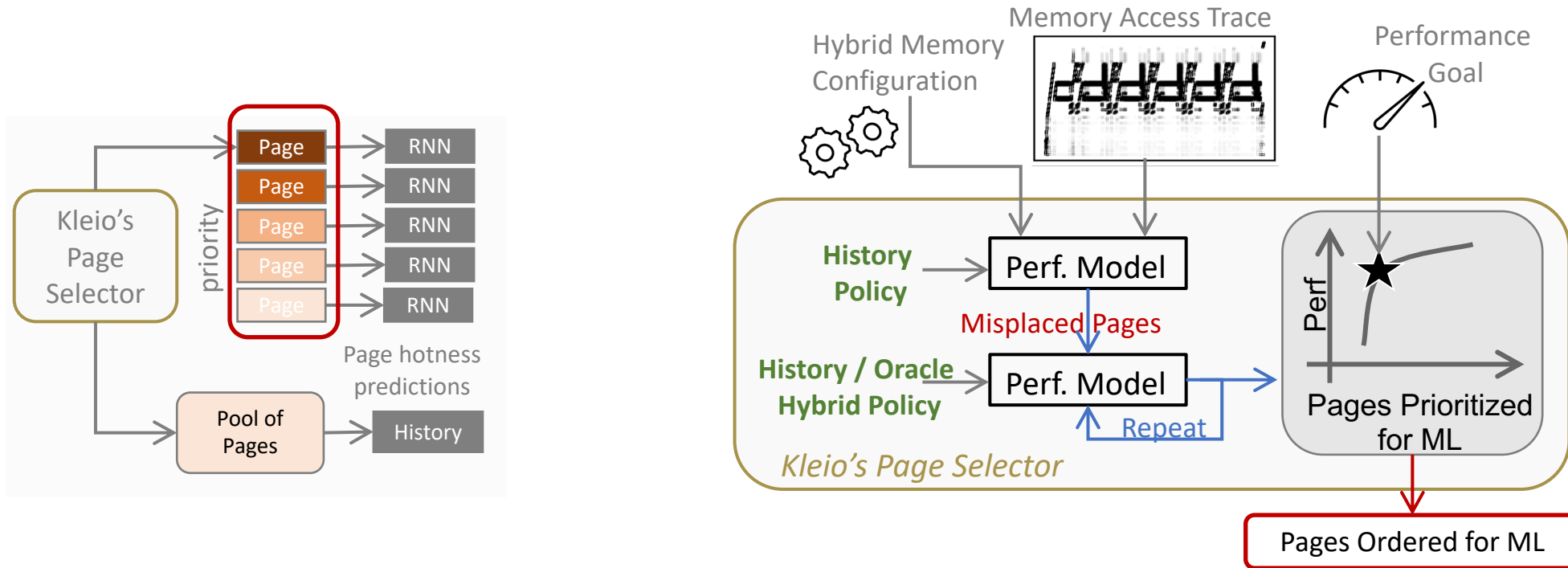
# System design of Kleio

Kleio is a hybrid memory page scheduler with machine intelligence. [HPDC '19]



# The Key to Success

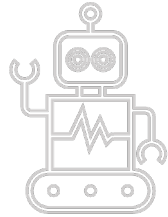
Selecting a small page subset for ML-based management.



*It is not a lightweight process, but necessary to maximize the effects of ML on application performance.*

Kleio bridges the **performance** gap left by existing solutions by 80%, on average.

# Talk Outline



## **Building *Smart* Systems**

Foundations for practical Machine Learning (ML)-based Management



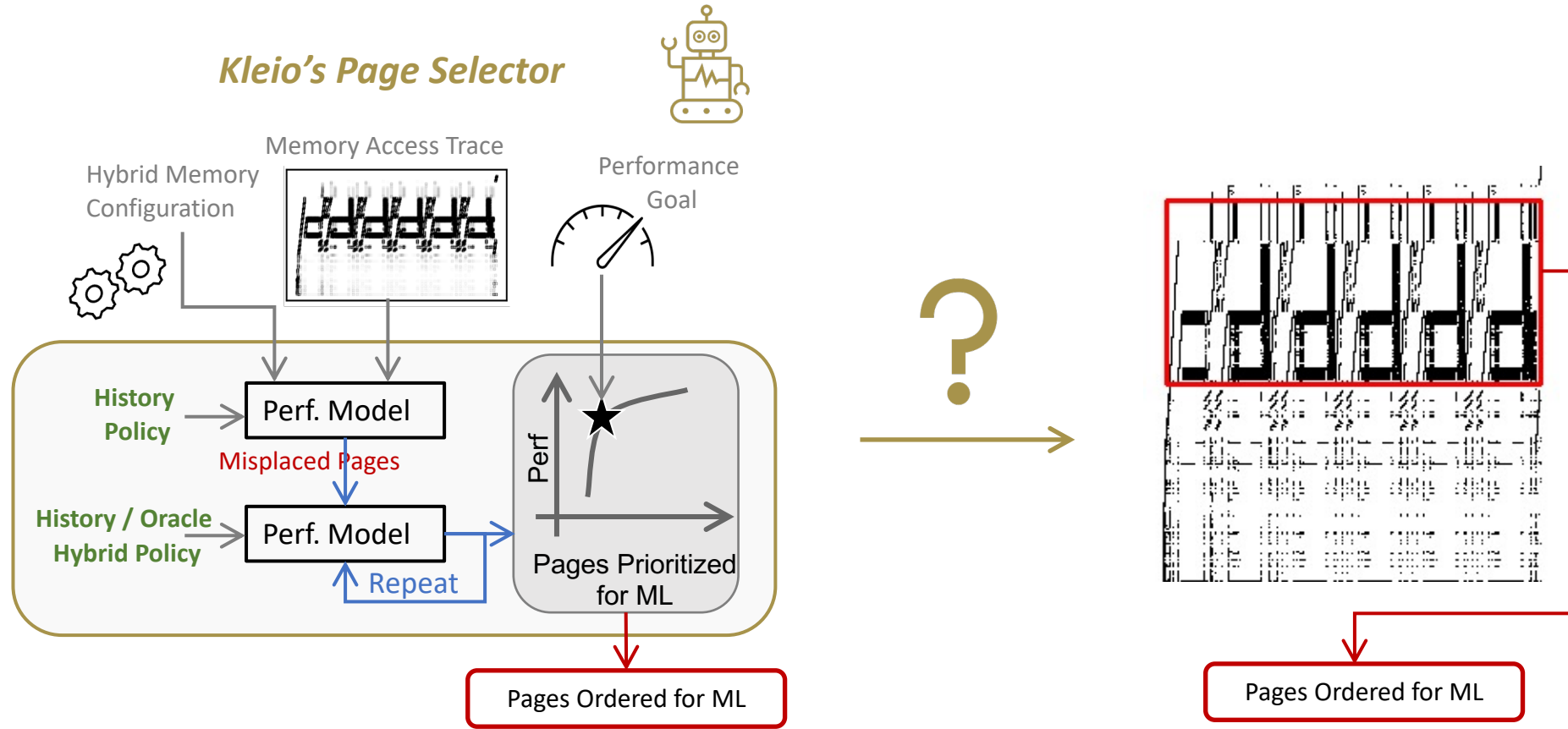
## **Building *Fast* Systems**

Reducing ML-based Management Overheads with Visualization



Open Research Questions

# Reducing Operational Overheads of ML-based Management

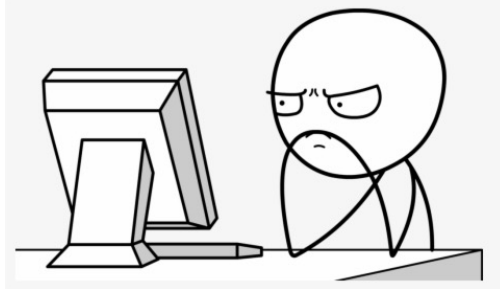


Can we accelerate the page selection process via image-based decisions?



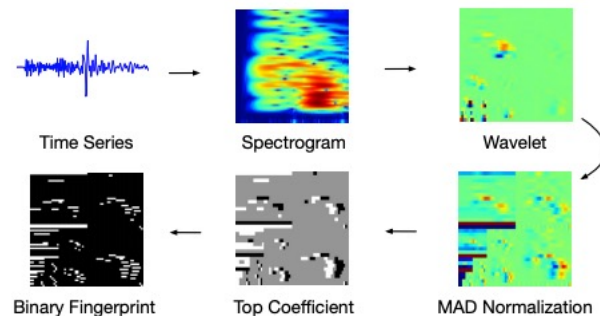
# Why images?

## Personal Experience



*I came up with great observations and insights, by visualizing memory access patterns.*

## Feature Extraction



**Figure 3:** The fingerprinting algorithm encodes time-frequency features of the original time series into binary vectors.

Source: Kexin Rong et al. at VLDB '18.

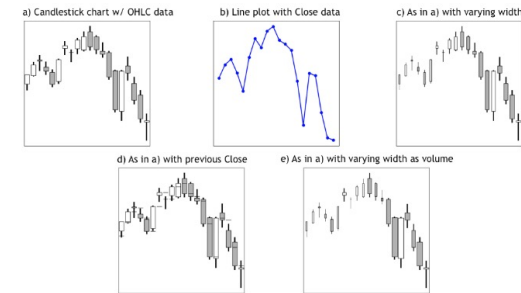
**Earthquake Detection:** Extract Frequencies of Seismic Waves.

## Image-based ML Classifiers



**Figure 1:** Typical workstation of a professional trader. Credit: Photoagriculture / Shutterstock.com.

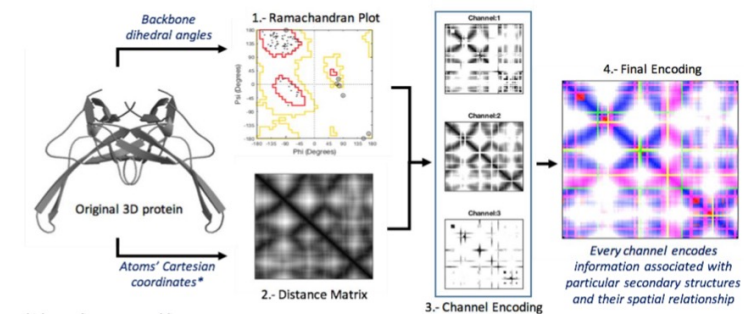
Source: J.P. Morgan AI labs.



**Figure 4:** Various visual representations of the same time-series data.

**Finance:** Trading by learning time series data as images.

## From Multi-fold Representation to Image Encoding



21

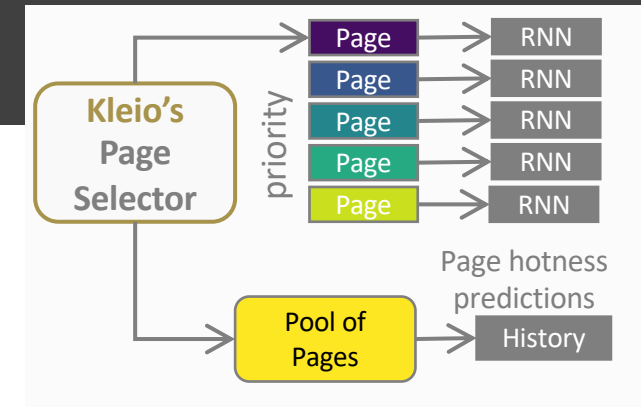
T. Estrada, J. Benson, H. Carrillo-Cabada, A. Razavi, M. Cuendet, H. Weinstein, E. Deelman, and M. Taufer. *Graphic Encoding of Proteins for Efficient High-Throughput Analysis. ICPP 2018.*



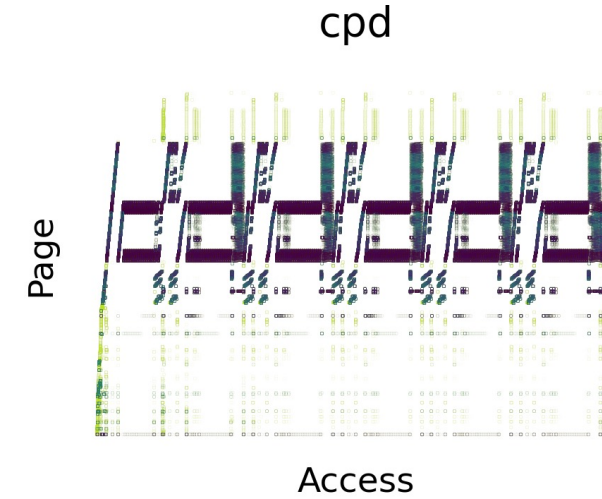
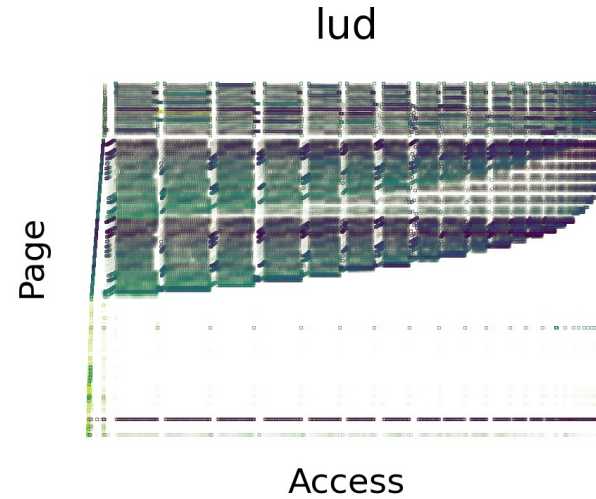
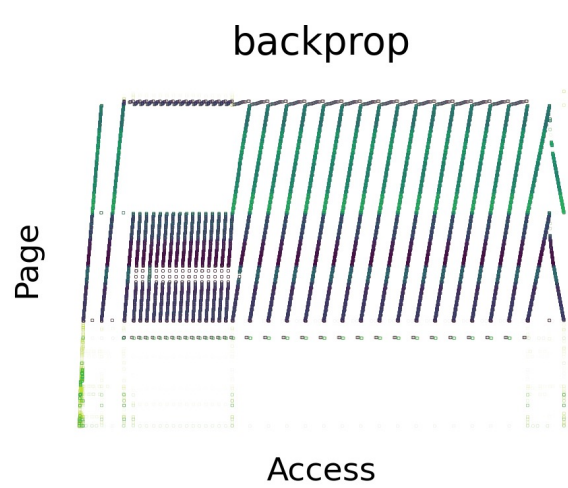
Source: Presentation from Michela Taufer.

**Bioinformatics:** Learn protein functions.

# Visualizing Pages Selected for ML



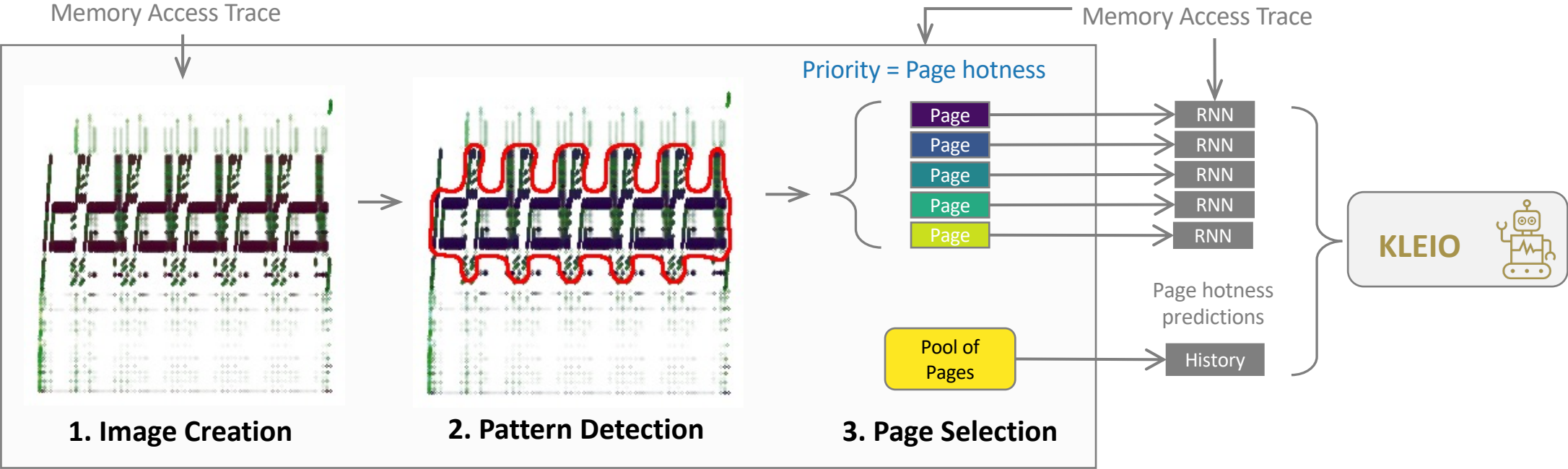
High Priority  Low Priority



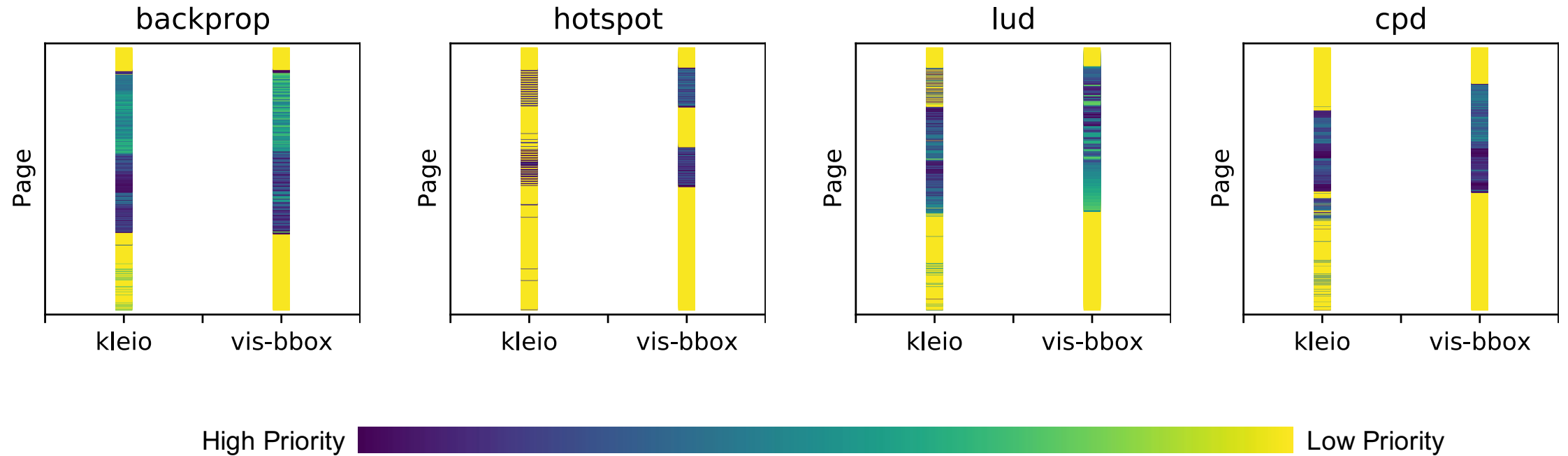
*Neighboring pages that are part of distinct access patterns across time receive similar priority for ML.*



# Towards Image-based Page Selection



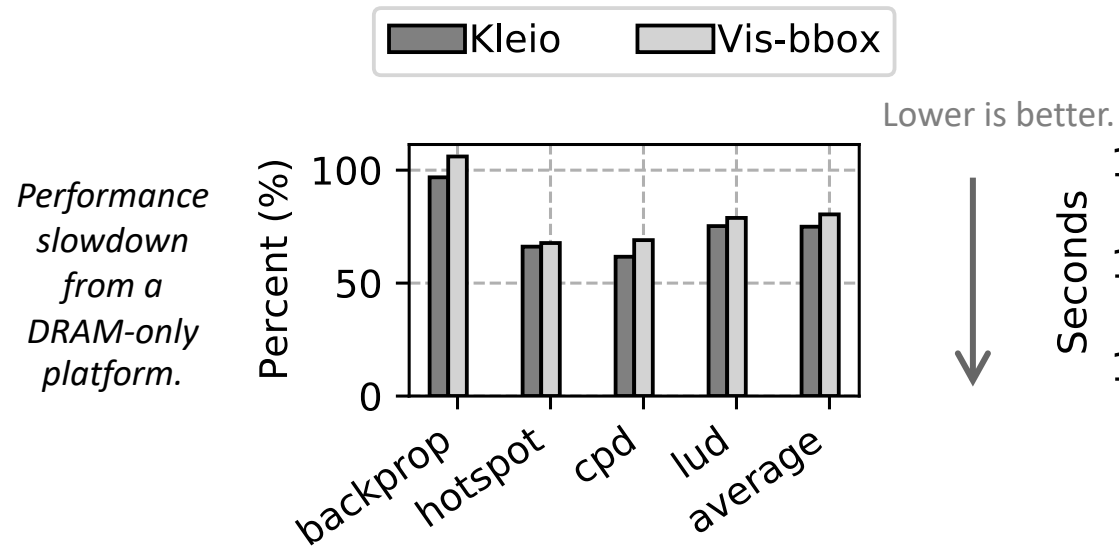
# Page Selection Comparison



*Similar page orderings between our initial approach (vis-bbox) and the performance-based selection of Kleio.*

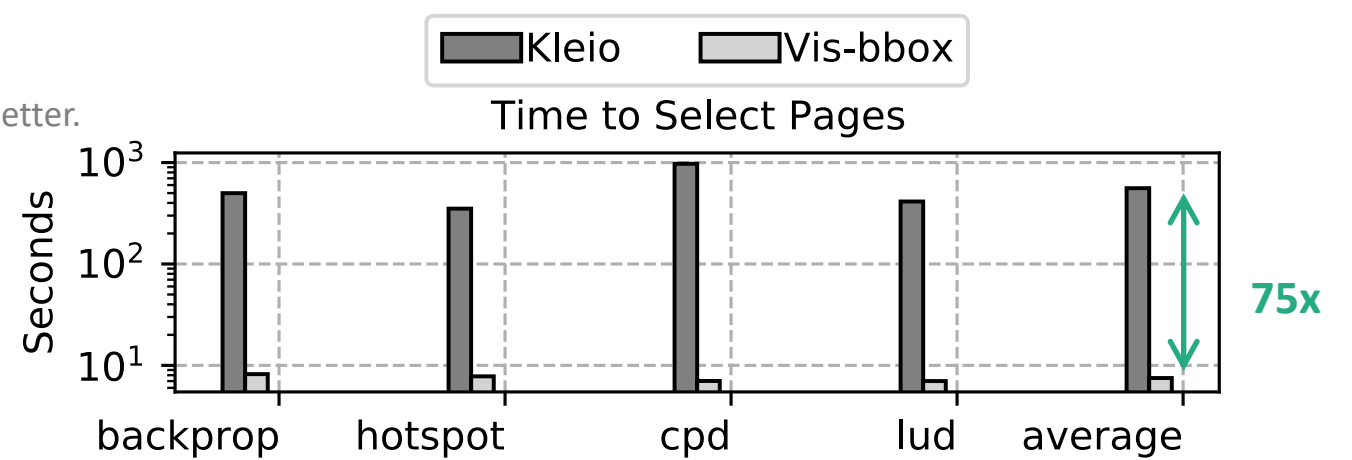
# Performance Evaluation

## 1. Application Performance



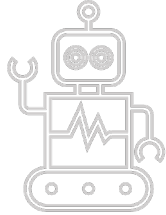
Comparable application performance levels.

## 2. Page Selection Time



Accelerates the time to select pages for ML-based management.

# Talk Outline



## **Building *Smart* Systems**

Foundations for practical Machine Learning (ML)-based Management



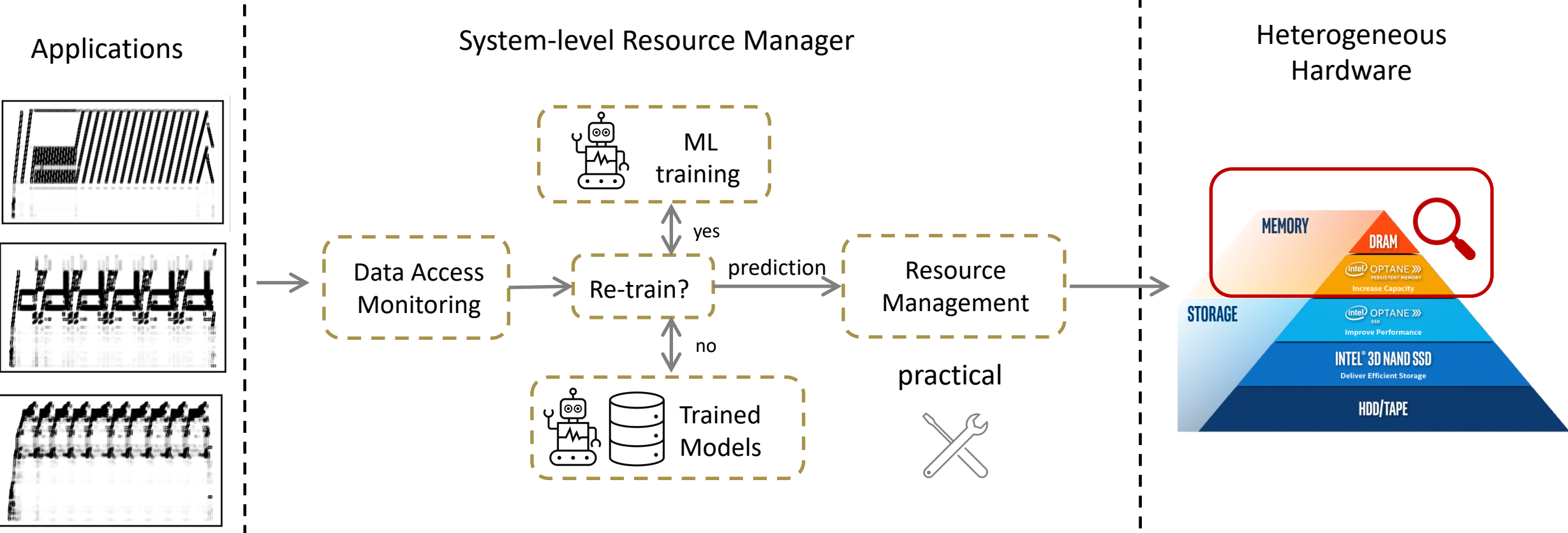
## **Building *Fast* Systems**

Reducing ML-based Management Overheads with Visualization



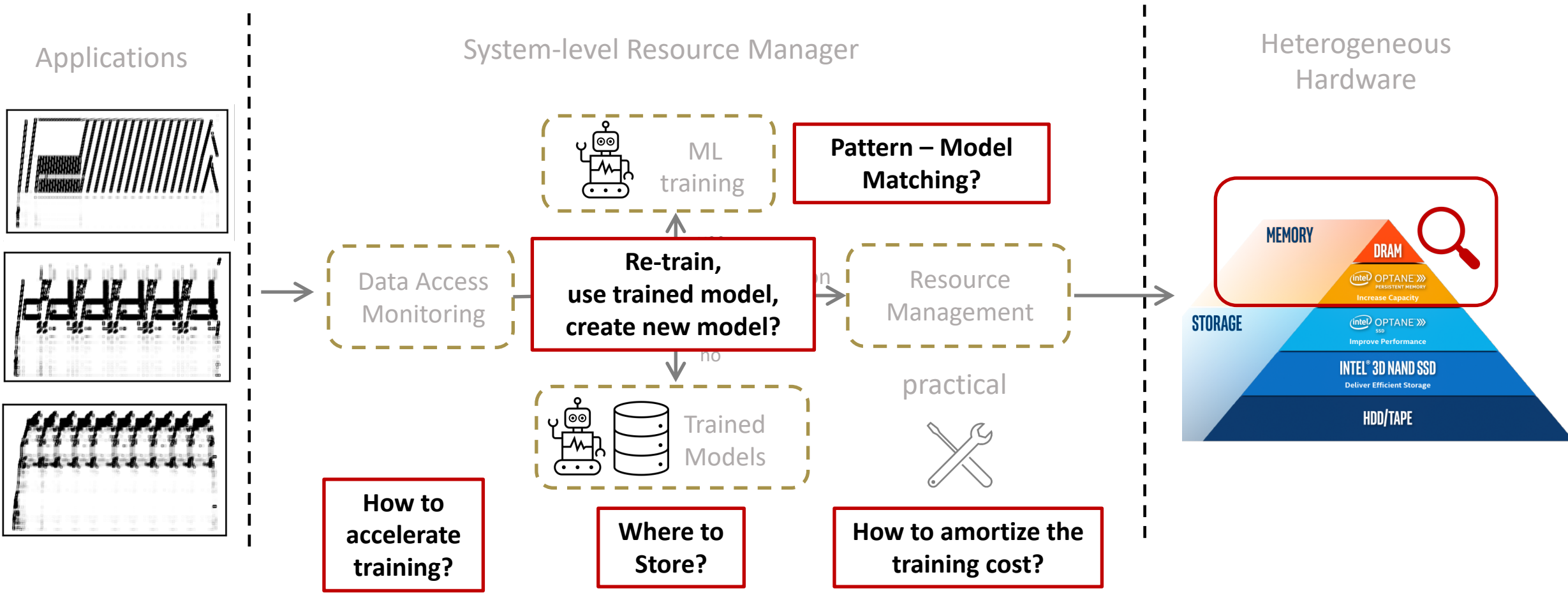
## **Open Research Questions**

# ML-augmented Heterogeneous Resource Manager



# Remaining Challenges

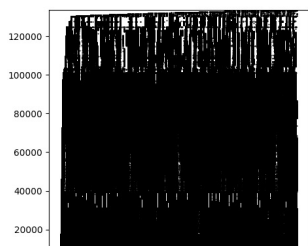
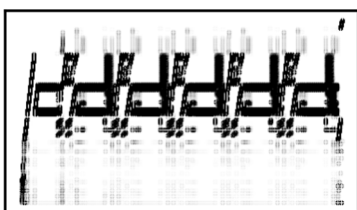
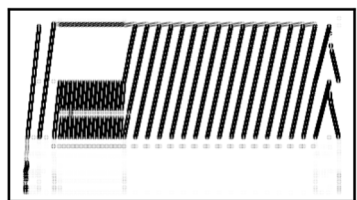
Fully integrated adaptive resource manager.





# Computer Vision + Machine Learning Pattern Recognition.

## Applications

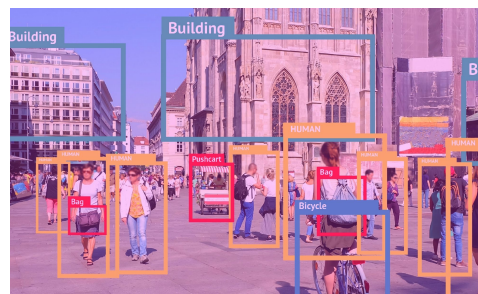
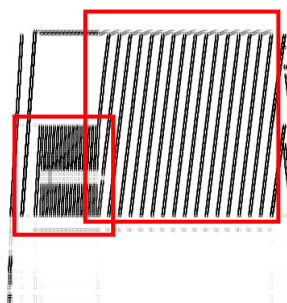


How to zoom-in?

Image Resolution?

## System-level Resource Manager

### Pattern Recognition



Pattern Class

### Example Use Cases

Select ML model

Take ML related action

Resource Management Decision

Build an "ImageNet" for memory access patterns.

Train classifiers to take actions upon recognizing a pattern.

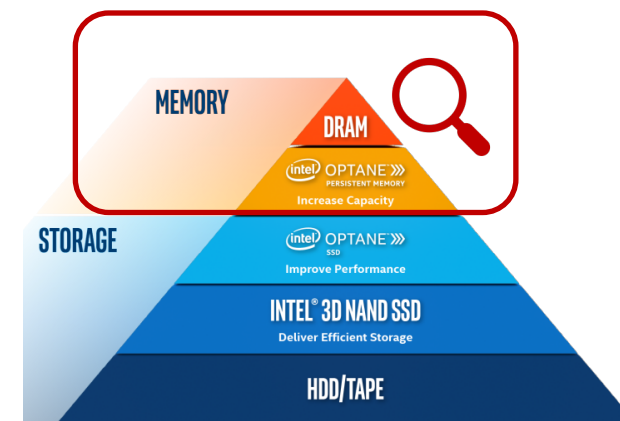
What Classes to define?

Metadata?

Storage?

Label guidelines for open contributions?

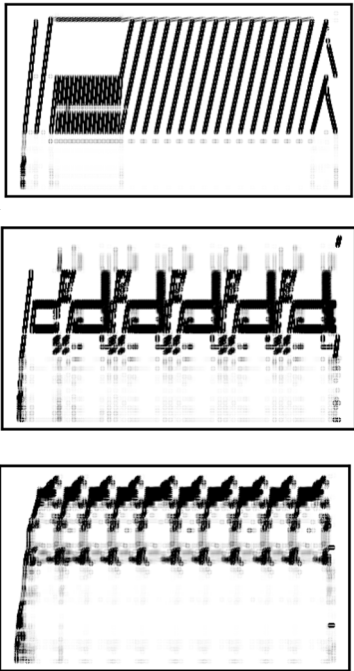
## Heterogeneous Hardware



# Computer Vision + Machine Learning

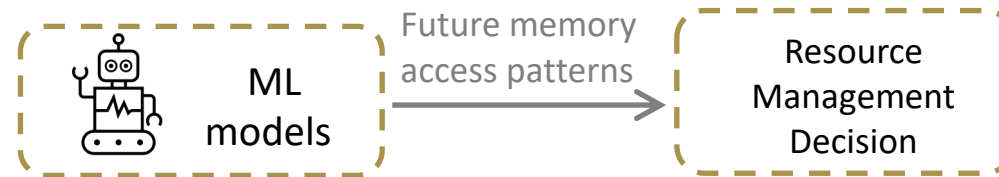
## Pattern Prediction.

### Applications

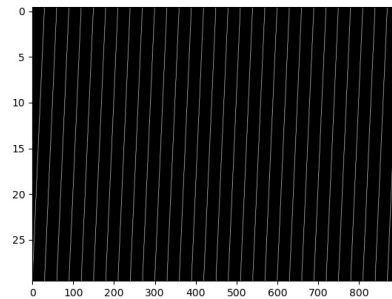


### System-level Resource Manager

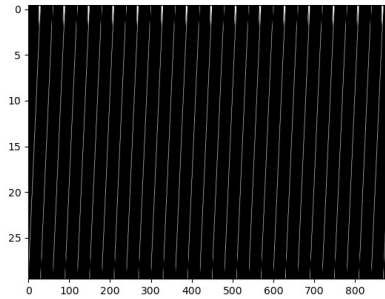
#### Pattern Prediction



Ground truth

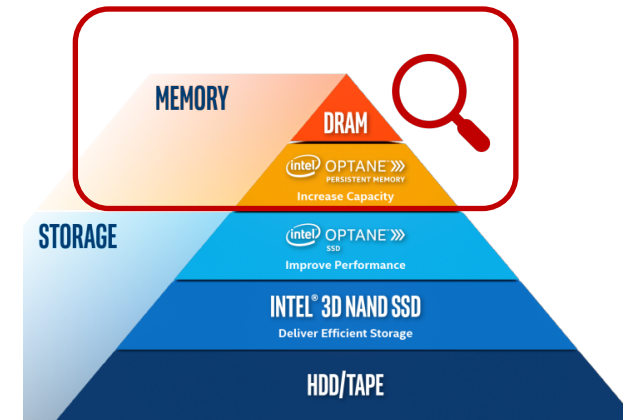


Prediction



Prediction result of a Convolutional + Recurrent Neural Network.

### Heterogeneous Hardware



Compare with RNN-only?

1 model per app or per pattern?

Page-to-image mapping

“ML models over images have better predictive capabilities.”

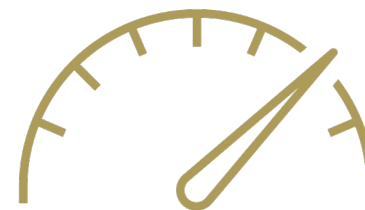
From OSDI '21 Keynote from J.P. Morgan AI labs.



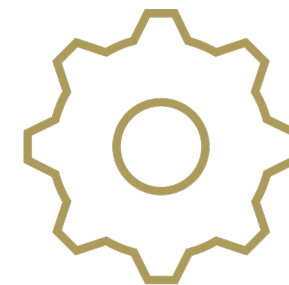
Scan this to find more about my work.



Smart

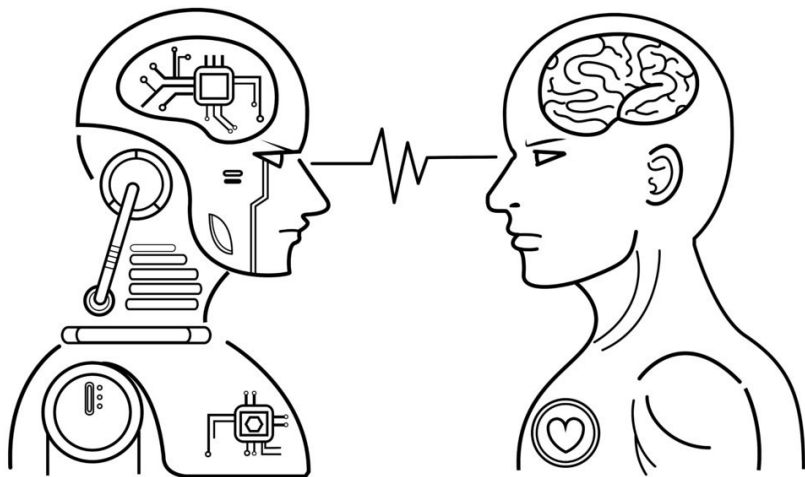


Fast

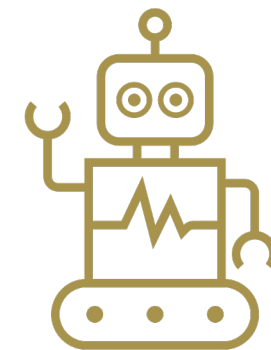


Systems

Artificial vs HUMAN Intelligence



How can we use our human intelligence to build **practical** systems that leverage machine learning and computer vision?



Machine Learning



Computer Vision