

Learning with Real and Unreal Data in the Era of Foundation Models

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November 08, 2023



30 Years Ago ...

RETRATO FALADO

Computador vai desenhar rosto de suspeitos

Três jovens criaram um programa que garante maior rapidez e precisão no momento de identificar pessoas acusadas de crimes

EUNICE LEME

Correspondente

Rio Grande — Os alunos do curso de Processamento de Dados do Colégio Técnico Industrial (CTI) de Rio Grande estão desenvolvendo um programa de computação que permite a criação do retrato falado de uma pessoa. O programa poderá em breve ajudar a polícia a identificar criminosos com maior rapidez e precisão. O sistema está em fase de aperfeiçoamento e a previsão é de que no próximo ano já esteja adequado às necessidades da Polícia Civil de Porto Alegre. O software é inédito no país. O que existe até hoje são sistemas importados, segundo afirma Delnir Monteiro Lemos, professor do curso. Em Brasília, a Polícia Federal também estuda a adoção do programa criado em Rio Grande. Se for aprovado, a PF pode utilizá-lo em seus escritórios de todo o país.



Na tela: a figura de uma pessoa fica mais precisa



Feito à mão: Paulo Severo usa um kit para identificar



Marco Gomes

Alessandro Bicho

Myself

Criativos: Marco (de óculos), Alessandro (sentado) e Rogério querem ver o projeto usado em investigações

O projetor de imagens é hoje a principal arma

O uso do computador no desenho de retratos falados reduz etapas e torna as informações precisas, admite o desenhista Paulo Severo. Atualmente à mão este tipo de trabalho era feito para toda a Polícia Civil gaúcha. Na elaboração de retratos falados, Severo utiliza um retroprojetor, um livro mostruário e um kit composto por lâminas de projeção, com figuras de cabelos, sobrancelhas, olhos, narizes, lábios, barbas, bigodes, orelhas e formatos de rostos variados. O livro é mostrado às testemunhas que vão selecionando os traços que mais se assemelham ao acusado de um crime. Com a superposição das lâminas, o desenhista forma uma figura, que, projetada numa parede, é reproduzida numa folha de papel, através de um desenho.

No programa de computador, explica, a imagem criada a partir da

Retrato falado é feito com informações das testemunhas




O programa possui 600 imagens de cada detalhe do rosto



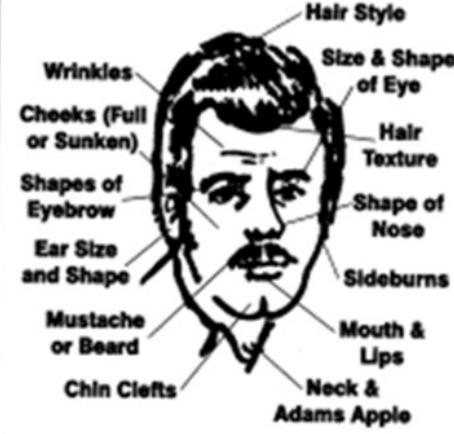
IBM Intelligent Video Analytics (2013)



People Search based on a Suspect Description Form

SEX	RACE	AGE	HEIGHT	WEIGHT	TYPE OF WEAPON		
							
						HAIR/FACIAL HAIR	HAT (color, type)
						GLASSES (type)	TIE
						TATTOOS	COAT
						COMPLEXION	SHIRT
						SCARS/MARKS	PANTS/SHOES
HARRISBURG EMERGENCY DIAL 1-911			DON'T HANG UP				
POLICE	FIRE						
MEDICAL							
NON-EMERGENCY 717-787-3199							
AUTO MAKE MODEL, COLOR		LICENSE NUMBER	DIRECTION OF ESCAPE	TIME OF DEPARTURE			

FACIAL APPEARANCE



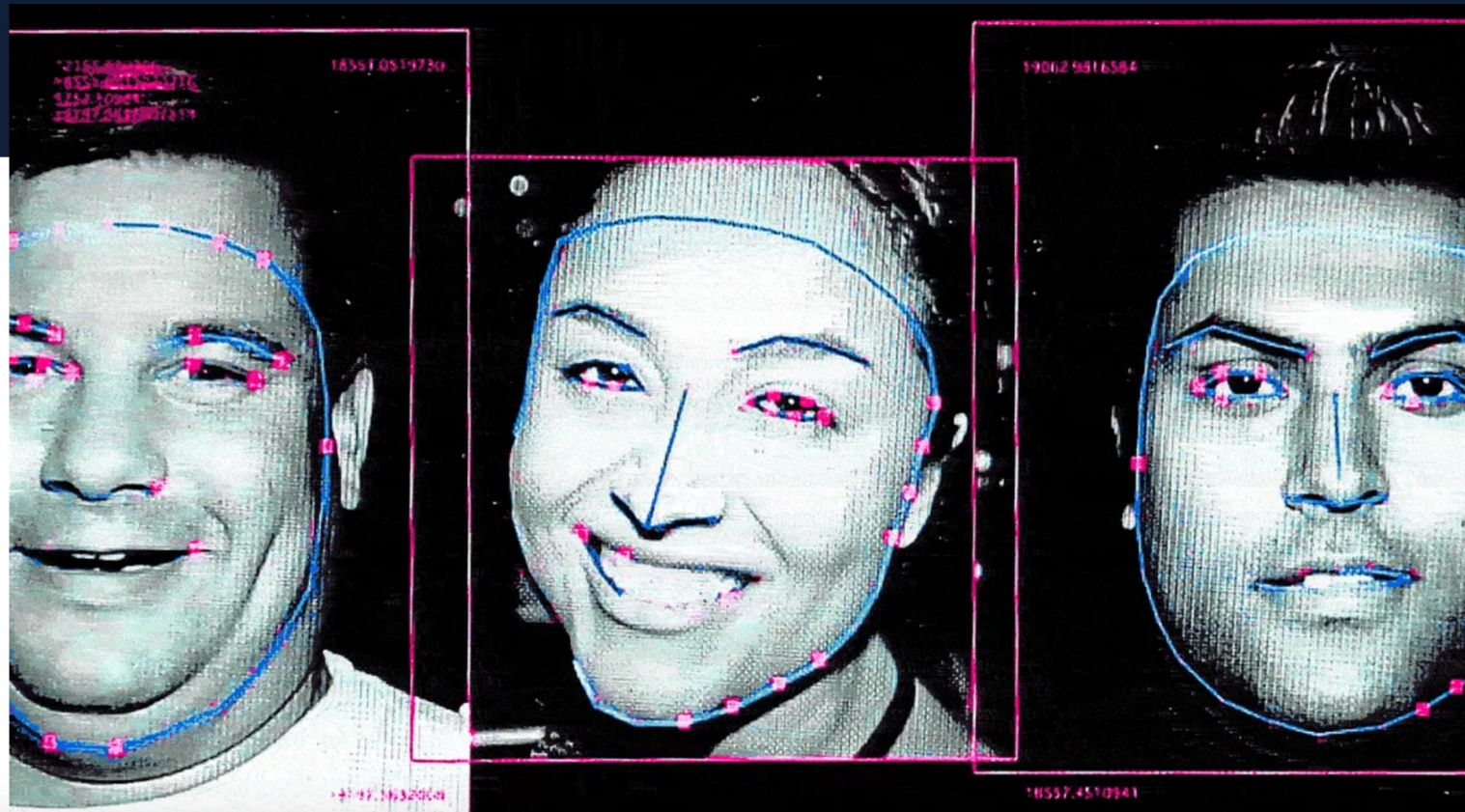
WRITE BELOW SPECIFIC FACIAL DETAILS-ONLY WHAT YOU DEFINITELY REMEMBER

WHAT DID SUSPECT SAY?


TECH & MEDIA

Facial recognition's 'dirty little secret': Millions of online photos scraped without consent

People's faces are being used without their permission, in order to power technology that could eventually be used to surveil them, legal experts say.



Facial Recognition Is Accurate, if You're a White Guy

 Share full article



By **Steve Lohr**

Feb. 9, 2018

Facial recognition technology is improving by leaps and bounds. Some commercial software can now tell the gender of a person in a photograph.

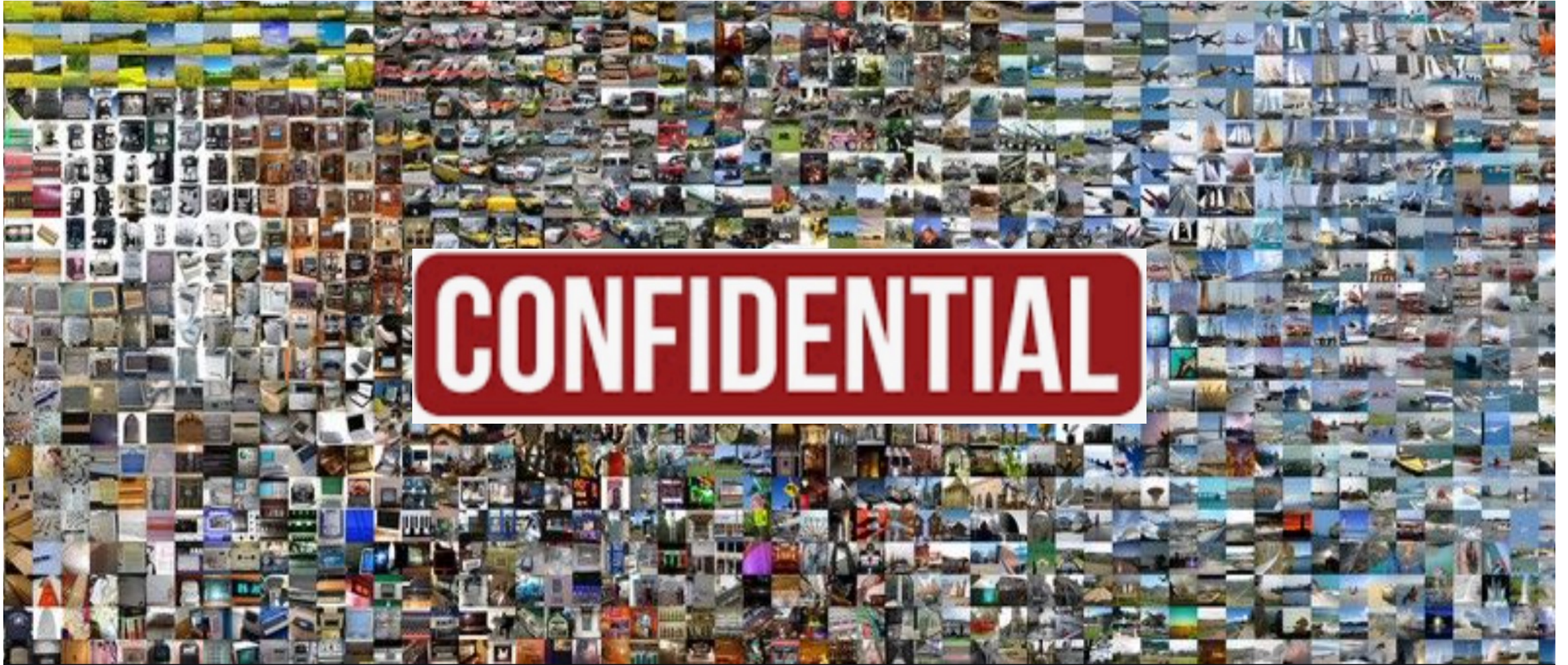
When the person in the photo is a white man, the software is right 99 percent of the time.

But the darker the skin, the more errors arise — up to nearly 35 percent for images of darker skinned women, according to a new study that breaks fresh ground by measuring how the technology works on people of different races and gender.



Era of Foundation Models: Pre-training on Massive Datasets of Web Images and Videos

Google JFT -- 3Billion images



Promising work-around: synthetic data

Embodied Perception



- In addition to privacy, synthetic data can address the issue of training data scarcity for certain applications

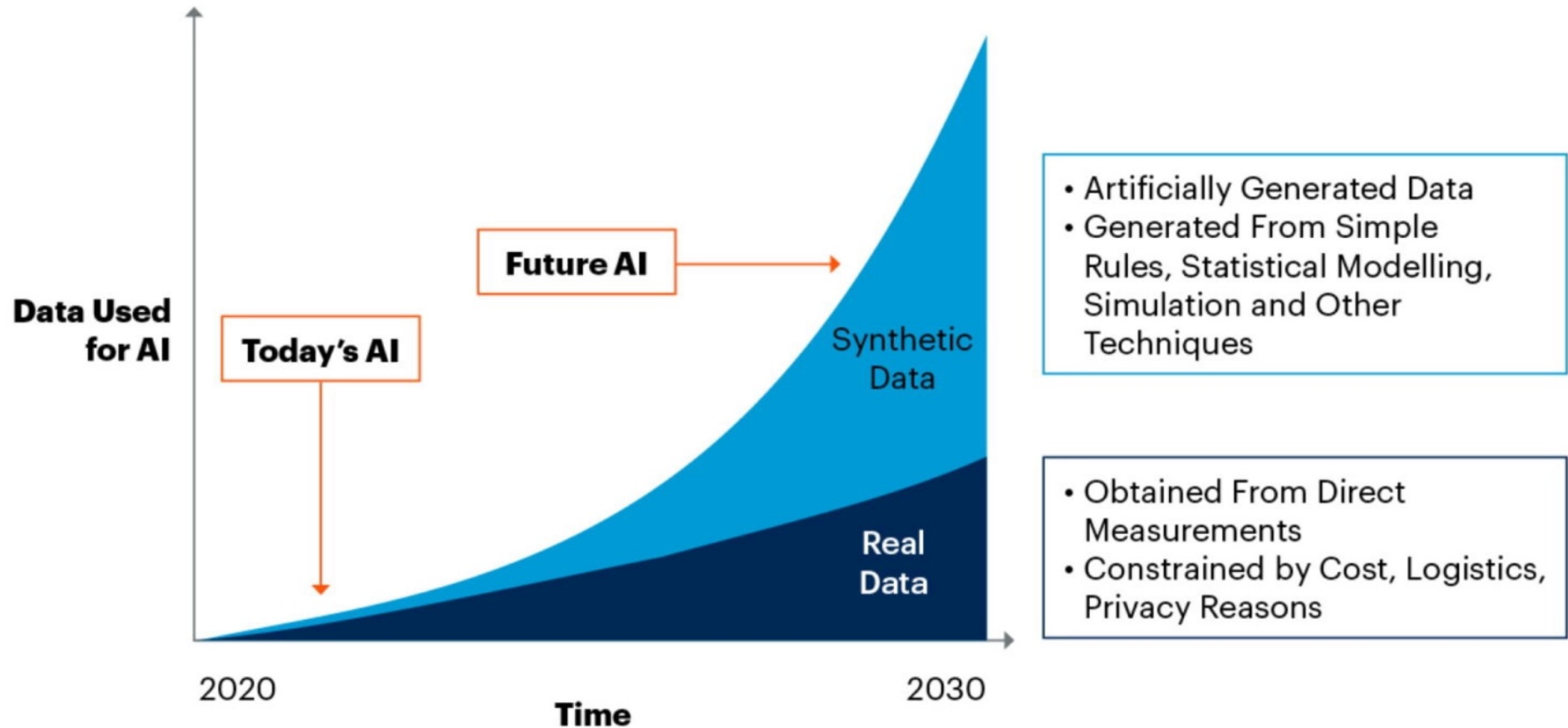
Face simulation



Autonomous Driving



By 2030, Synthetic Data Will Completely Overshadow Real Data in AI Models



Source: Gartner

750175_C

Outline:

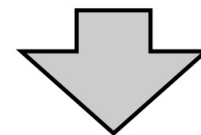
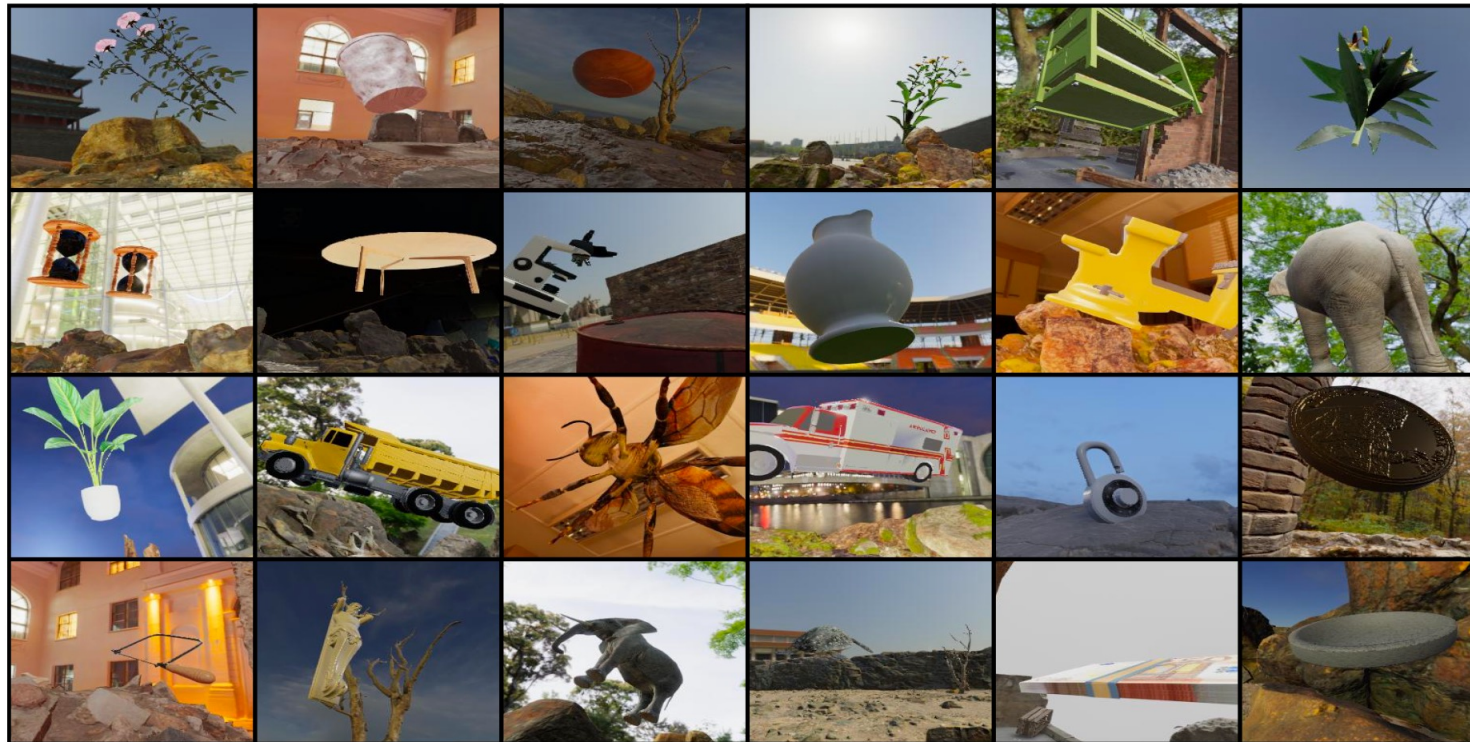
- Pre-training based on Synthetic Data for Vision Tasks
- Synthetic Data for Specialization of Large Language Models
- Application: AI Commentary @ Wimbledon and US Open 2023

Sim2Real domain
adaptation
(same label set)

Our Setting:

New problem: synthetic
Data Pretraining and
Transfer to Diverse
Downstream Tasks
(disjoint label set)

Synthetic Data Pre-training



Downstream Tasks from Various Domains (Real Images)



ChestX



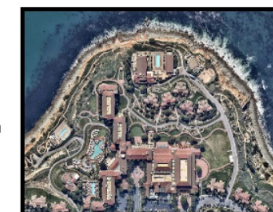
Sketch



Flowers

**SVHN**





■ ■



EuroSAT

Observation: Different simulation parameters have different effects on different downstream tasks

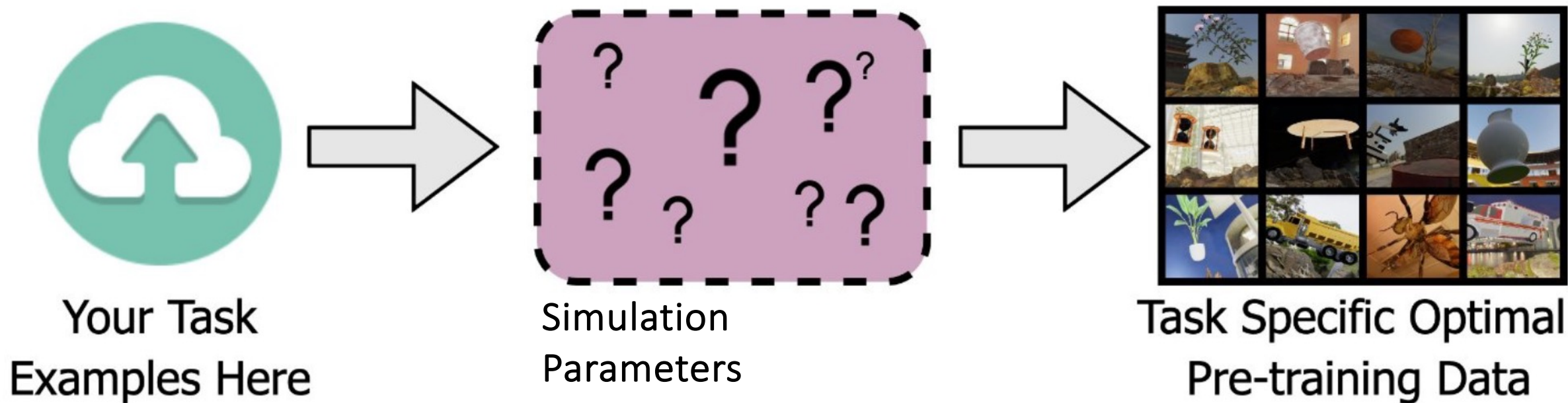
Resnet-50, linear probing

Pretraining Data Variations	Downstream Accuracy			
	EuroSAT	SVHN	Sketch	DTD
Pose	87.01	28.49	37.89	37.39
+Lighting	88.57	32.36	38.81	40.32
+Blur	 90.20	35.58	35.53	37.66 
+Materials	84.54 	44.84	30.81	 38.51
+Background	80.44	29.93	14.60	32.39

Suggests best parameter settings for synthetic data are task-specific

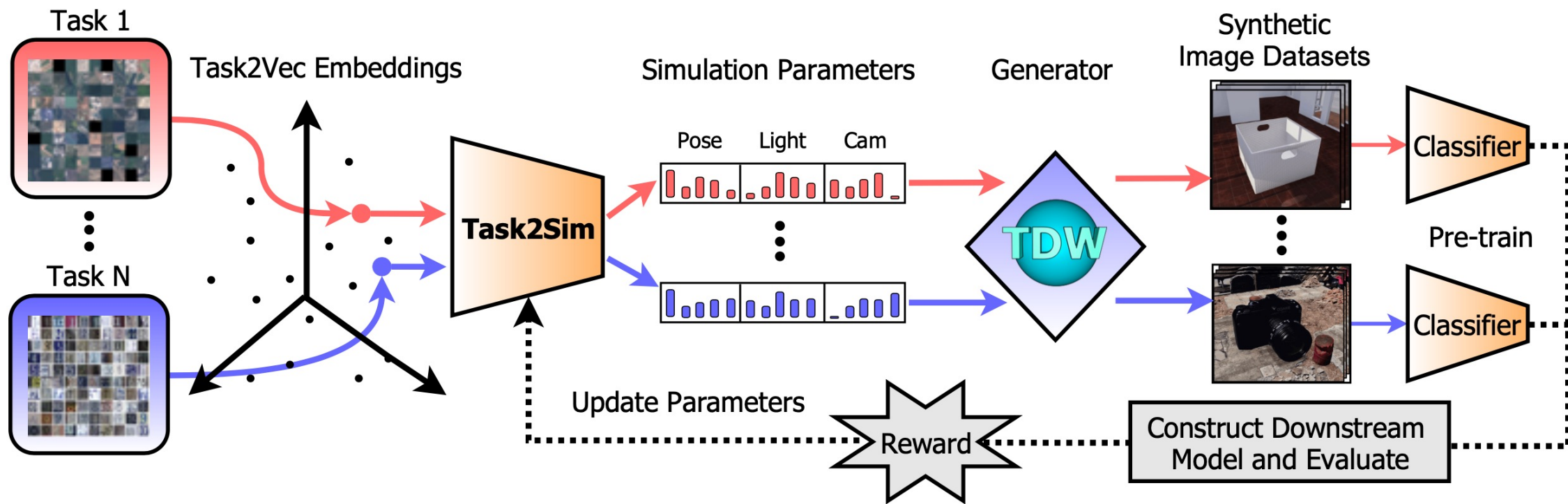
Task2Sim: Towards Effective Pre-training and Transfer from Synthetic Data (CVPR 2022)

S. Mishra, R. Panda, C. Phoo, R. Chen, L. Karlinsly, K. Saenko, V. Saligrama, and R. Feris



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Task2Sim: Simulation Parameters

Data parameterization : 8 discrete-valued binary scene parameters controlling variation



Cam distance



Materials



Background



Focus Blur



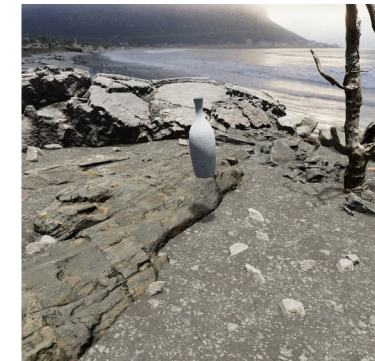
Light color



Obj Rotation



Light Direction

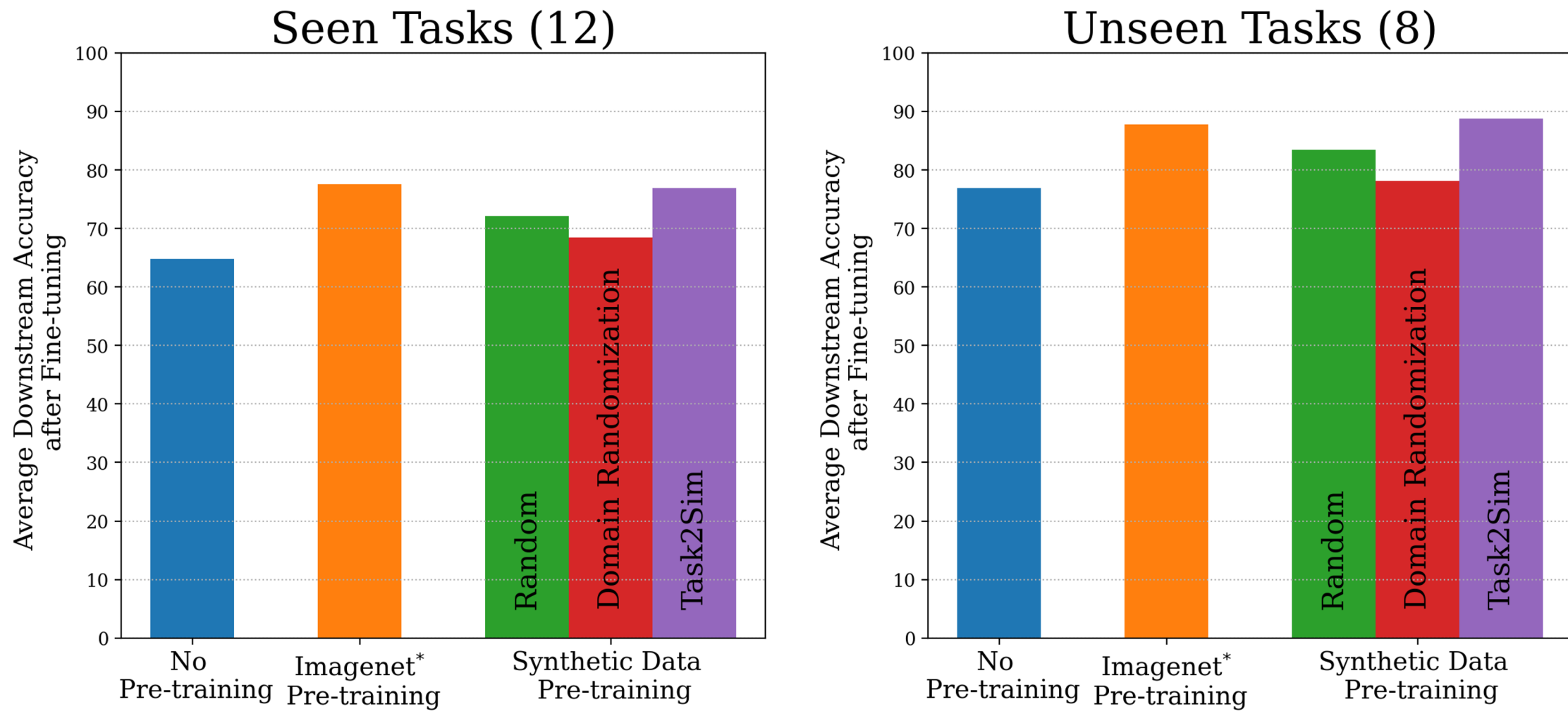


Light Intensity

Experiments:
20 downstream tasks
from various domains

Category	Dataset	Train Size	Test Size	Classes
Natural	CropDisease [39]	43456	10849	38
	Flowers [42]	1020	6149	102
	DeepWeeds [44]	12252	5257	9
	CUB [65]	5994	5794	200
Satellite	EuroSAT [18]	18900	8100	10
	Resisc45 [4]	22005	9495	45
	AID [75]	6993	3007	30
	CactusAerial [34]	17500	4000	2
Symbolic	Omniglot [30]	9226	3954	1623
	SVHN [40]	73257	26032	10
	USPS [21]	7291	2007	10
Medical	ISIC [7]	7007	3008	7
	ChestX [67]	18090	7758	7
	ChestXPneumonia [25]	5216	624	2
Illustrative	Kaokore [60]	6568	821	8
	Sketch [66]	35000	15889	1000
	PACS-C [32]	2107	237	7
	PACS-S [32]	3531	398	7
Texture	DTD [6]	3760	1880	47
	FMD [81]	1400	600	10

Task2Sim Downstream Performance



Task2Sim matches ImageNet pre-training and generalizes to tasks not encountered in training

*Indicates subset of Imagenet with same number of total images and total object classes as in synthetic data

How Transferable are Video Representations Based on Synthetic Data?

NeurIPS 2022 Dataset Track

Yo-whan Kim, Samarth Mishra, SouYoung Jin, Rameswar Panda, Hilde Kuehne, Leonid Karlinsky, Venkatesh Saligrama, Kate Saenko, Aude Oliva, Rogerio Feris

Synthetic Pretraining



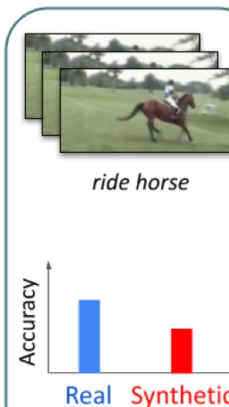
Downstream Tasks

Decreasing scene-object bias

UCF101



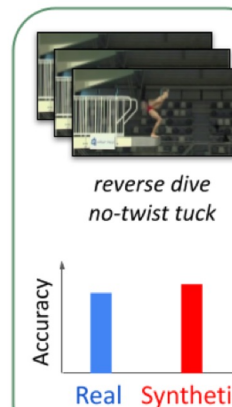
HMDB51



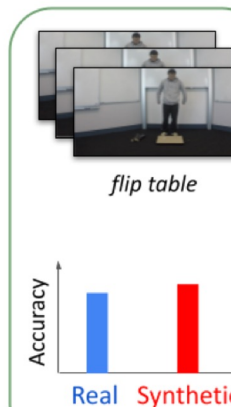
Mini-SSV2



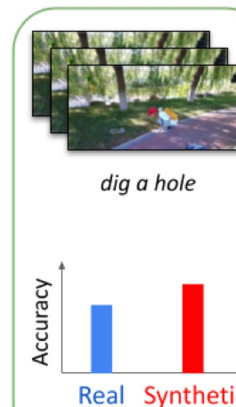
Diving48



IkeaFA



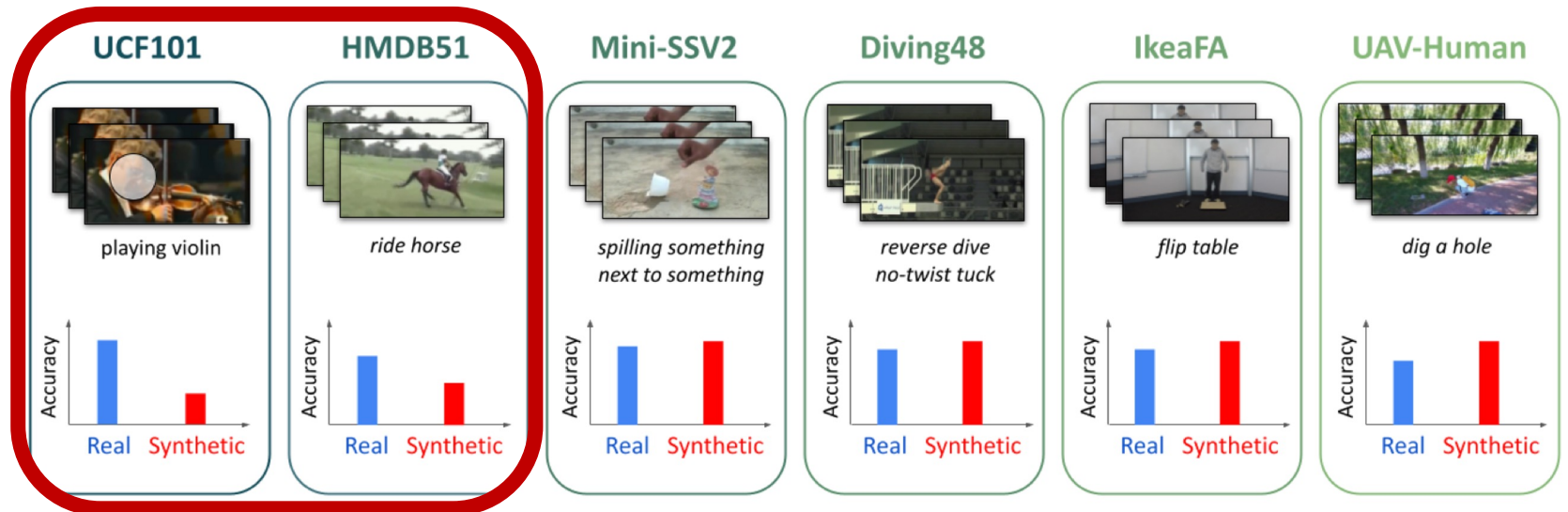
UAV-Human



Downstream Tasks

Decreasing scene-object bias

Real pre-training
outperforms synthetic
pre-training on datasets
with high bias.

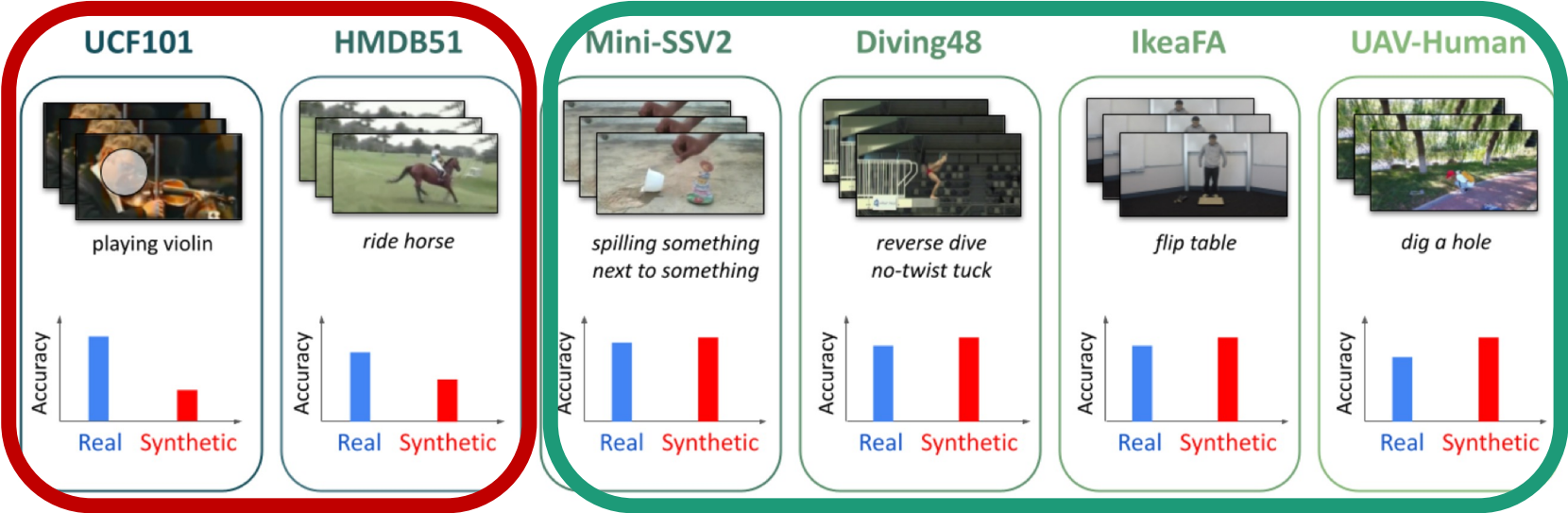


* Sim2real gap is largely due to contextual features (instead of temporal dynamics of the actions)

Synthetic pre-training achieves similar or better accuracy than real pre-training on datasets with low bias

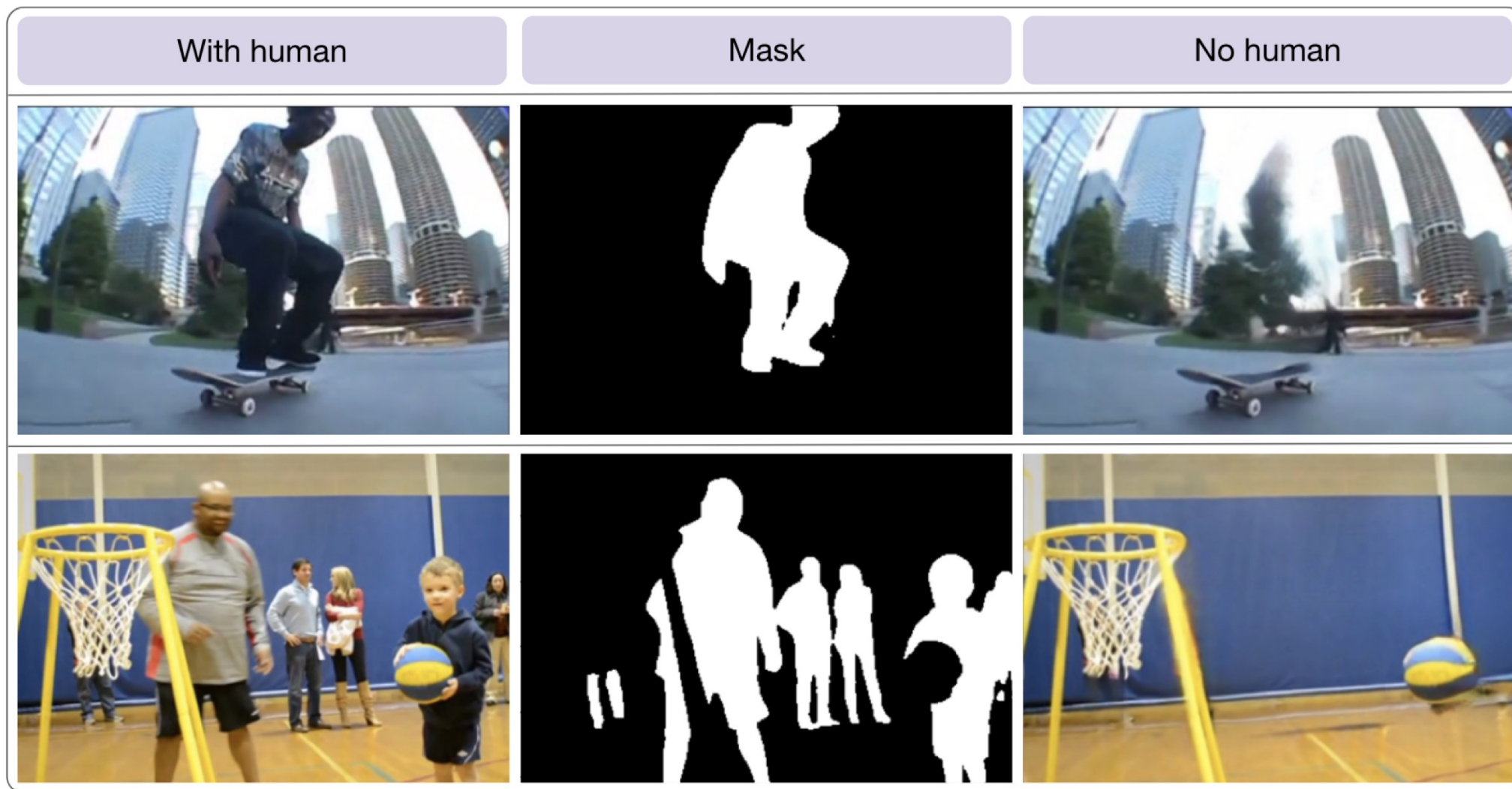


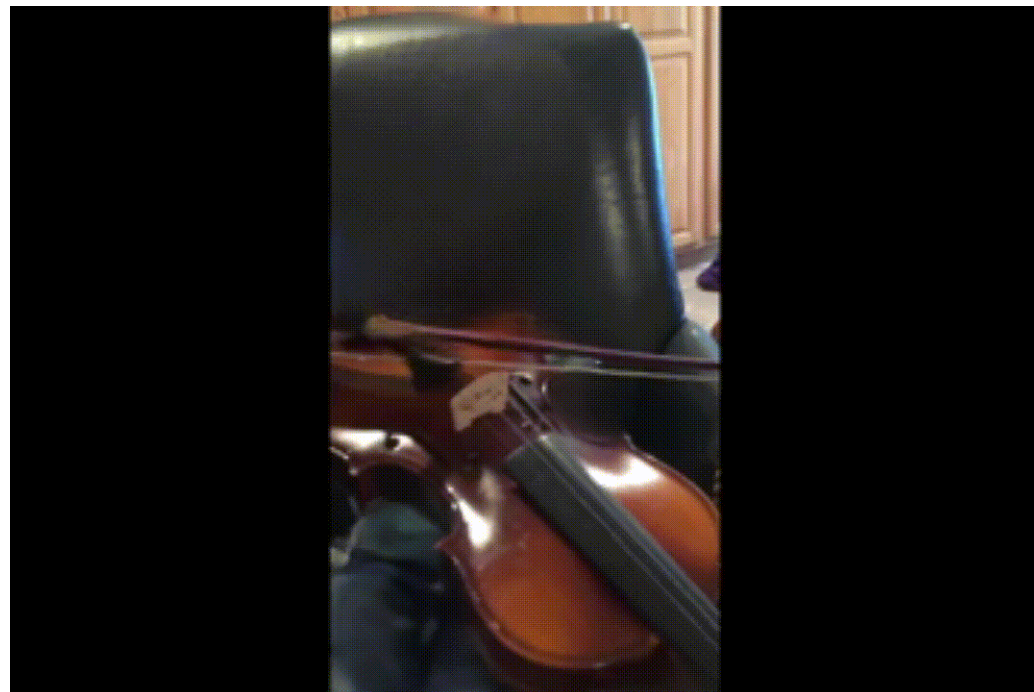
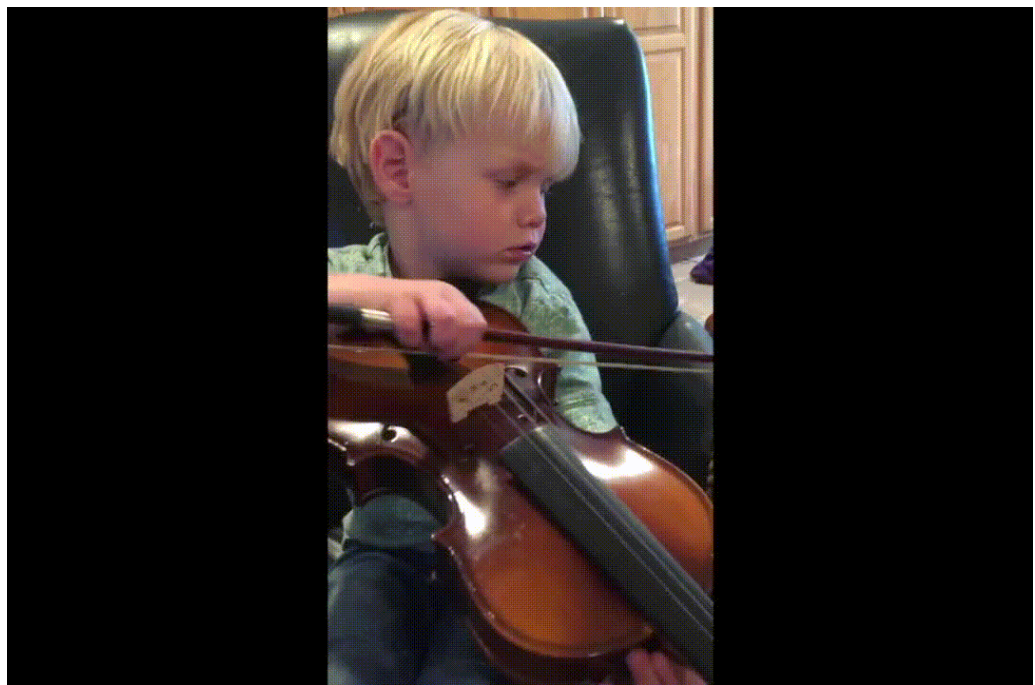
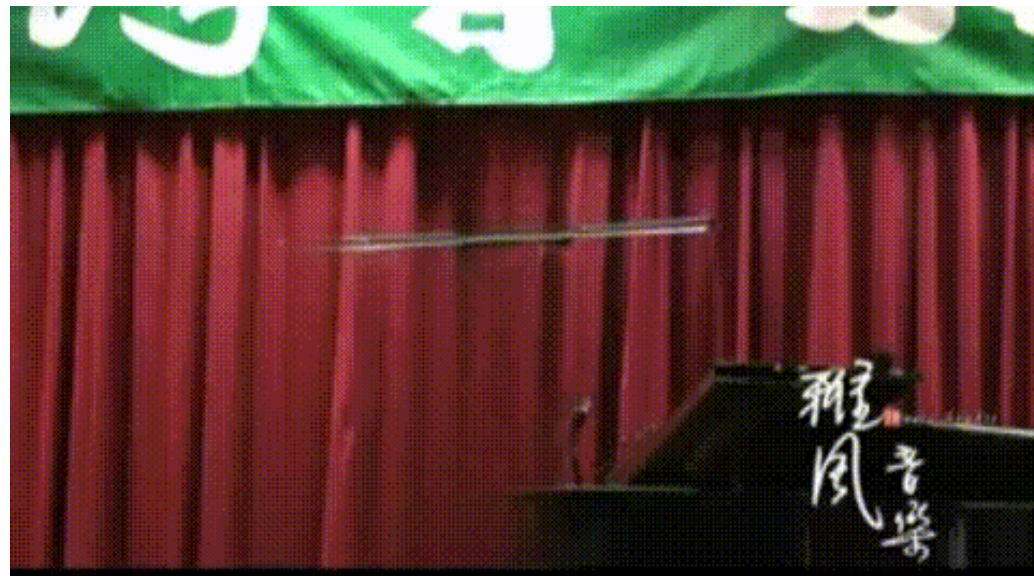
Real pre-training outperforms synthetic pre-training on datasets with high bias



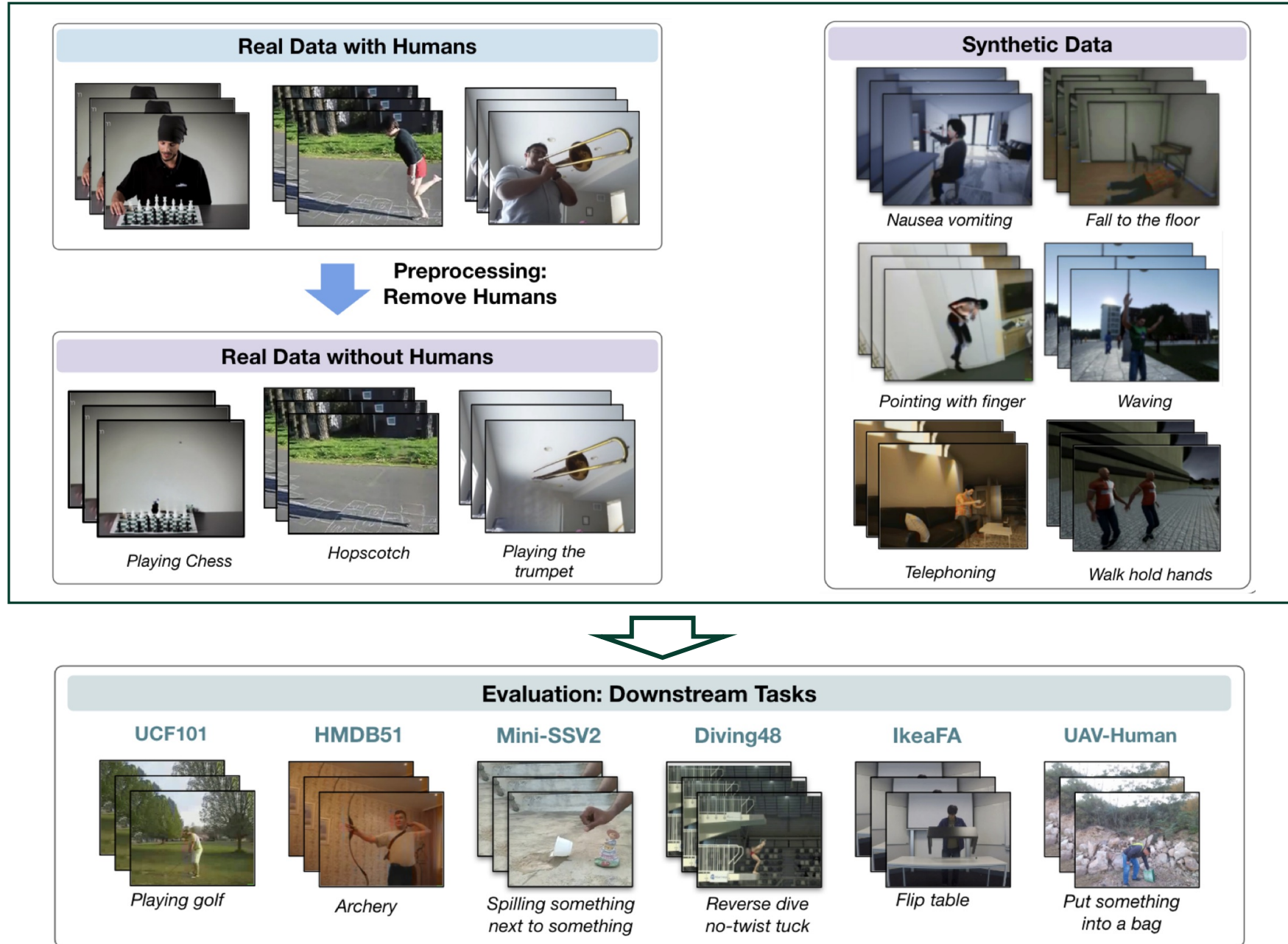
How can we better model contextual features
while preserving privacy?

Learning Human Action Recognition Representations Without Humans [Howard Zhong et al, 2023]



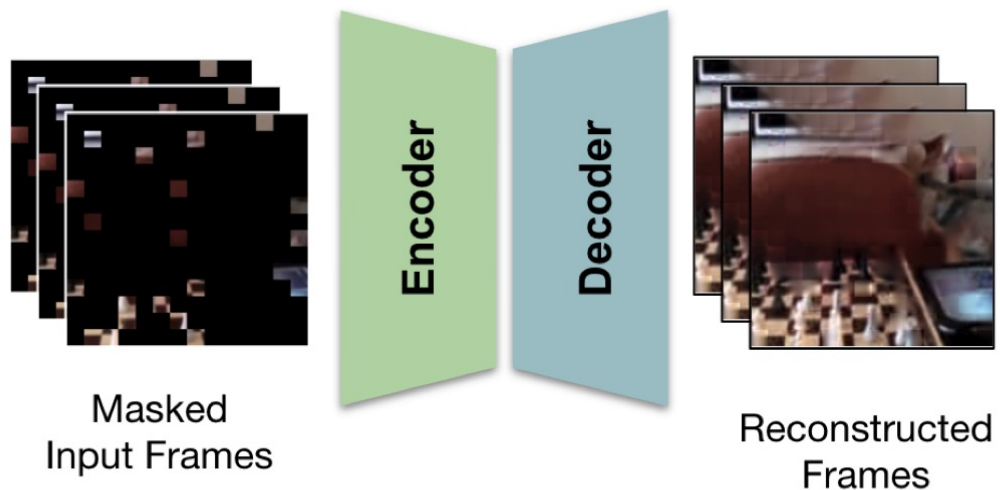


Pre-training: Real Data (no Humans) + Synthetic Data (Virtual Humans)

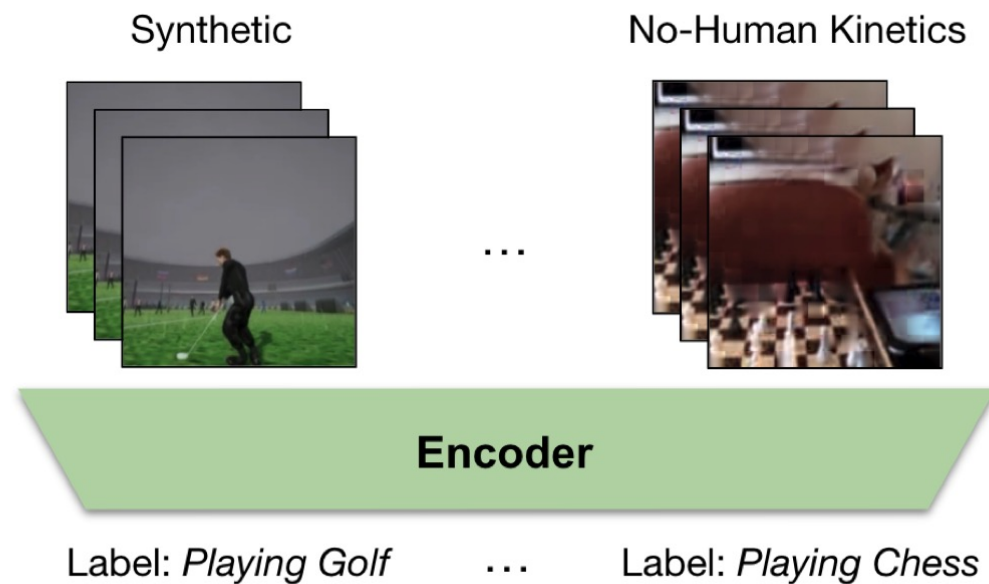


Privacy-Preserving MAE-Align Pre-training

Stage 1: MAE Pre-training on No-Human Kinetics



Stage 2: Supervised Pre-training for Label-Alignment



Experimental Results

Our approach outperforms previous baselines by up to 5% and closes the performance gap between human and no-human action recognition representations

Pre-trained Model	Privacy Preserving	Stage 1: MAE	Stage 2: Alignment	Downstream Task Accuracy												Average	
				UCF101		HMDB51		Mini-SSV2		Diving48		IkeaFA		UAV-Human		FT	LP
				FT	LP	FT	LP	FT	LP	FT	LP	FT	LP	FT	LP		
MAE-Align w. real humans	✗	Kinetics	Kinetics	93.3	91.4	73.4	69.5	68.8	37.4	66.3	19.9	72.0	58.3	34.9	13.9	68.1	48.4
TimeSformer Kinetics ^{*†}	✗	x	Kinetics	92.1	89.4	59.5	55.4	48.9	21.5	46.4	17.0	61.9	47.7	23.3	8.4	55.3	39.9
TimeSformer Synthetic ^{*†}	✗	x	Synthetic	89.0	82.1	54.4	49.2	51.1	21.2	44.9	19.2	63.6	45.5	25.0	13.8	54.7	38.5
Scratch	✓	x	x	30.1	-	14.8	-	16.0	-	9.3	-	19.5	-	0.7	-	15.1	-
TSN (RN50 backbone) [†]	✓	x	Synthetic	83.4	28.0	54.4	20.9	49.7	12.8	63.5	10.9	42.7	36.0	35.6	5.7	54.9	19.1
I3D (RN50 backbone) [†]	✓	x	Synthetic	82.1	27.6	55.7	22.6	50.7	12.3	55.3	10.1	42.7	33.2	35.1	5.8	53.6	18.6
R(2+1)D (RN50 backbone) [†]	✓	x	Synthetic	80.0	26.4	53.3	22.2	52.0	13.3	57.3	10.0	41.5	35.7	31.8	5.5	52.6	18.9
MAE-Align w. Synthetic	✓	Synthetic	Synthetic	88.7	76.6	69.7	59.7	64.3	26.0	61.1	16.7	67.3	57.7	36.1	20.6	64.5	42.9
Ours : Privacy-Preserving MAE-Align (PPMA)	✓	NH Kinetics	NH Kinetics + Synthetic	92.5	88.4	71.2	64.9	67.8	34.9	64.0	21.9	67.9	57.7	38.5	19.3	67.0	47.9

Outline:

- Pre-training based on Synthetic Data for Vision Tasks
- Synthetic Data for Specialization of Large Language Models
- Application: AI Commentary @ Wimbledon and US Open 2023

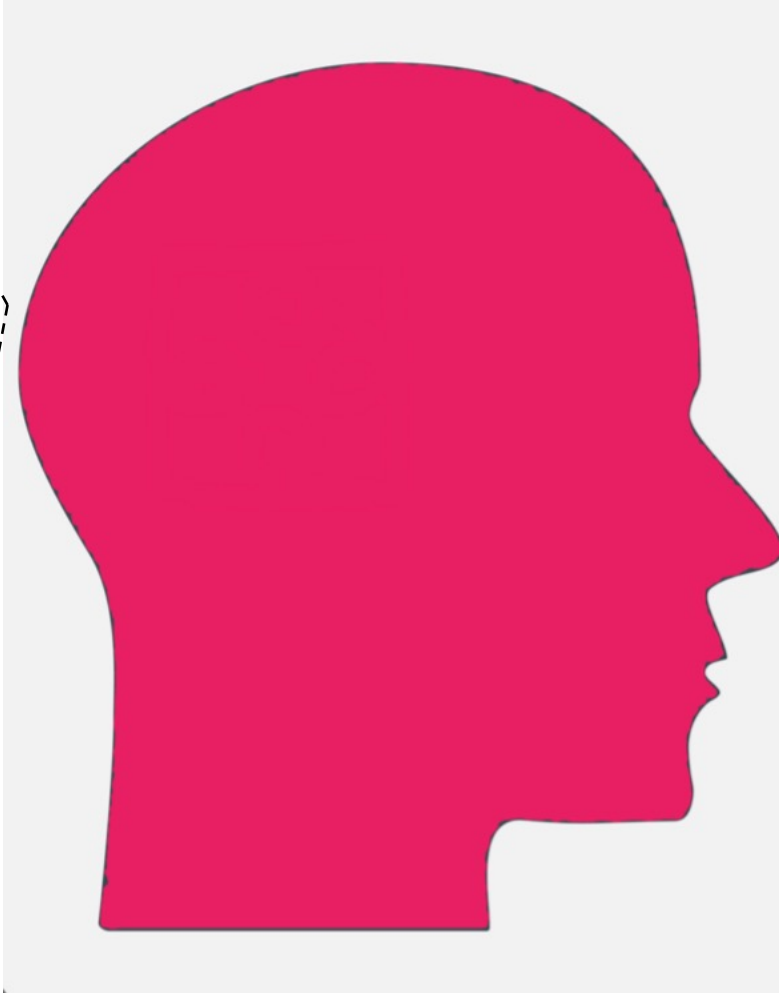
Generalist



- Does well in many domains
- Requires more capacity (bigger models)

Specialist

- Best in a specific domain
- Requires less capacity (smaller models)



Generalist

- Does well in many domains
- Requires more capacity (bigger models)



Can we adapt a general large language model to perform better in a specific domain?

- Naive Approach: Instruction fine-tuning, but it requires time-consuming and difficult to scale manual annotation

SELF-SPECIALIZATION: UNCOVERING LATENT EXPERTISE WITHIN LARGE LANGUAGE MODELS

Junmo Kang^{*1}

Hongyin Luo²

Yada Zhu³

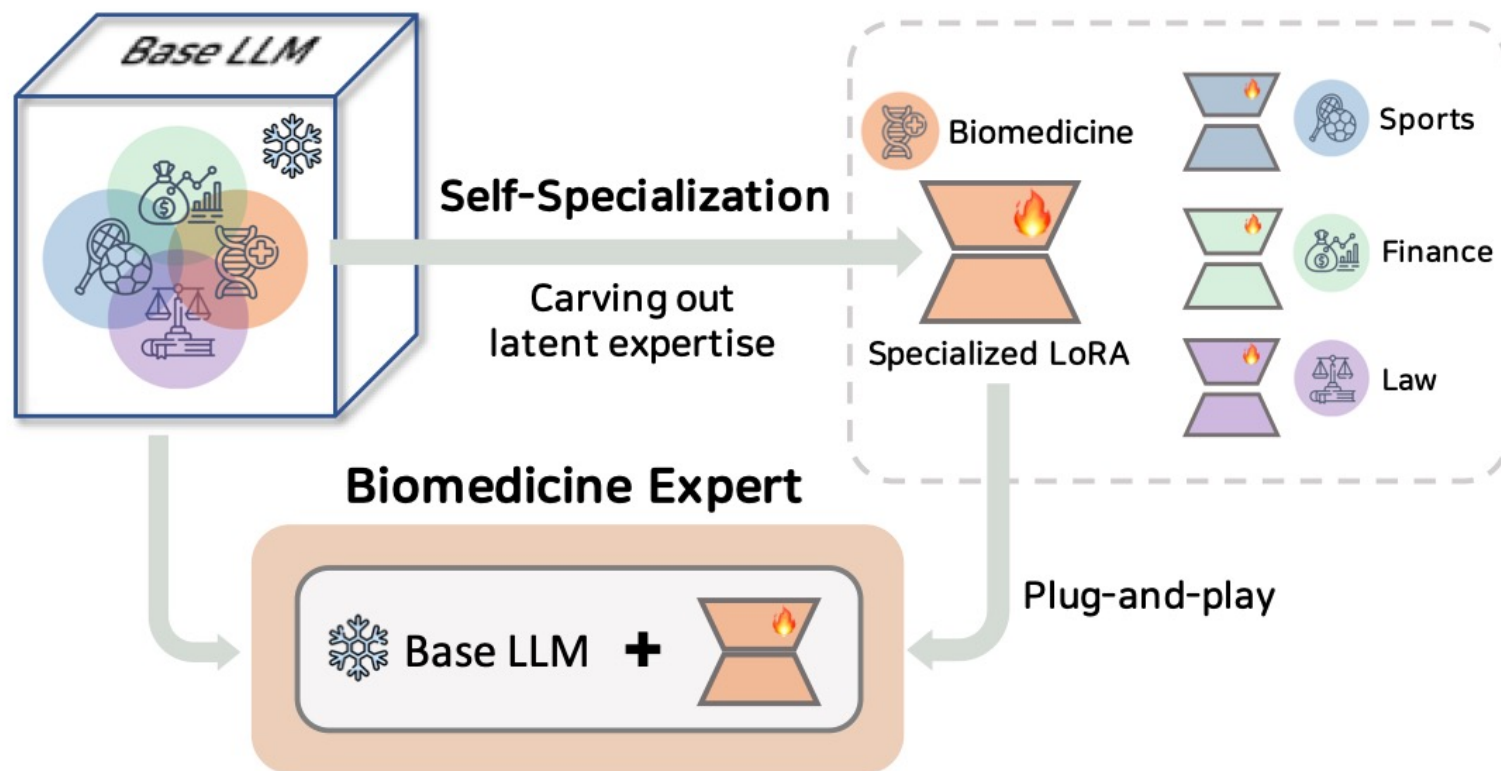
James Glass²

David Cox³

Alan Ritter¹

Rogério Feris³

Leonid Karlinsky³



Real-Data:

Domain-specific
seed instructions
(small set – e.g., 80)



Instruction:

In this task, you are given a short article and question.
Read the short article and answer the question.

Input:

Short article: The extract from the opium poppy was tested on breast cancer cells and was found to inhibit the migration and invasion of breast cancer cells.

Question: Was the extract more potent in its inhibitory effect on the migration of breast cancer cells than its effect on the invasion of breast cancer cells?

Real-Data:

Domain-specific
seed instructions
(small set – e.g., 80)



Base LLM



Synthetic Data:

Self-generation of
domain-specific
instructions



Instruction: Generate a list of drugs
which can be used for the treatment of
the given symptom.

Instruction: Given medication
records, predict possible drug-drug
interactions.

Instruction: You are given data of
genetic variations and mutations,
generate a comprehensive report.

⋮

Instruction: Provide an answer to the
following question about the patient's
medical history.

Real-Data:

Domain-specific
seed instructions
(small set – e.g., 80)



Base LLM



Synthetic Data:

Self-generation of
domain-specific
instructions

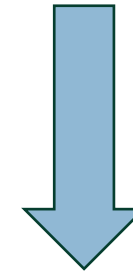
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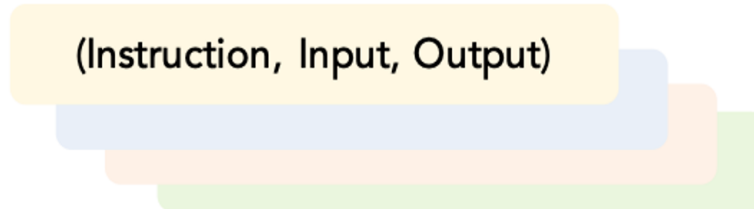
Base LLM



Domain-specific
knowledge
(unlabeled)



(Instruction, Input, Output)



Real-Data:

Domain-specific
seed instructions
(small set – e.g., 80)



Synthetic Data:

Self-generation of
domain-specific
instructions

Base LLM



Instruction: Generate a list of drugs
which can be used for the treatment of
the given symptom.

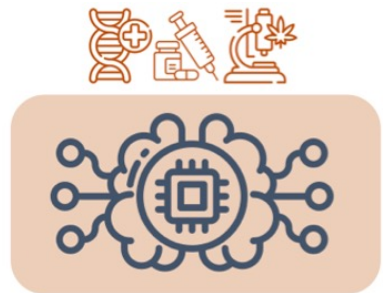
Instruction: Given medication
records, predict possible drug-drug
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Instruction: You are given data of
genetic variations and mutations,
generate a comprehensive report.

⋮

Instruction: Provide an answer to the
following question about the patient's
medical history.

Specialized LLM (e.g., Biomedicine)



Parameter-Efficient
Fine Tuning (LoRA)

(Instruction, Input, Output)

Domain-specific
knowledge
(unlabeled)

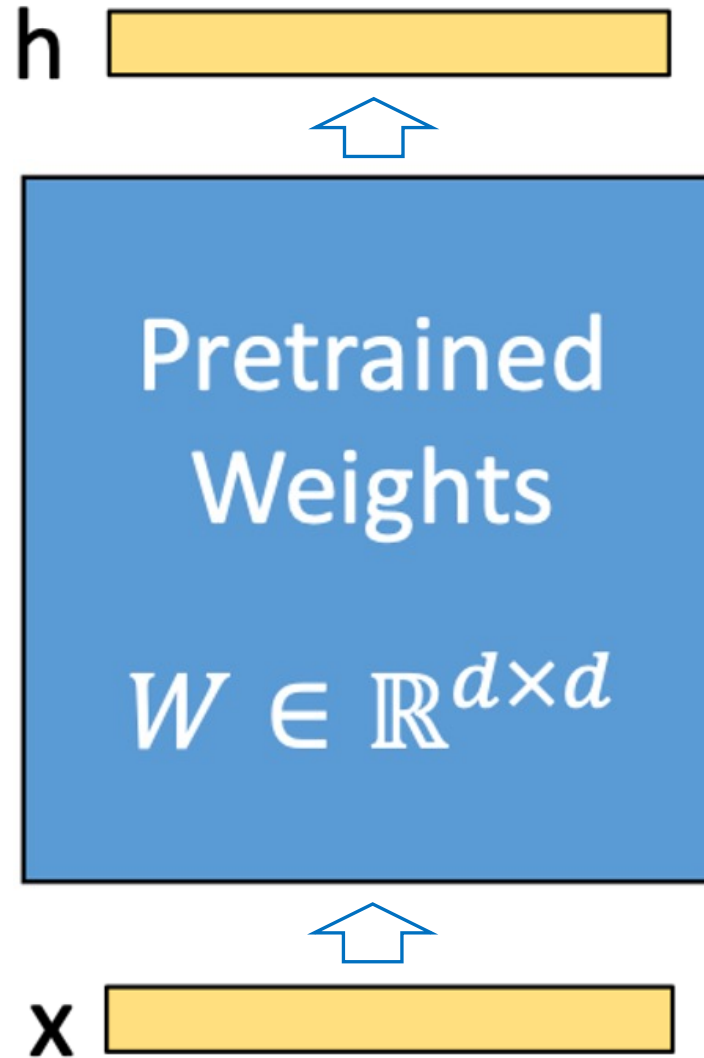


Base LLM



Traditional Fine-tuning

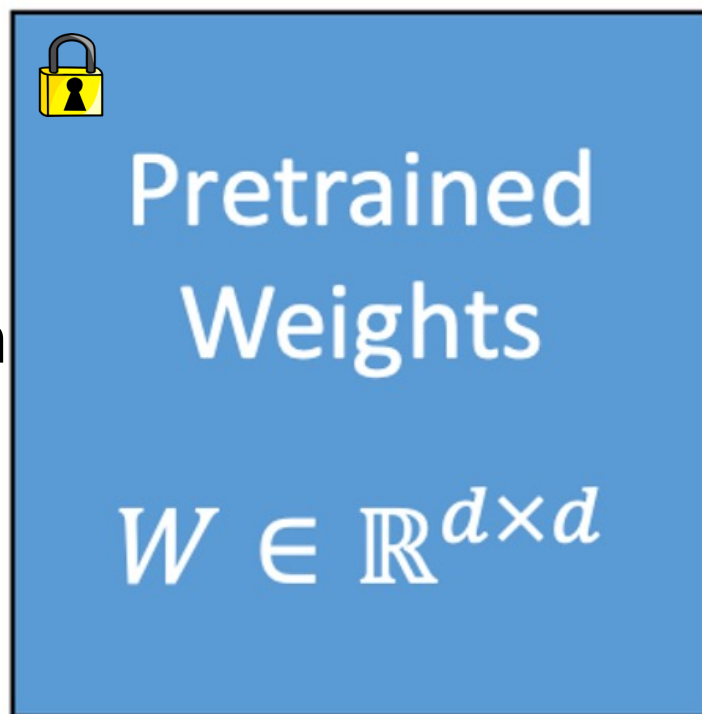
- All pre-trained model weights are optimized
- Not feasible for billion-scale models



LoRA: Low-rank adaptation of large language models

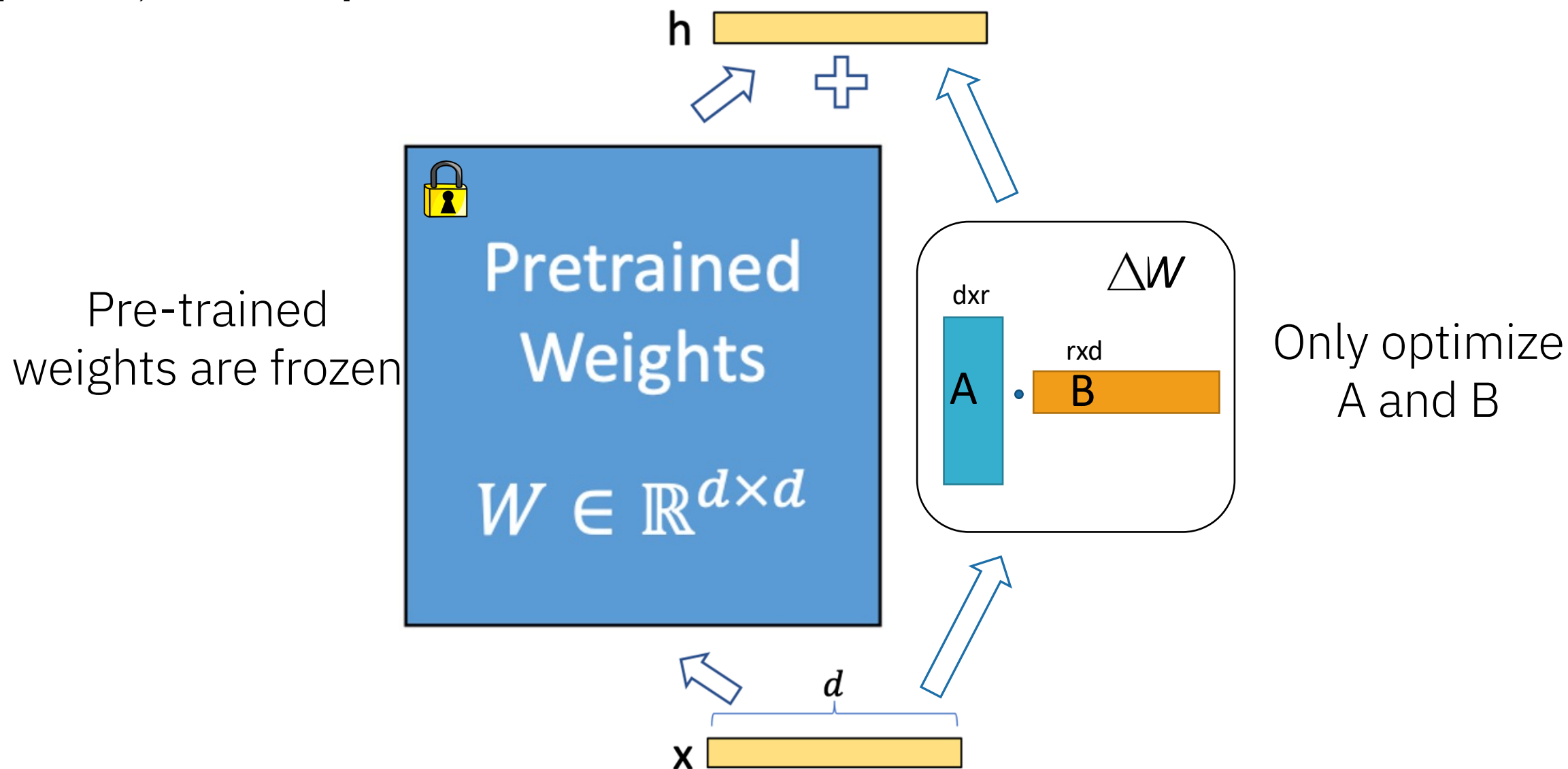
[Hu et al, ICLR 2022]

Pre-trained
weights are frozen



LoRA: Low-rank adaptation of large language models

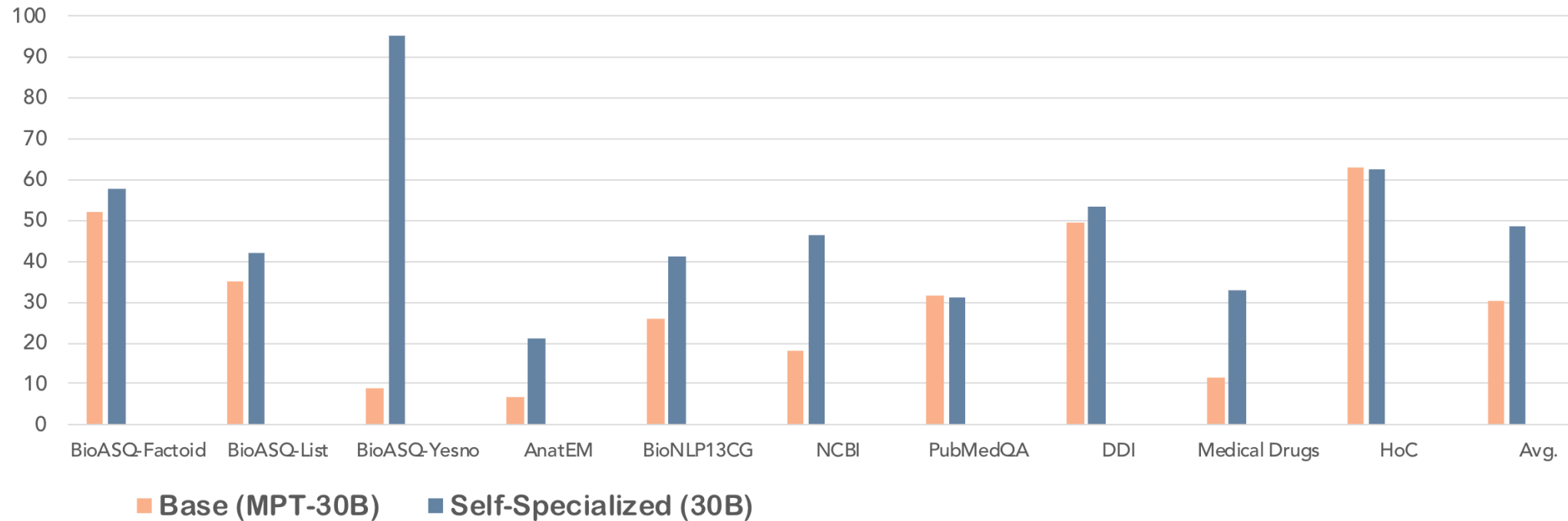
[Hu et al, ICLR 2022]



Main Results

- Our self-specialized model (30B) outperforms its base model, MPT-30B by a large margin (up to 18 points)

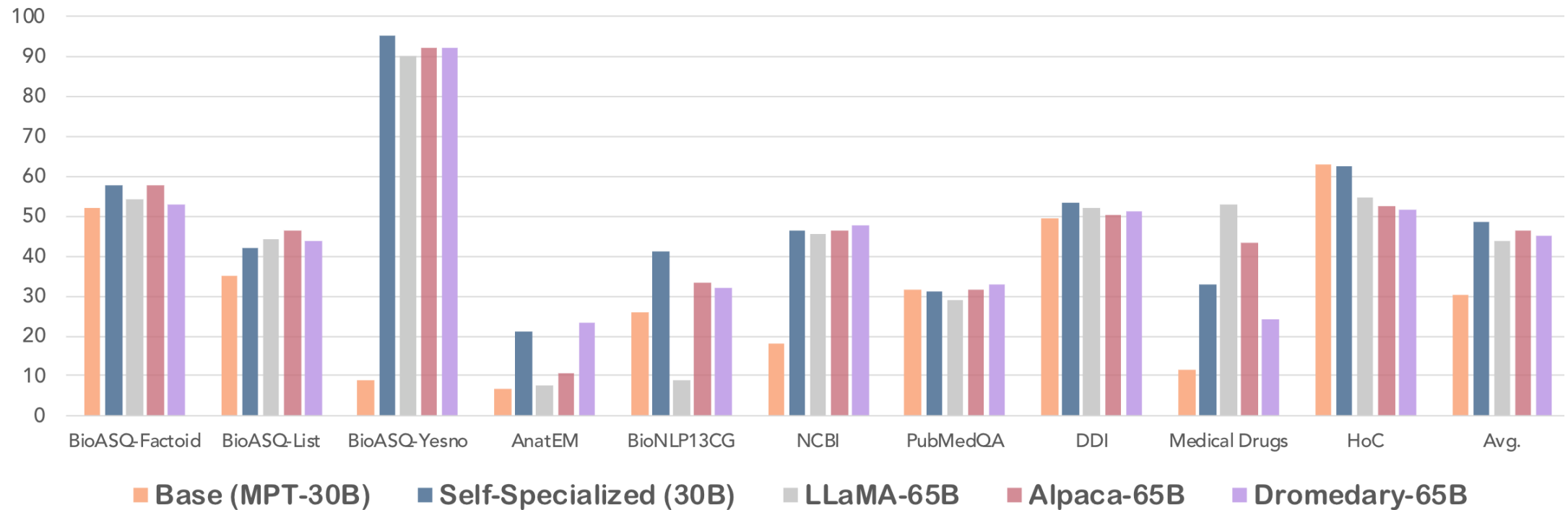
5-shot results in 10 datasets (biomedicine)



Main Results

- Our self-specialized model (30B) even outperforms 65B models (2.2x larger), including Llama-65B

5-shot results in 10 datasets (biomedicine)



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90 players

18 holes

4 days

100s
hours of footage



since **1877**

>600 players

19 courts

13 days

254 matches

100s hours of footage



14 days

400 players

304 hours of coverage

24K clips processed



2016 - FINAL RD
LOUIS OOSTHUIZEN
16TH HOLE



OVERALL EXCITEMENT LEVEL

COMMENTATOR EXCITEMENT

ACTION RECOGNITION

CROWD CHEERING

CURRENT TIME: 1:34 PM CLIP TIME: 2:15 PM APRIL 9, 2017

HOLES 15 & 16: LOUIS OOSTHUIZEN m HOLE

COMMENTARY:



2:15 PM APRIL 9, 2017
HOLES 15 & 16
0.78
EXCITEMENT LEVEL



2:11 PM APRIL 9, 2017
BROADCAST
0.18
EXCITEMENT LEVEL



2:11 PM APRIL 9, 2017
BROADCAST
0.24
EXCITEMENT LEVEL



2:11 PM APRIL 9, 2017
BROADCAST
0.51
EXCITEMENT LEVEL

Wimbledon AI Highlights

With
Watson™

IBM®

All Events All Rounds All Statistics All Days



Jack Draper vs Nicolas Mejia

Set 3 : 40-AD : Match Point; Draper wins the point with a forehand smash winner.

1.00

Crowd Cheering

1.00

Match Analysis

0.90

Player Gestures

0.98

Overall Excitement

Type Player Name



COVERAGE

45144 653

HIGHLIGHTS

261 38

TOP 5 MAIN EVENTS HIGHLIGHTS

38 6619 5981

Set Excitement Threshold

Most Excitement | Most Recent

0.00 0.3 1.00



0.98

Friday, July 13, 2018, 2:32 PM

Jack Draper vs Nicolas Mejia

Set 3 : 40-AD : Match Point; Draper wins the point with a forehand smash winner.



0.92

Monday, July 9, 2018, 1:11 PM

Jiri Vesely vs Rafael Nadal

Set 3 : 0-0 : Vesely wins the point with a forehand volley winner.



0.92

Tuesday, July 10, 2018, 9:34 AM

Daria Kasatkina vs Angelique Kerber

Set 2 : 40-40 : Kerber wins the point with a forehand volley winner.



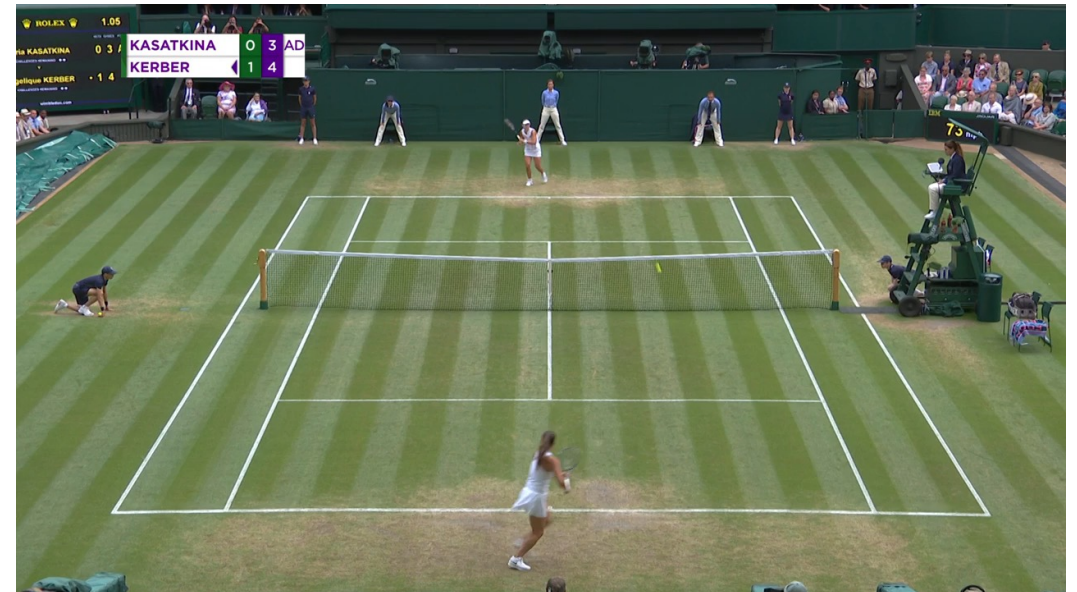
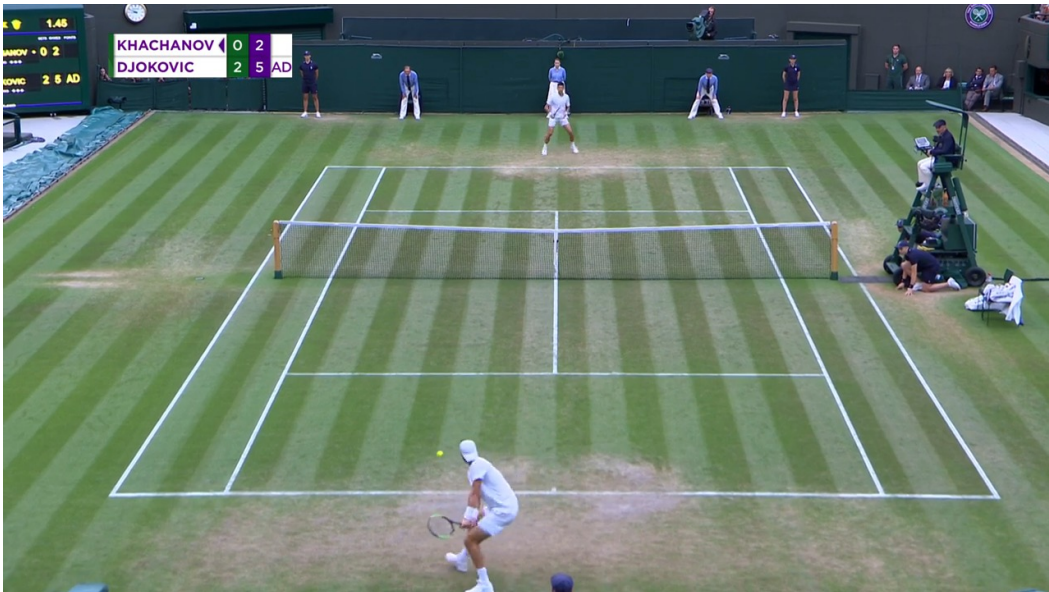
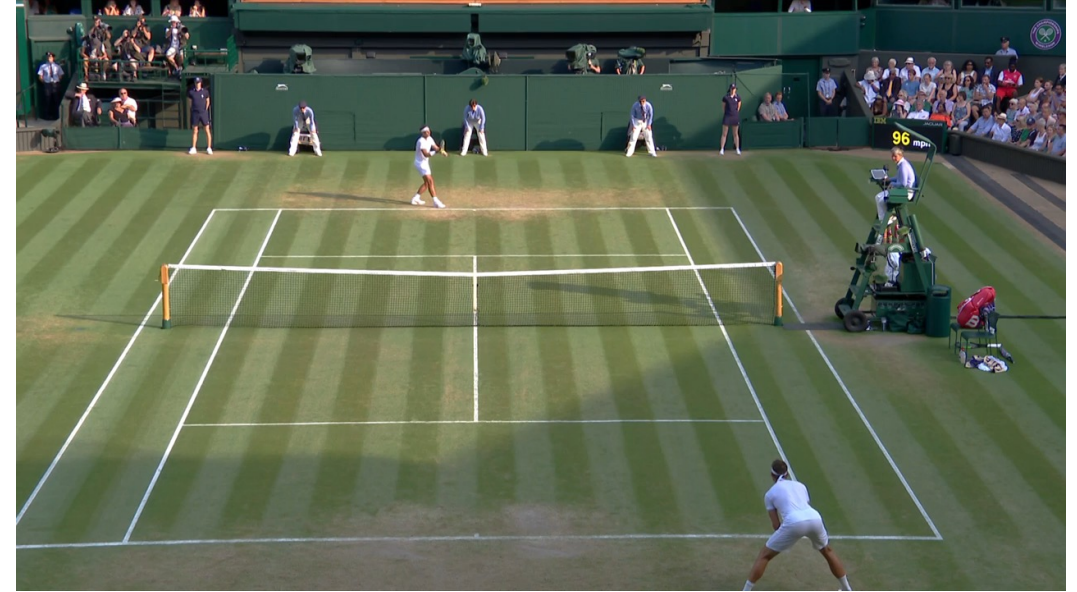
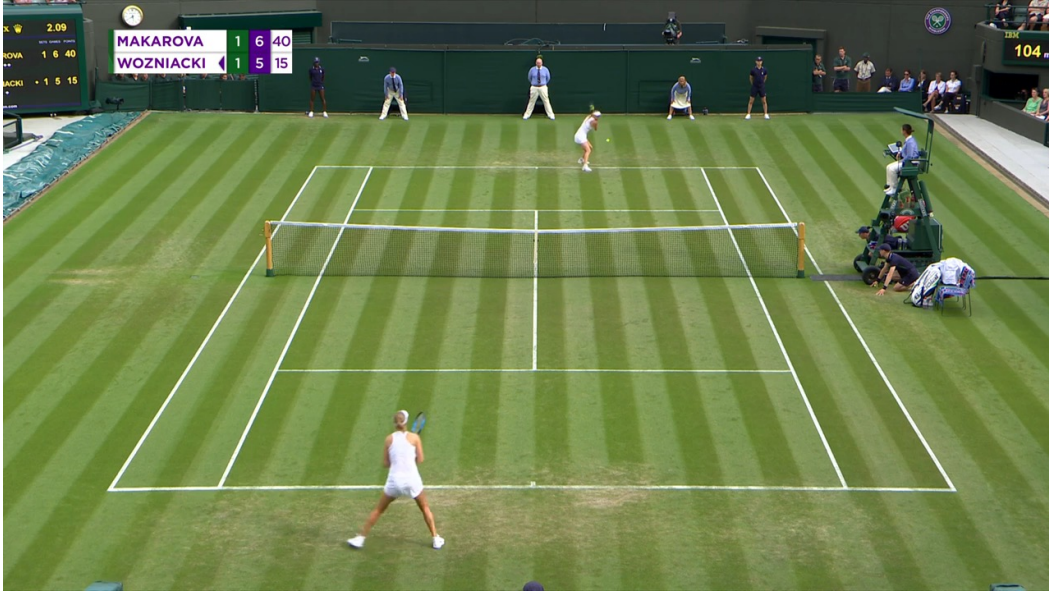
0.91

Saturday, July 14, 2018, 10:09 AM

Novak Djokovic vs Rafael Nadal

Set 5 : 40-AD : Djokovic wins the point with a forehand winner.

Multi-modal curation of sports highlights (Wimbledon)



Impact

- Our system has been used to produce the official highlights of the US Open, Wimbledon, and Masters tournaments
- Watched by millions of fans worldwide (mobile app, website, ...)

The New York Times

***Enjoy Those U.S. Open Highlights. A
Computer Picked Them for You.***

(and many more media outlets)

Our work on AI/ML for auto-curation of sports highlights has been selected for a Technology and Engineering **Emmy Award** !

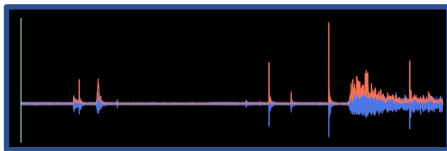
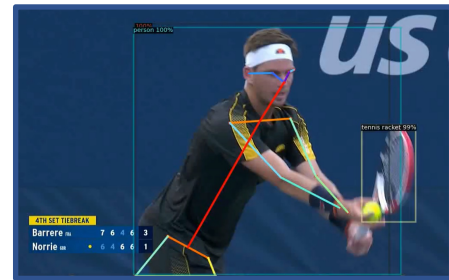


AI Commentary

Wimbledon and US Open 2023

Turning multi-sensory data into fluent commentary using Large Language Models

After a wide out serve by Peniston, Laaksonen returns the service with a short backhand. Peniston then approaches the net, hits a forehand, but Laaksonen strikes back with a magnificent backhand. What a beautiful shot! Peniston makes all efforts to save the point but hits the net with a volley unforced error. The game is now tied: 30-30



Video
Sound
Match Data
Excitement Scores

Projection into a
unified
embedding
space
(v1: Data to Text)

Billion-scale IBM Language Model

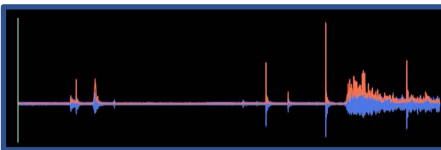
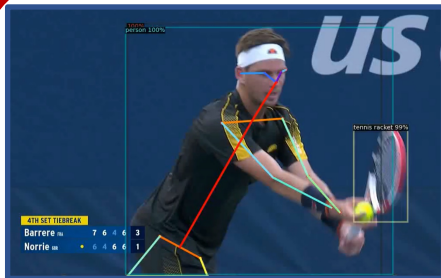
- IBM Sandstone.3B model
- Pretrained on petabytes of data
- Fine-tuned on real and synthetic commentary data

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Video
Sound
Match Data
Excitement Scores

Projection into a unified embedding space
(v1: Data to Text)



Play-by-play metadata extraction using computer vision

Understanding every detail of the game in real-time



Court and Net
Detection



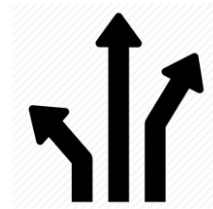
Player Tracking
and pose/keypoint estimation



Ball Tracking



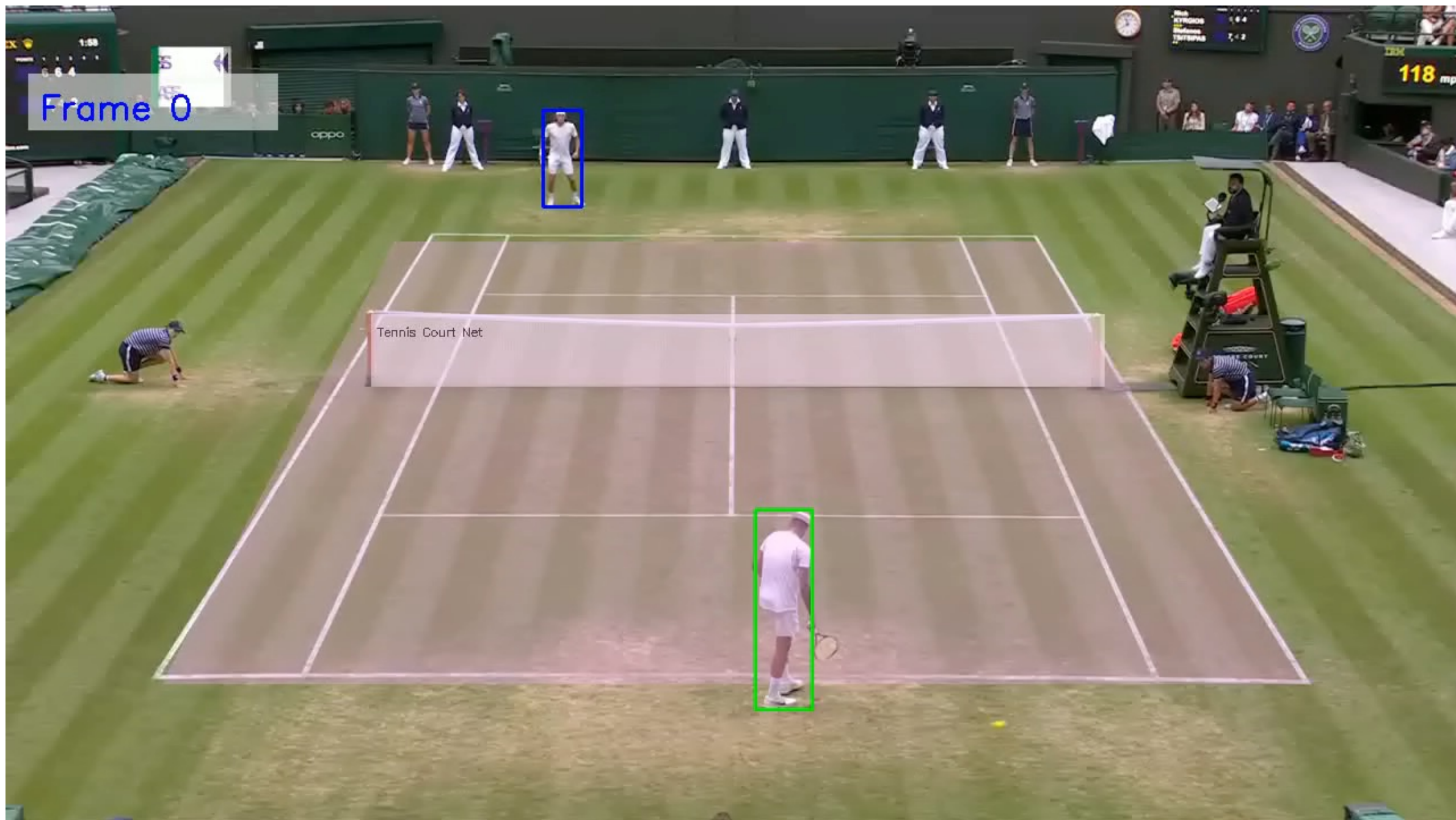
Action
Recognition
Serve, Forehand, Volley, ...



Shot Directionality

Frame 0

Tennis Court Net



Metadata from other sensors and modalities



Score information



Ball speed (Radar)



Audio and vision excitement scores



Player information



Rally Length

DJOKOVIC	6-1 6-7 6-3	ALMAGRO
MATCH TIME 2:02		
12	ACES	10
0	DOUBLE FAULTS	2
71	1st SERVE IN %	57
78	1st SERVE PTS WON %	70
58	2nd SERVE PTS WON %	53
40	WINNERS	29
21	UNFORCED ERRORS	32
6/7	NET POINTS WON	6/11
3/3	BREAK POINTS WON	0/1

Match analysis



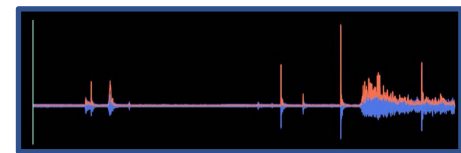
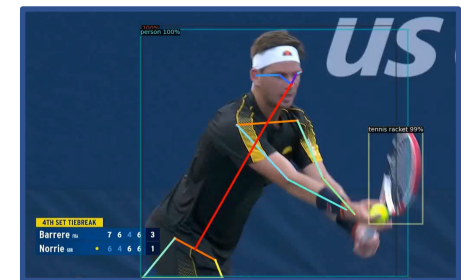
Winner vs Error

Metadata: Computer vision + other modalities (input to large language model)

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Turning multi-sensory data into fluent commentary using Large Language Models

After a wide out serve by Peniston, Laaksonen returns the service with a short backhand. Peniston then approaches the net, hits a forehand, but Laaksonen strikes back with a magnificent backhand. What a beautiful shot! Peniston makes all efforts to save the point but hits the net with a volley unforced error. The game is now tied: 30-30

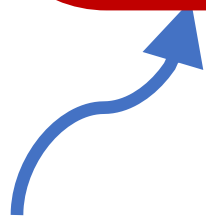


Video
Sound
Match Data
Excitement Scores

Projection into a unified embedding space
(v1: Data to Text)

Billion-scale IBM Language Model

- IBM Sandstone.3B model
- Pretrained on petabytes of data
- Fine-tuned on real and synthetic commentary data

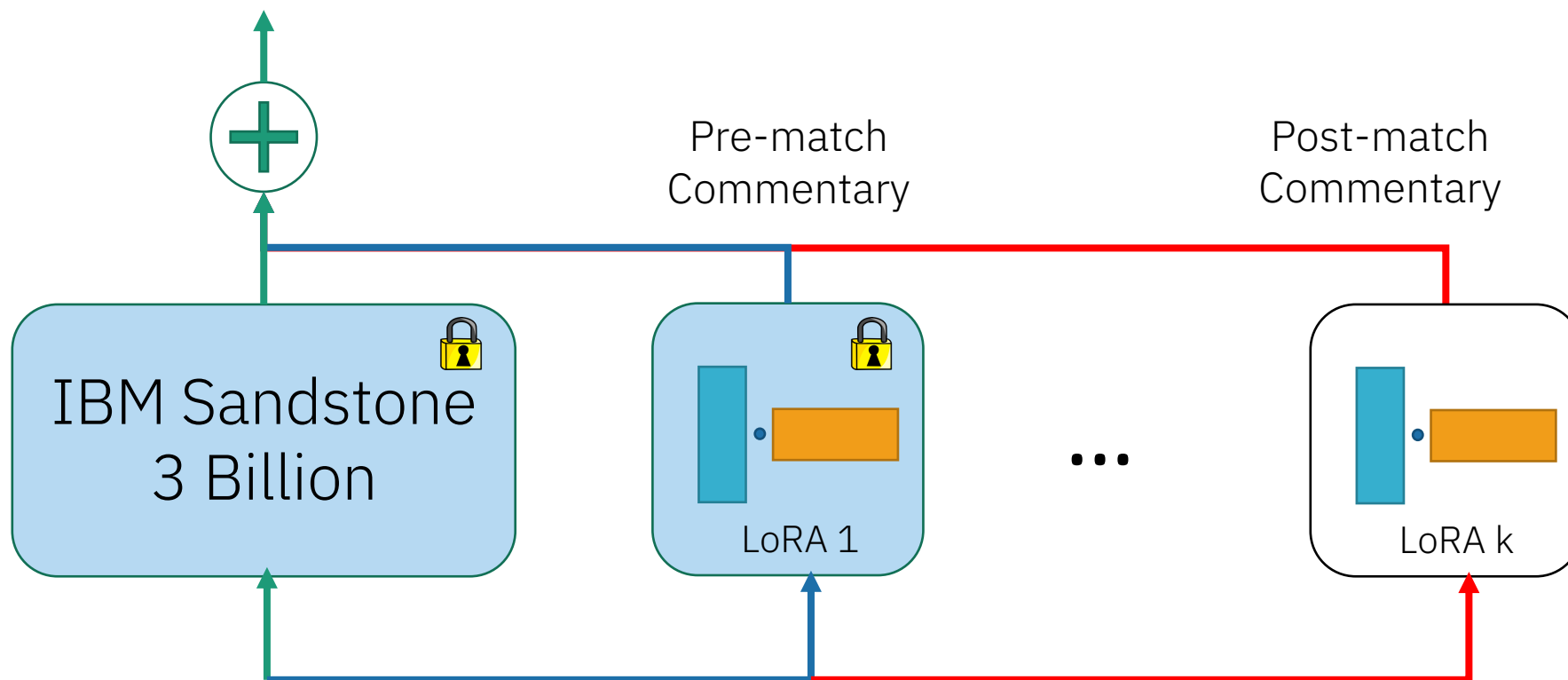


IBM Sandstone 3B model

- Encoder-Decoder Architecture
 - Base model pre-trained on petabytes of data
 - Long context modeling with ALiBi
-
- Tuned for AI commentary using Synthetic and Real Data
 - Efficient tuning is achieved using Layered LoRA (next)

Our Layered LoRA Architecture (LaLo)

[James Smith et al, CVPR 2023]



LaLo can support task modularity & efficient recall of past models

Efficient adaptation: < 3% of all parameters are adapted !

Wimbledon Multimodal Perception and Commentary

IBM

All Events

All Rounds

All Statistics

Day 3 Wed 3 July

Type Player Name

COVERAGE

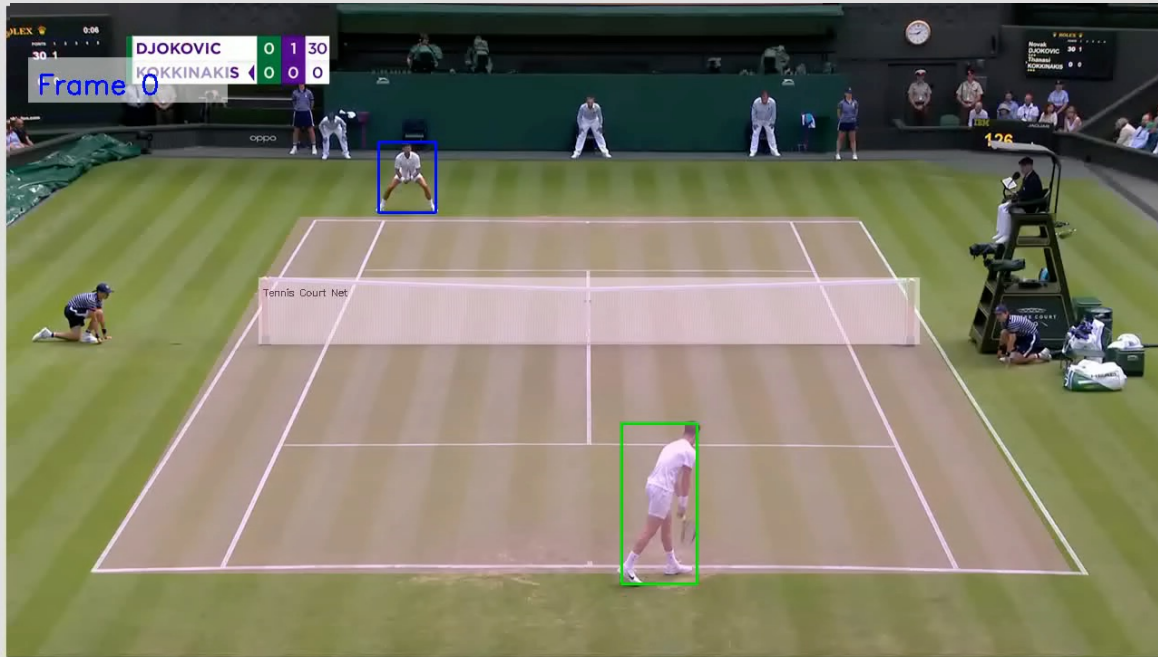
13169 48
Points Hours

HIGHLIGHTS

14 0
Produced Published

TOP 5 MAIN EVENTS HIGHLIGHTS

0 0 900
Published Secs Published Secs Remaining



Djokovic vs Kokkinakis

Set 1: 1-0; 15-30

Auto-Generated Commentary:

After a Kokkinakis serve up the middle, Djokovic hits a backhand. After a Kokkinakis forehand up the middle, the athletes jump into a sequence of backhand to backhand. Unlucky miss by Djokovic as he directs a backhand unforced error into the net to lose the point.

Most Excitement | Most Recent

Set Excitement Threshold

0.00 0.3 1.00



0.90

Wednesday, 3 July 2019, 13:08

Novak Djokovic vs. Thanasi Kokkinakis

Set 1: 15-30 : Djokovic loses the point with a backhand unforced error.



0.15

Wednesday, 3 July 2019, 13:08

Serena Williams vs. Alize Cornet

Set 3: 15-40 : Break Point; Williams wins the point with a forehand winner.



0.27

Wednesday, 3 July 2019, 13:07

Novak Djokovic vs. Thanasi Kokkinakis

Set 1: 0-30 : Kokkinakis loses the point with a forehand unforced error.



0.86

Wednesday, 3 July 2019, 13:06

Jeremy Chardy vs David Goffin

Set 1: 15-40 : Break Point; Goffin wins the point with a forehand winner.

Conclusions

- Era of Foundation Models: Training with billions of images (vision) and trillions of tokens (language)
- **Take-away message:** focus on data quality (instead of just data quantity) for safer and more efficient models
- Learning with real + synthetic data is a promising way to achieve this goal
- Outlook: Synthetic data for understanding what makes for a good pre-training model

Thank you!

See more at <http://rogerioferis.org>