

Sharif University of Technology

# Life as a Data Scientist

#### **System Analysis and Design**

#### S.M.F. Sani

Data Science and Machine Learning Lab (DML) Department of Computer Engineering, Sharif University of Technology *Oct 2024* 



Data Science and Machine Learning Lab



## Introduction

#### Introduction About Me













## Introduction (Cont.) **About Me**









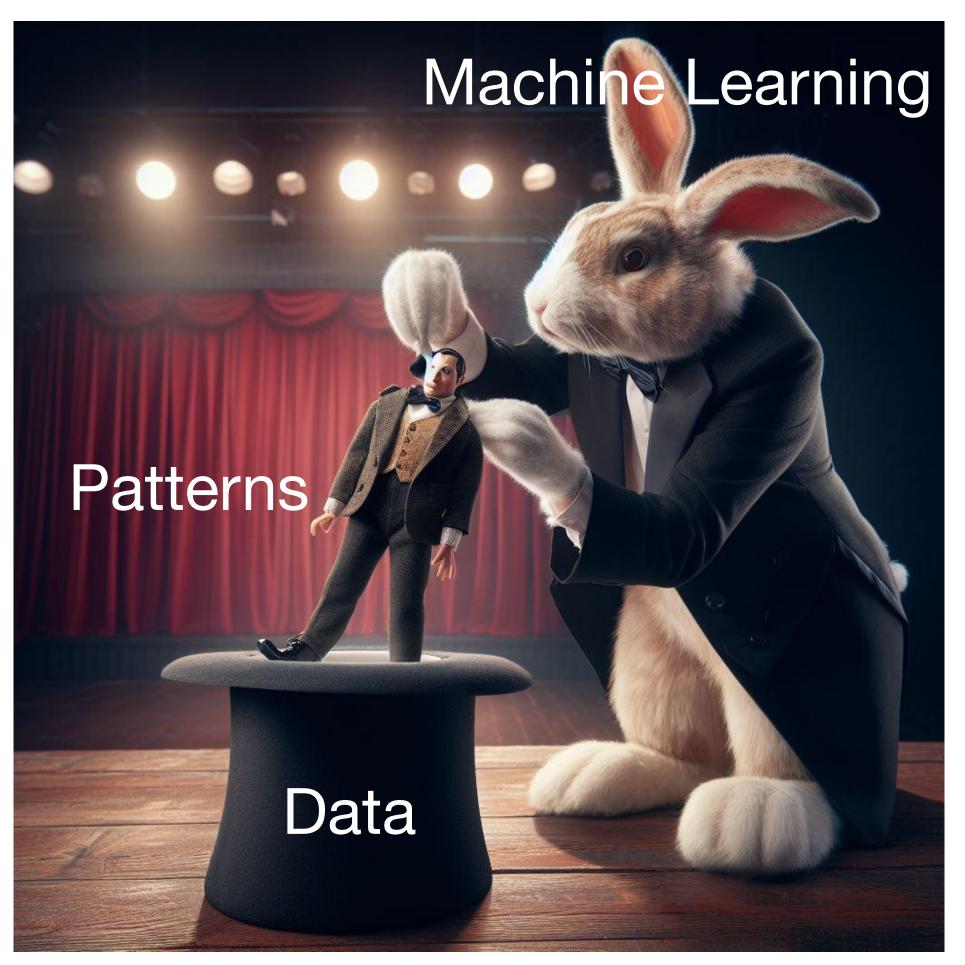
## **NJ2PSell**





## What is this machine learning thing...

#### Machine Learning Magic of our era!



#### Machine Learning Magic of our era!



Cyberpunk Style Mercedes Benz in Night City

super realistic portrait of an Iranian model, little smile, rainy day, standing in crossway

A blonde woman with elvish ears waring green dress that has silver buttons puts flower hairpin in her hairs sitting on a wooden chair inside of wooden cottage.



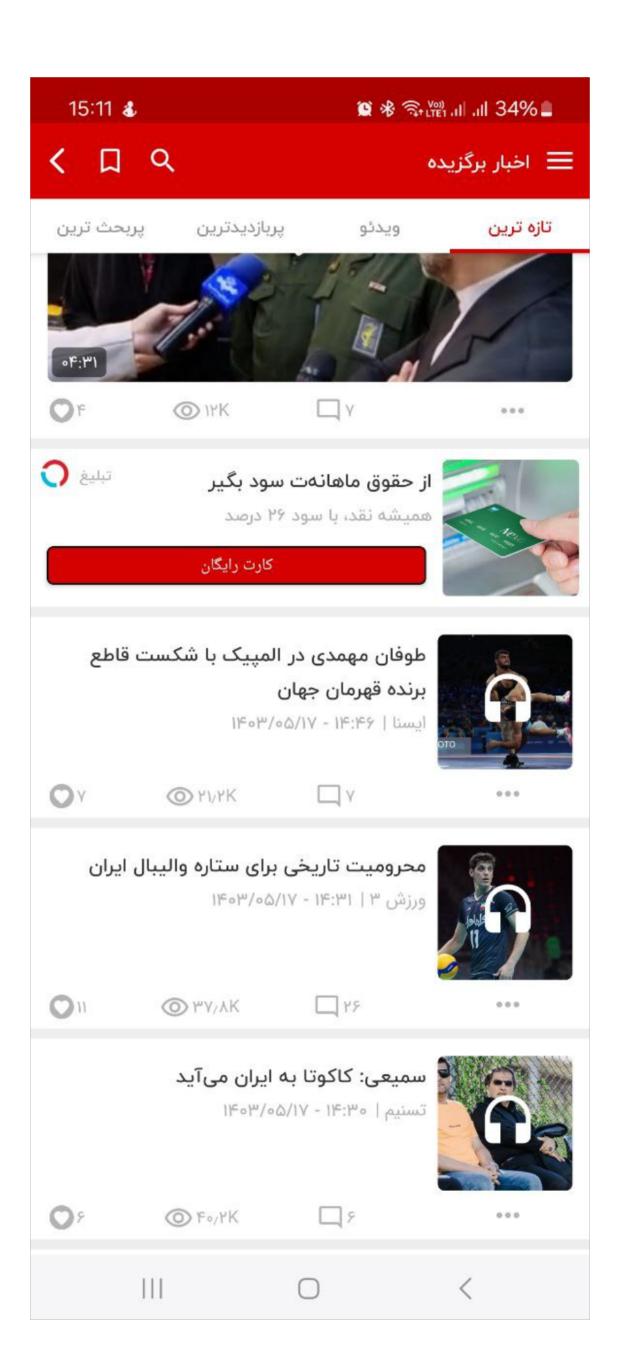
# **Applications (Iran)**Digital Advertising

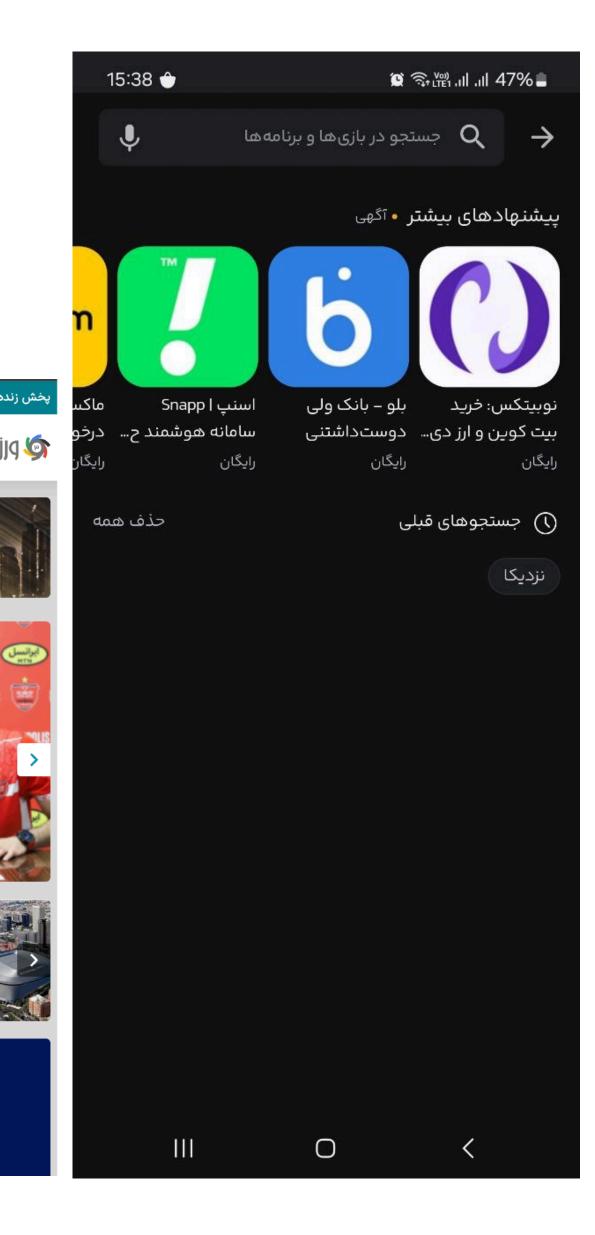
پخش زنده اپلیکیشن ورزش سه پیش بینی گزارش تصویری اخبار ورزشی ثبت نظر پادکست تبلیغات سوژههای شما

🖬 شنبه ۲۱ مهر - 🔍 🛛 I۵:۳۳ مهر - 🔍 ورود

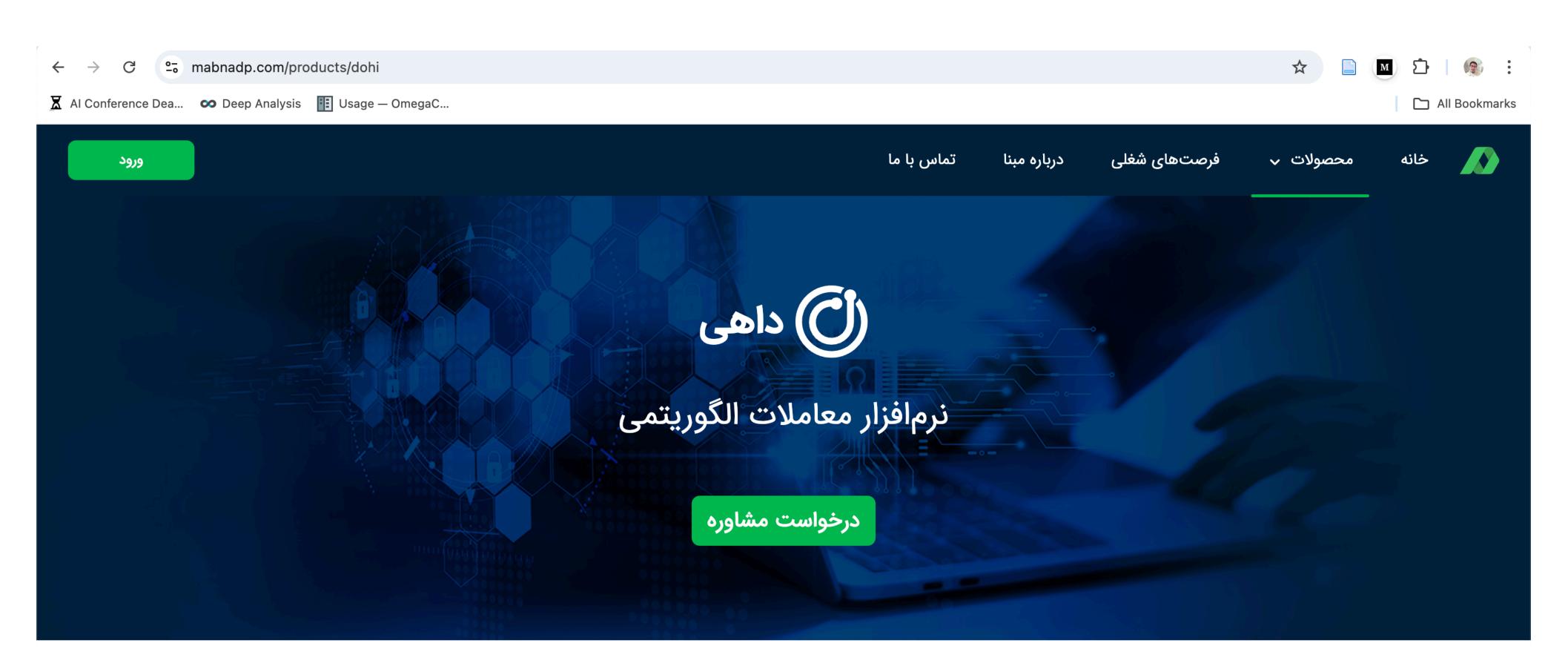
🏠 کارش 🗰 🖉 صفحهاصلی جدول لیگ برتر جام حذفی ویدیو روزنامه نقل و انتقالات نتایج زنده 🔹



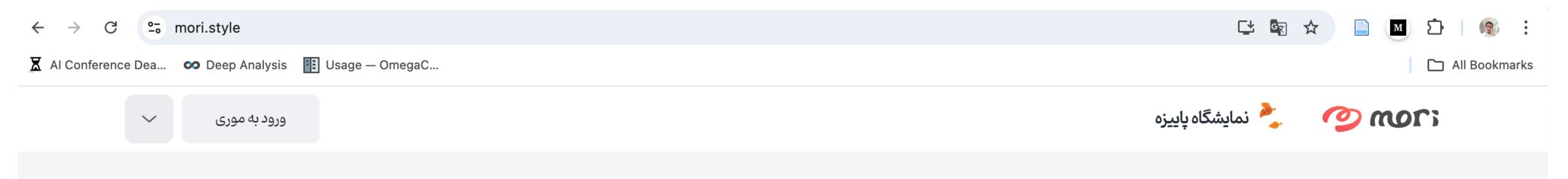




## **Applications (Iran)** Trade



## **Applications (Iran) Product Search**







o/\ooo

 $\leftarrow$ 

#### جستجوبين ١.۴ ميليون لباس از صدها فروشنده

۹ به دنبال چه جور لباس یا اکسسوری می گردی؟ توصیفش کن تا برات پیدا کنیم.

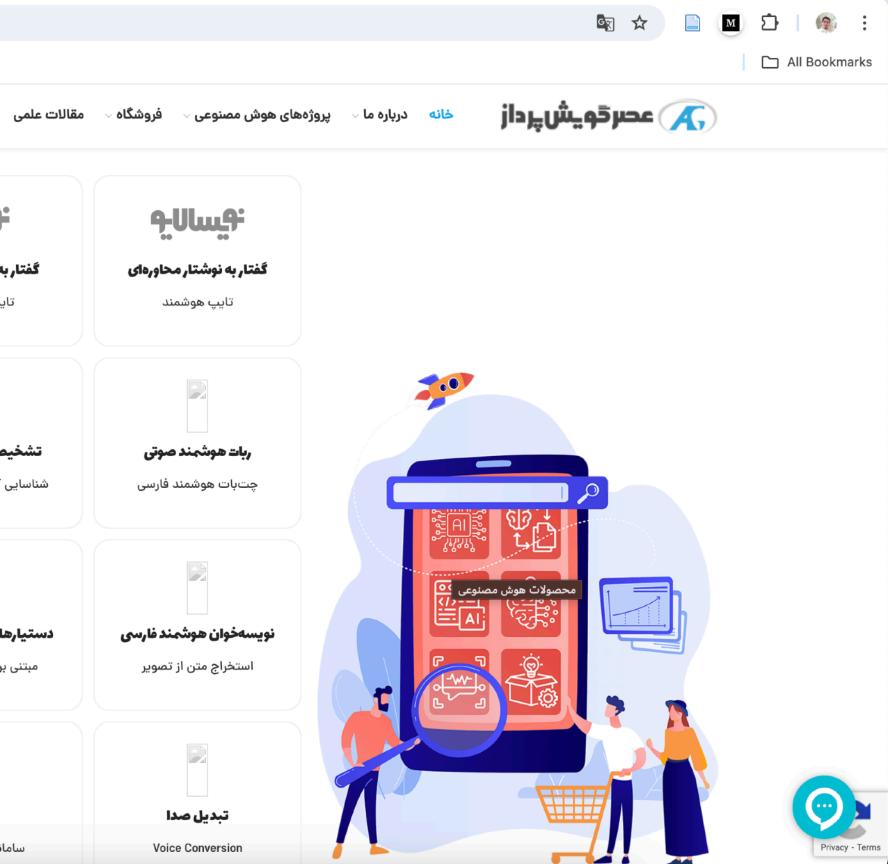
## **Applications (Iran)** Navigation



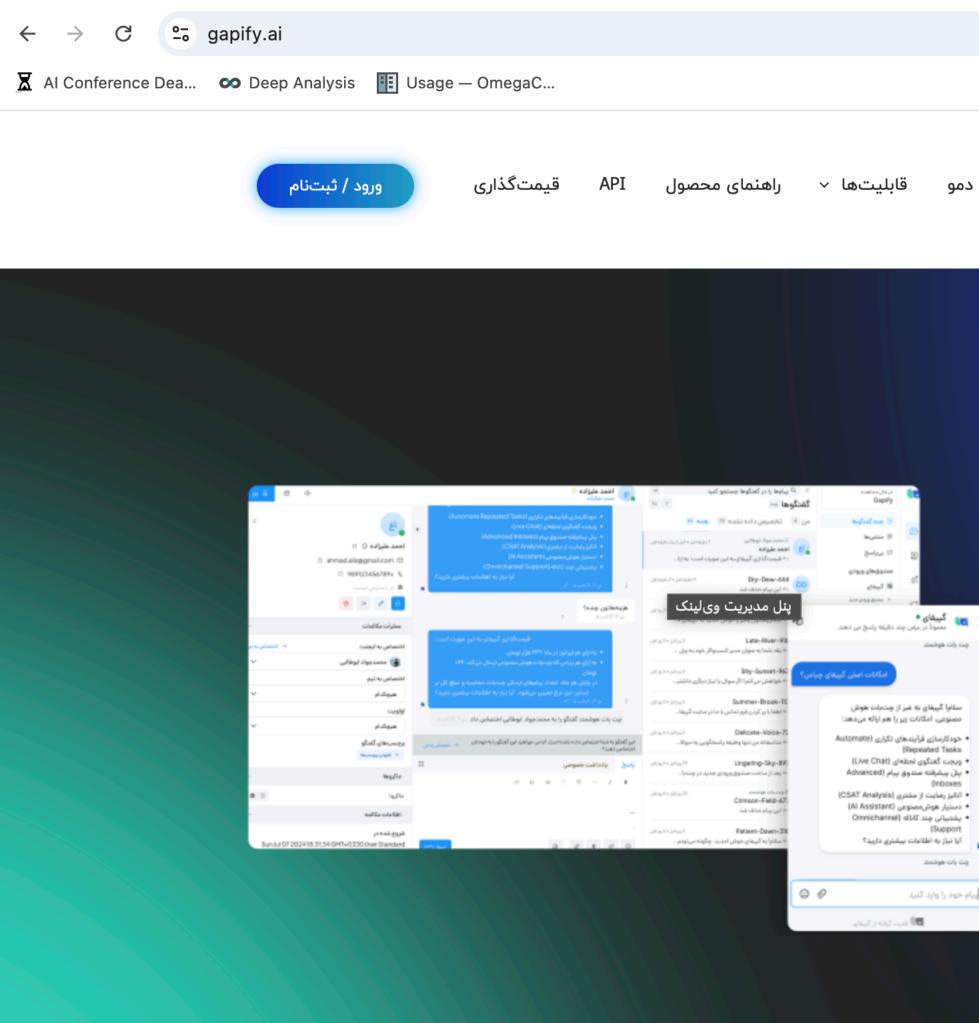


## **Applications (Iran) Speech Processing**

← → C	asr-gooyesh.com/fa/#proje	ctai	
X AI Conference Dea.	🗴 Deep Analysis 📑 Usag	e — OmegaC	
	° <u>°</u> QQ •"	مشاوره EN AI مشاوره ۱۰۶٬۱۹۳۱	وبلاگ 🗸 تماس با ما
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			••
	تلفن گویای هوشمند	متن به گفتار آریانا	ر به نوشتار رسمی
	پاسخگوی هوشمند تلفنی	متنخوان هوشمند فارسی	تایپ هوشمند
		1.5	
		رمزآوا	
	جستجو در گفتار	احراز هویت گفتاری	یص هویت صوتی
	جستجوگر کلمات کلیدی در صوت	رمز صوتی	بی گوینده از روی صدا
	پارسيا	تهيه و توليد	رهای صوتی هوشمند
	مترجم هوشمند گفتار به گفتار	دادگان هوش مصنوعی	ں بر هوش مصنوعی
$\bigcirc$		پروژەھا	شيوا
$\sim$		چ <b>ری</b> پروژههای سفارشی هوش مصنوعی	امانه بهبود گفتار
		پرورەمەي سەرسى موس سسوغى	المالة بهبود عسار



### Applications (Iran) Chat Bot



			G	τ <b>γ</b>	\$	M	Ď		3	•
								All B	lookma	arks
دریافت دمو	سوالات متداول	صفحه نخست	پیفای - Gapify	ػؖ	Í.					

#### ابزار چت آنلاین با مشتریان + هـوشمصنوعـی برای پاسخدهی سریع

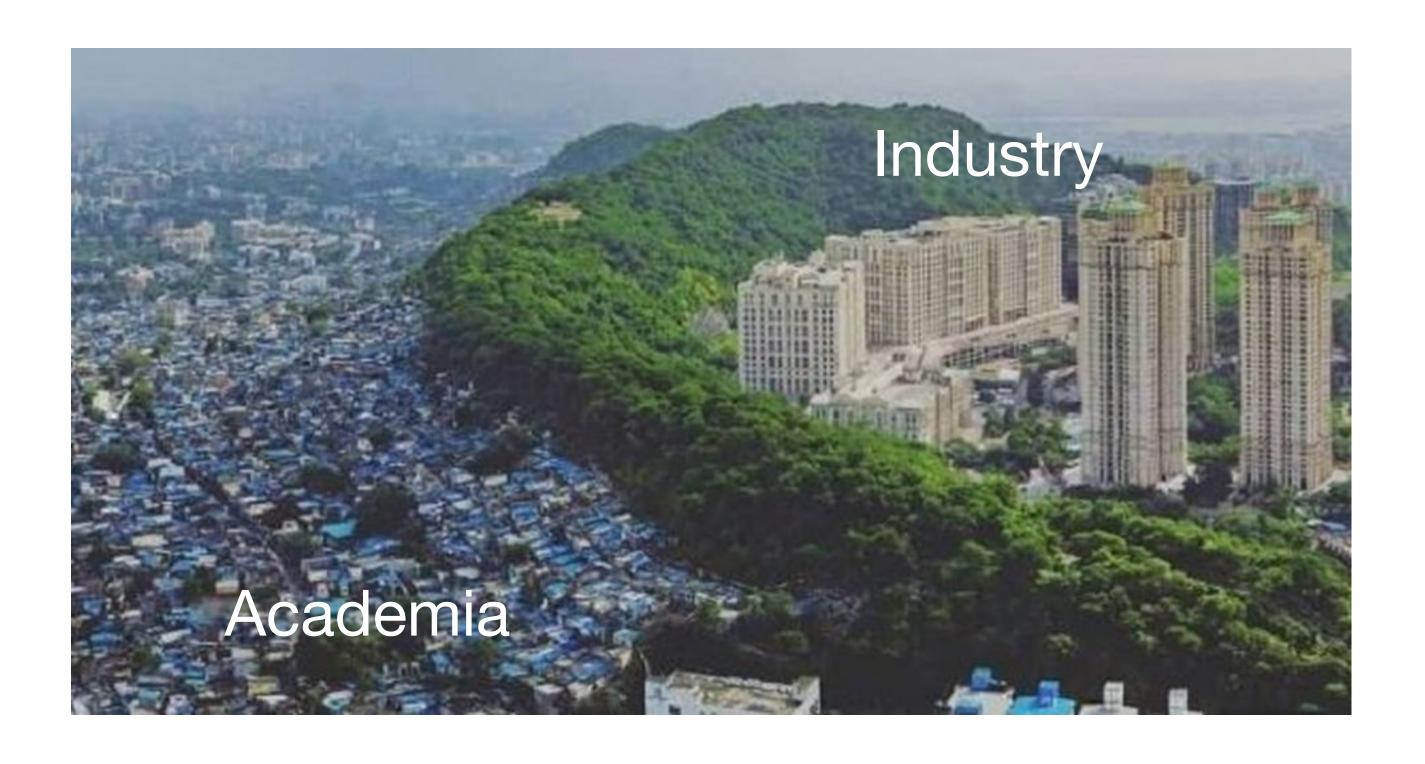
گپیفای ترکیبی از امکانات جامع چت آنلاین با مشتریان به اضافهی چتبات هوش مصنوعی (باتیفای) برای خودکارسازی حداکثری پاسخدهی به مشتریان است. با گپیفای هیچ مشتری بیپاسخ نمیماند.

گپیفای را امتحان میکنم

نیاز به اطلاعات بیشتری دارم

## Academia vs Industry

## Academia Vs Industry Money





Team of Multi Skill Experts

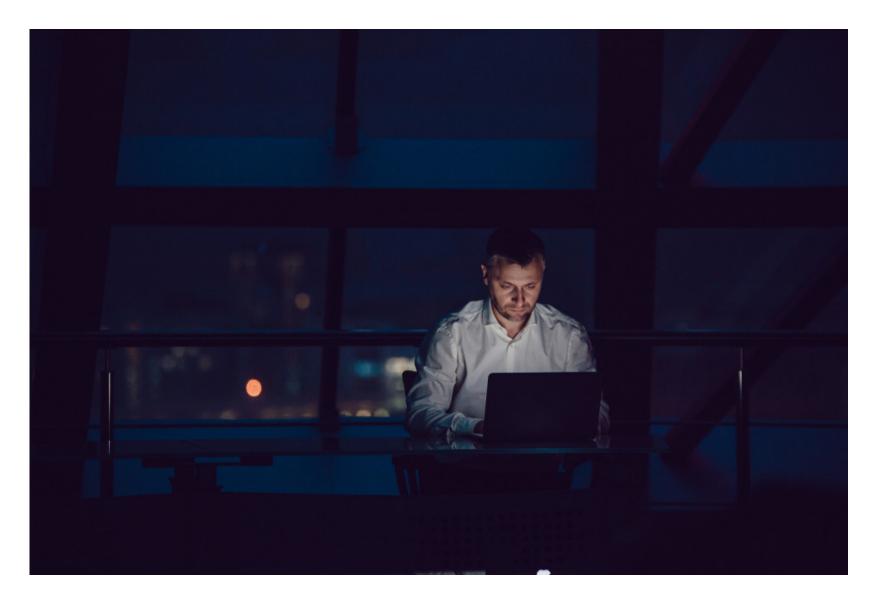


Infrastructure

#### **Academia Vs Industry** Team Work, Communication (Maybe our Problem)







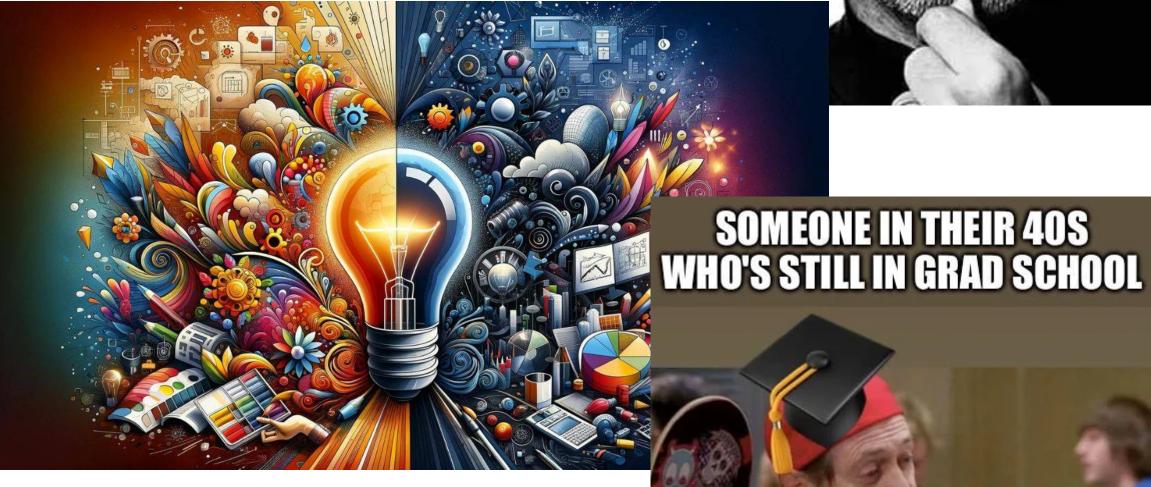


#### **Academia Vs Industry Priorities, Deadlines**



"The only way to do great work is to love what you do. If you haven't found it, keep looking. Don't settle."

- Steve Jobs







music # Band

## Academia Vs Industry Importance of Data



#### **Facebook and Google Are the New Data Brokers**

By Chris Hoofnagle | University of California, Berkeley

They claim not to sell data, but Facebook and Google have paid developers with data. This selling of your personal data is central to platform economics. Could quality signals change the terms of the bargain?

"We do not sell your personal information to anyone" [https://safety.google/privacy/ads-and-data/] – Google, Ads and Data Webpage

"No, we don't sell any of your information to anyone and we never will." [https://www.facebook.com/help/152637448140583?helpref=related] – Facebook, Does Facebook sell my information? Webpage





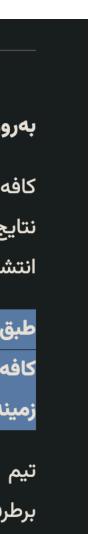
2 2	
157	
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159	سلام خیلی خوشحالم ک این وقت شب برای ما شب بیدار ها فعال هستید، لطفا در صورت امکان شیرموز پسته شکر نداشته باشد بهتر است. دکتر کاظمی
160	
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بهروزرسانی ۲: توضیح کافهبازار در مورد آسیبپذیری امنیتی گزارششده

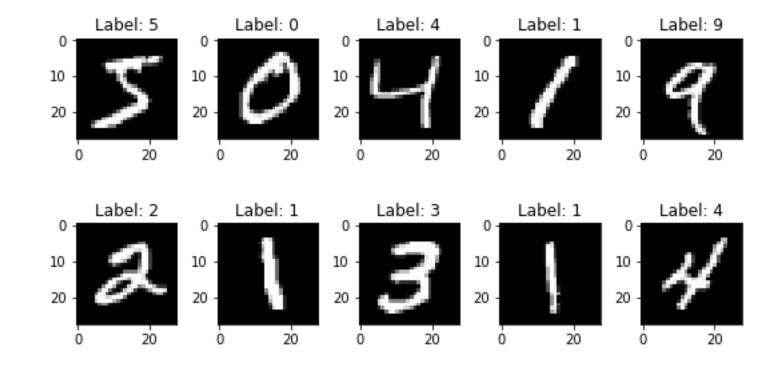
کافهبازار پیرو بیانیهی قبلی و در راستای سیاست خود مبنی بر شفافیت، کاربران را در جریان جزئیات جدید نتایج بررسیهای تیم فنی کافهبازار دربارهی گزارشی که صبح روز شنبه ۷ اردیبهشت در شبکههای اجتماعی انتشار یافت، قرار میدهد.

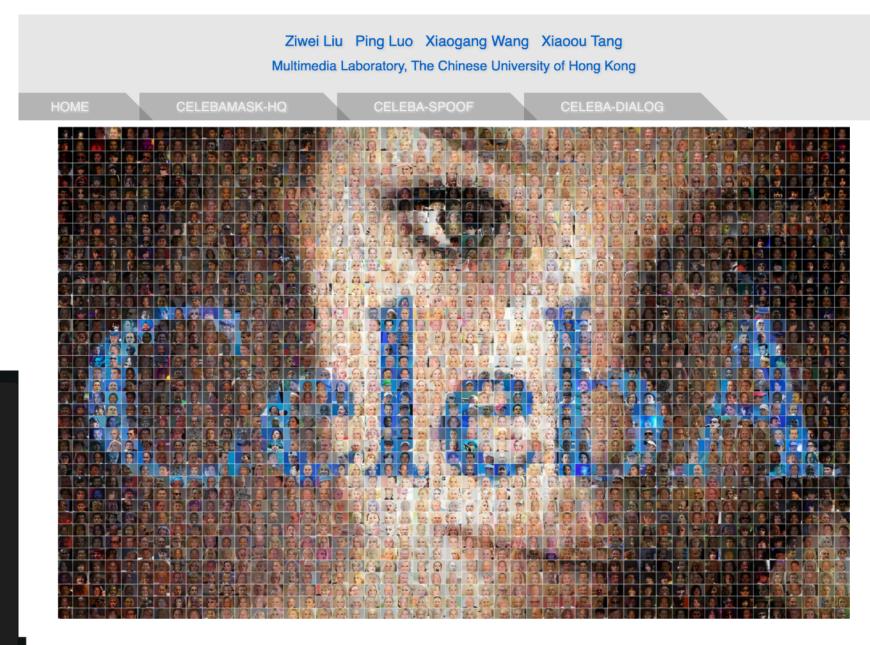
طبق بررسیهای تیم فنی کافهبازار مشخص شده است که منبعکد یکی از زیرسیستمهای وبسایت کافهبازار به دست افرادی خارج از مجموعه رسیده است؛ در عین حال، همچنان هیچگونه شواهدی در زمینه نشت اطلاعات حسابهای کاربری به دست نیامده است.

تیم فنی این مسئله را از ساعات ابتدایی روز شنبه با جدیت پیگیری کرد و اشکال امنیتی را شناسایی و برطرف کرد.



### **Academia Vs Industry** Data Type, Volume





```
"sort_timestamp": 1634275259292,
"rating": 3.0,
"helpful_votes": 0,
```

```
"title": "Meh",
```

"text": "These were lightweight and soft but much too small for my liking. I would have preferred two of these together to make one loc. For that reason I will not be repurchasing.",

```
"images": [
   "small_image_url": "https://m.media-amazon.com/images/I/81FN4c0VHzL._SL256_.jpg",
    "medium_image_url": "https://m.media-amazon.com/images/I/81FN4c0VHzL._SL800_.jpg",
    "large_image_url": "https://m.media-amazon.com/images/I/81FN4c0VHzL._SL1600_.jpg",
    "attachment_type": "IMAGE"
"asin": "B088SZDGXG",
"verified_purchase": true,
 'parent_asin": "B08BBQ29N5",
"user_id": "AEYORY2AVPMCPDV57CE337YU5LXA"
```

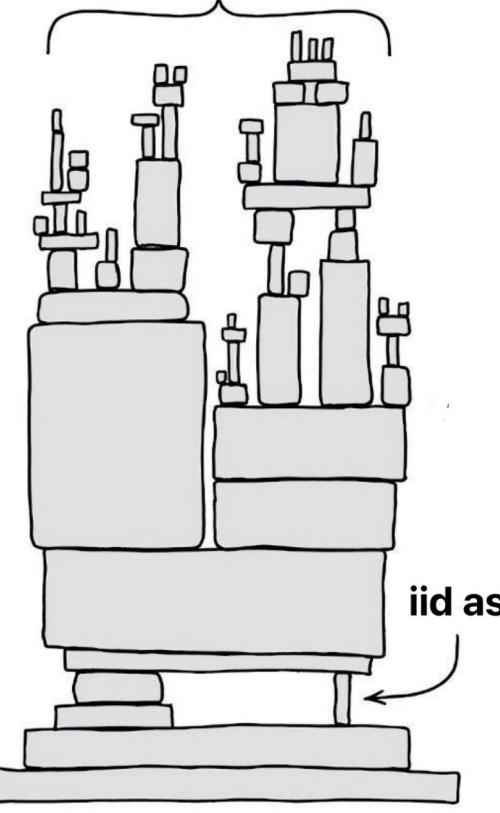






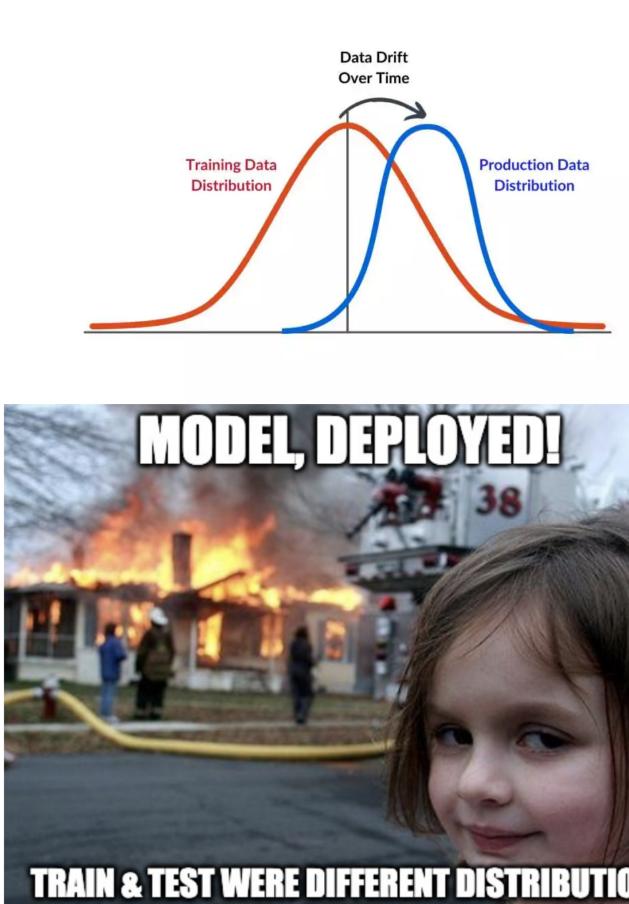
#### **Academia Vs Industry Data Quality**

Handling Imbalanced Dataset in ML





#### statistics and machine learning



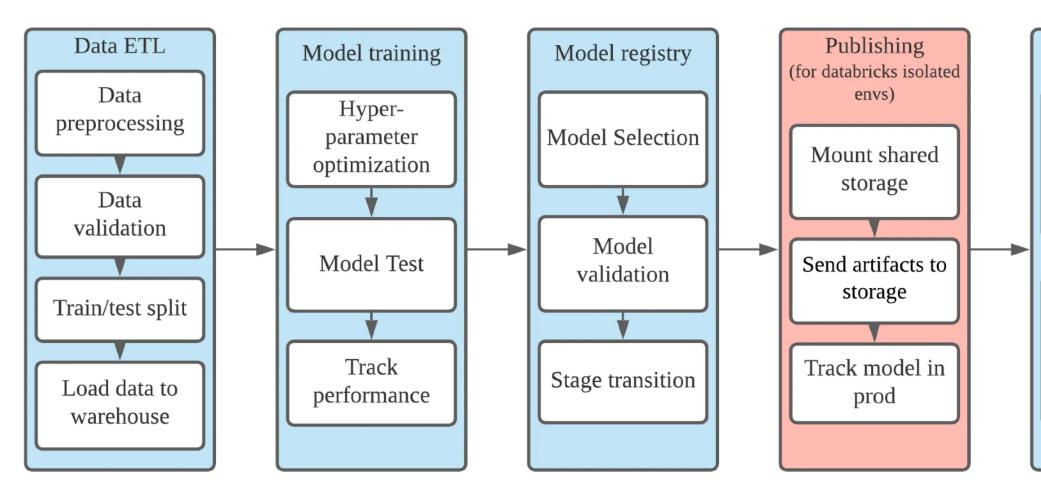
iid assumption



## Academia Vs Industry Infrastructure, Continuous Training

Some key facts about GPT-4:

- **Total parameters**  $\sim 1.8$  trillion (over 10x more than GPT-3)
- Architecture Uses a <u>mixture of experts (MoE)</u> model to improve scalability
- **Training compute** Trained on ~25,000 Nvidia A100 GPUs over 90-100 days
- **Training data** Trained on a dataset of ~13 trillion tokens
- Inference compute Runs on clusters of 128 A100 GPUs for efficient deployment
- Context length Supports up to 32,000 tokens of context



#### Running Large Language Models (Llama 3) on Apple Silicon with Apple's MLX Framework

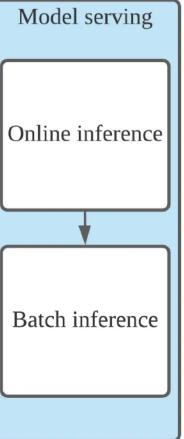
Step-by-Step Guide to Implement LLMs like Llama 3 Using Apple's MLX Framework on Apple Silicon (M1, M2, M3, M4)

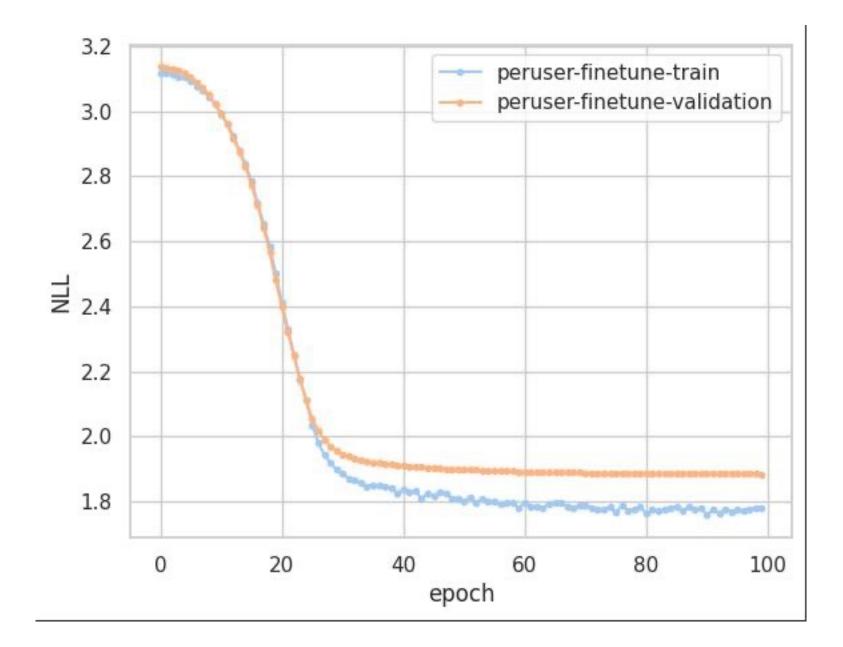


Manuel · Follow

5 min read • Jun 10, 2024

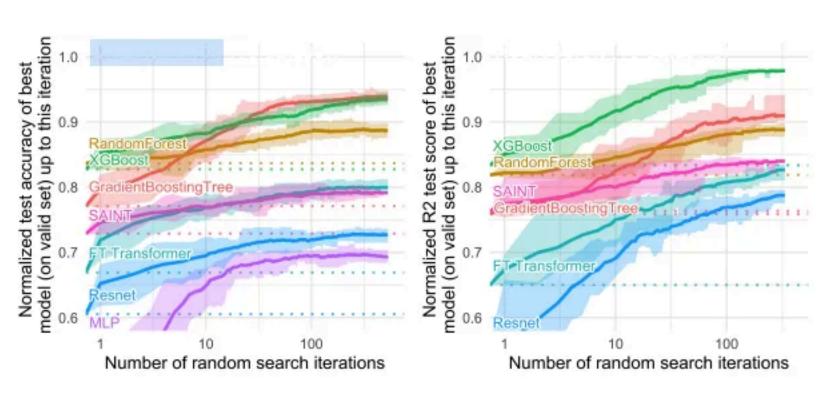
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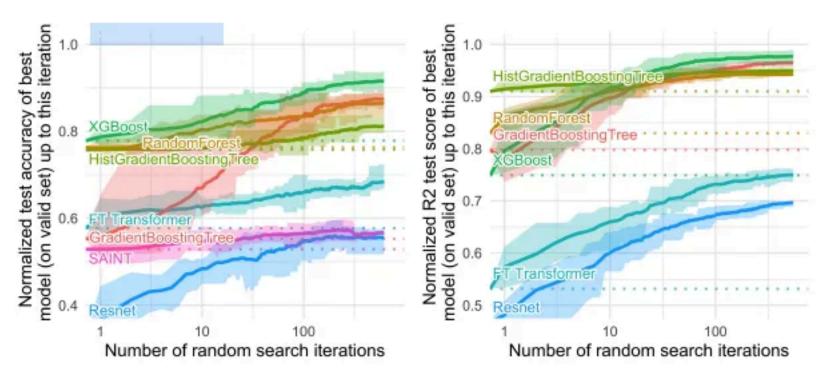


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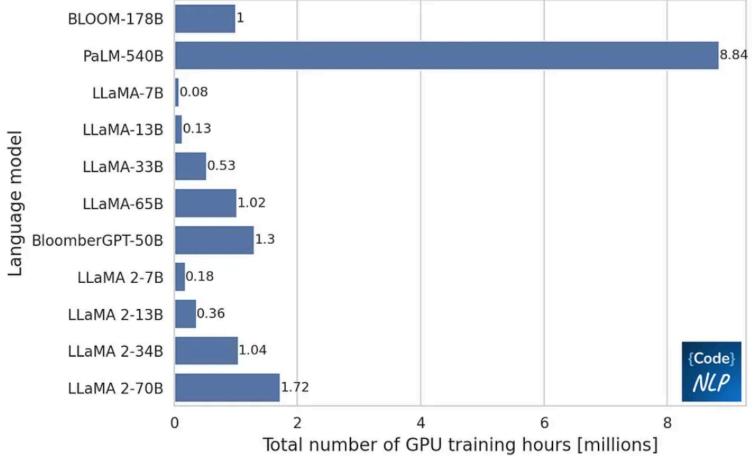
#### **Academia Vs Industry Solution Complexity**



Medium-sized datasets, with only numerical features







Model Name	Rossman	CoverType	Higgs	Gas	Eye	Gesture
XGBoost	$490.18 \pm 1.19$	$3.13\pm0.09$	$21.62\pm0.33$	$2.18\pm0.20$	56.07±0.65	$80.64 \pm 0.80$
NODE	$488.59 \pm 1.24$	$4.15 \pm 0.13$	$21.19\pm0.69$	$2.17\pm0.18$	$68.35 \pm 0.66$	$92.12\pm0.82$
DNF-Net	$503.83 \pm 1.41$	$3.96 \pm 0.11$	$23.68 \pm 0.83$	$1.44 \pm 0.09$	$68.38 \pm 0.65$	$86.98 \pm 0.74$
TabNet	485.12±1.93	$3.01 \pm 0.08$	$21.14 \pm 0.20$	$1.92\pm0.14$	$67.13 \pm 0.69$	$96.42 \pm 0.87$
1D-CNN	$493.81\pm2.23$	$3.51\pm0.13$	$22.33 \pm 0.73$	$1.79\pm0.19$	$67.9 \pm 0.64$	$97.89 \pm 0.82$
Simple Ensemble	$488.57 \pm 2.14$	$3.19\pm0.18$	$22.46 \pm 0.38$	$2.36 \pm 0.13$	$58.72 \pm 0.67$	$89.45 \pm 0.89$
Deep Ensemble w/o XGBoost	$489.94 \pm 2.09$	$3.52\pm0.10$	$22.41 \pm 0.54$	$1.98\pm0.13$	$69.28 \pm 0.62$	$93.50\pm0.75$
Deep Ensemble w XGBoost	$485.33 \pm 1.29$	$2.99 \pm 0.08$	$22.34 \pm 0.81$	$1.69 \pm 0.10$	$59.43 \pm 0.60$	78.93 ±0.73

Each column is a dataset. Better performances are highlighted with bold numbers.

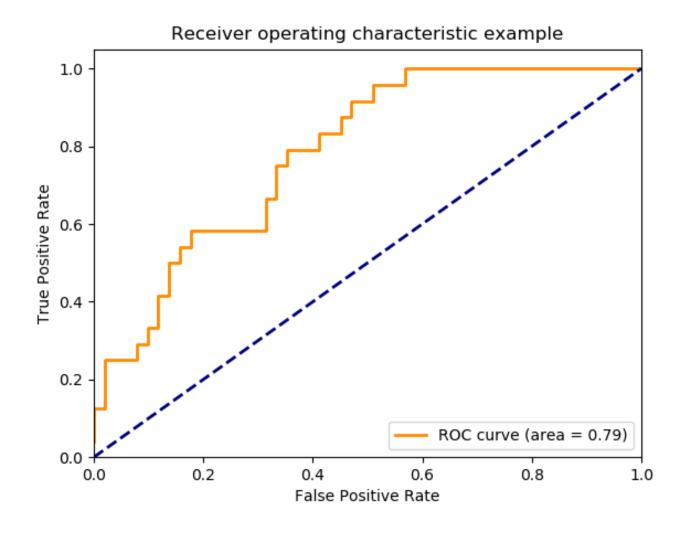
Model Name	YearPrediction	MSLR	Epsilon	Shrutime	Blastchar
XGBoost	$77.98 \pm 0.11$	55.43±2e-2	11.12±3e-2	$13.82\pm0.19$	$20.39 \pm 0.21$
NODE	$76.39 \pm 0.13$	55.72±3e-2	10.39±1e-2	$14.61 \pm 0.10$	$21.40 \pm 0.25$
DNF-Net	$81.21 \pm 0.18$	56.83±3e-2	12.23±4e-2	$16.8\pm0.09$	$27.91 \pm 0.17$
TabNet	$83.19\pm0.19$	56.04±1e-2	11.92±3e-2	$14.94\pm, 0.13$	$23.72 \pm 0.19$
1D-CNN	$78.94 \pm 0.14$	$55.97 \pm 4e-2$	11.08±6e-2	$15.31\pm0.16$	$24.68 \pm 0.22$
Simple Ensemble	$78.01 \pm 0.17$	$55.46 \pm 4e-2$	11.07±4e-2	$13.61\pm, 0.14$	$21.18 \pm 0.17$
Deep Ensemble w/o XGBoost	$78.99 \pm 0.11$	55.59±3e-2	10.95±1e-2	$14.69\pm0.11$	$24.25 \pm 0.22$
Deep Ensemble w XGBoost	76.19 ±0.21	55.38±1e-2	11.18±1e-2	13.10±0.15	20.18±0.10

Each column is a dataset. Better performances are highlighted with bold numbers.





#### **Academia Vs Industry Evaluation Metrics**



#### **Doubly Robust**

Estimating ATE with Doubly Robust estimator:

$$ATE_{DR} = \frac{1}{n} \sum_{i=1}^{n} \left[ \frac{T_i Y_i}{\hat{e}(X_i)} - \frac{\{T_i - \hat{e}(X_i)\}}{\hat{e}(X_i)} \hat{m}_1(X_i) \right] - \frac{1}{n} \sum_{i=1}^{n} \left[ \frac{(1 - T_i) Y_i}{1 - \hat{e}(X_i)} + \frac{\{T_i - \hat{e}(X_i)\}}{1 - \hat{e}(X_i)} \hat{m}_0(X_i) \right]$$

Unbiased if either propensity score or regression model is correct

This property is referred to as double robustness

Predicted values		ed values		
		True	False	
	True	True	False	Pacall - Sonsitivity - TP
		Positive (TP)	Negative (FN)	$\operatorname{Re} call = \operatorname{Sensitivity} = \frac{TP}{TP + FN}$
Actual			Type 1 Error	
Act	False	False	True Negative	Specificity = $\frac{TN}{TN + FP}$
		Positive (FP)	(TN)	$\frac{\text{Specificity}}{TN + FP}$
		Type 1 Error		
		D TP		$Accuracy = \frac{TP + TN}{TP + TN + FP + FN}$
		$\Pr ecision = \frac{11}{TP + FP}$		$F1 = \frac{2 x \operatorname{Pr} ecision x \operatorname{Recall}}{\operatorname{Pr} ecision + \operatorname{Recall}}$

#### B. Classification Model Results

$$m_0 = E(Y|T = 0, X)$$
$$m_1 = E(Y|T = 1, X)$$



#### Academia Vs Industry Values, Blue or Red pill

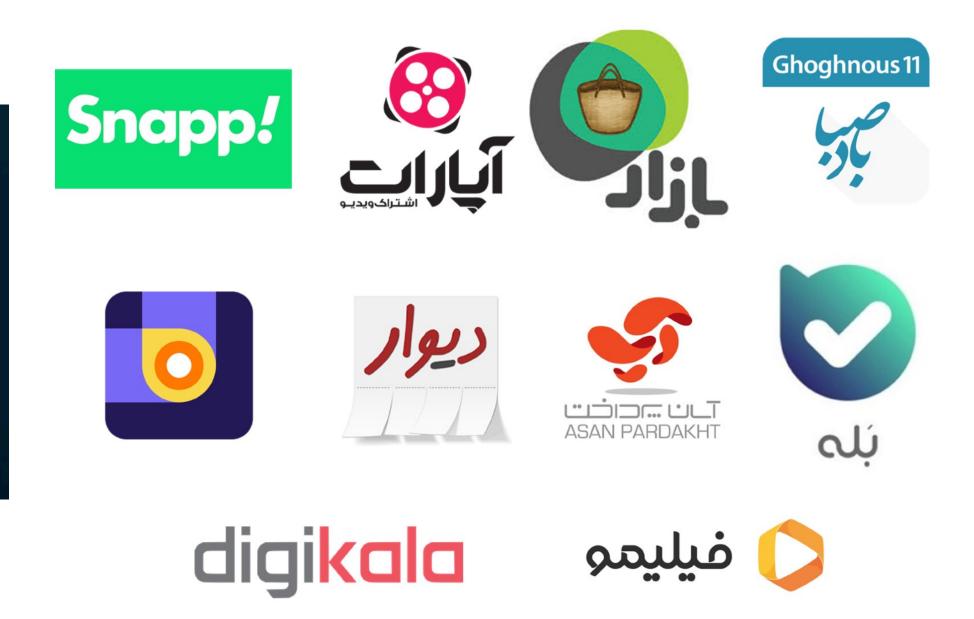
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•	Articles	About 8,700,000 results (0.06 sec)	
	Any time	[PDF] Attention is all you need	
	Since 2024	A Vaswani - Advances in Neural Information Processing Systems, 2017 - user.phil.hhu.de	
	Since 2023	Attention is all you need Attention is all you need	
	Since 2020	☆ Save ワワ Cite Cited by 136701 Related articles ≫	
	Custom range		





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The Choice is Yours!



#### **Academia Vs Industry Project Management**











Sprint Review Meeting Release

**Product Backlog** 

#### **SCRUM** METHODOLOGY

**Sprint Week** 

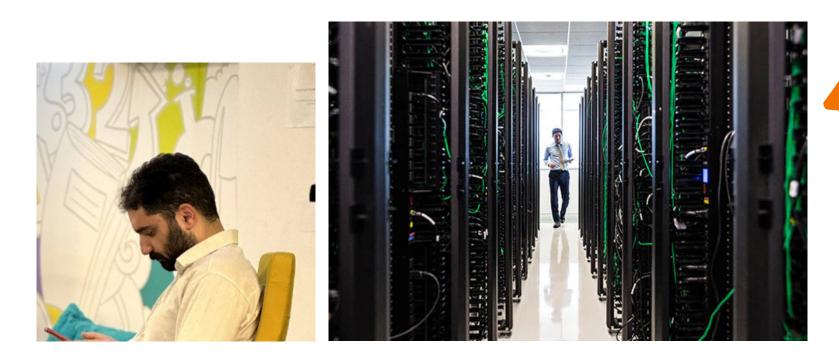


## Scrum and Machine Learning

#### **Positions Machine Learning Roles**



ML Engineer



Data Engineer

Infrastructure



Machine Learning



**Data Scientist** 





**Business** 

Data Analyst

#### Scrum and ML **Product Manager is Super Important**

- View ML system as product is HARD.
- Maintaining Backlog is a Job.
- Finding Business Opportunities.
- Academic People do not pay attention to business.
- Intra-team Communications.
- Ego in academic people.
- Technical Product Manager (TPM).







### Scrum and ML Standup Meetings

- Sense of Progress.
- Fail-fast Mindset.
- Pickup fights.
- Speed Control.
- Inter-team speed difference.
- Irrelevant tasks Issue.
- Means Start of The Day.
- Easily Could Break the Time Box.



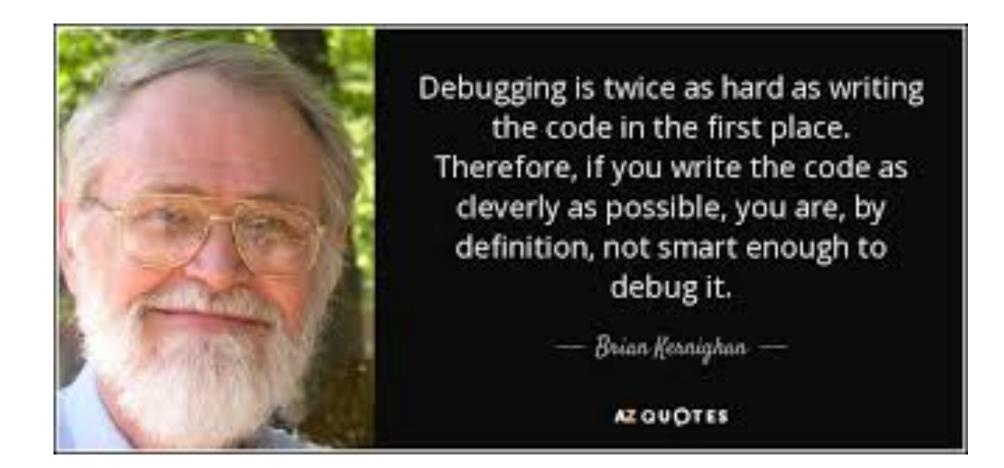
#### LACINAI Lack of Clarity in Communication



## Scrum and ML **Refinement / Vague Backlog**

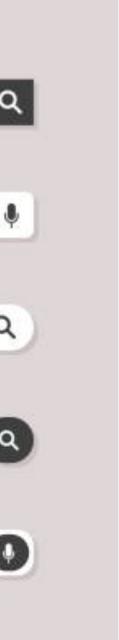
- **Define DoDs are challenging in ML Systems.**
- Data Driven Scrum Insights.

How to break bigger down into prioritized TASKS



Time Boxing Issues. (EDA, Increasing performance x%, debugging, …)

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S	Search	٩	Search	
	٩	Search	Q   Search	
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## Scrum and ML Refinement

- Curse of Brilliant People.
- Performance Monitoring.
- Maintenance.
- POC in machine learning.
- Double checking core idea/implementation.
- Novel things estimation?
- It's enough.
- Consider Buffer.

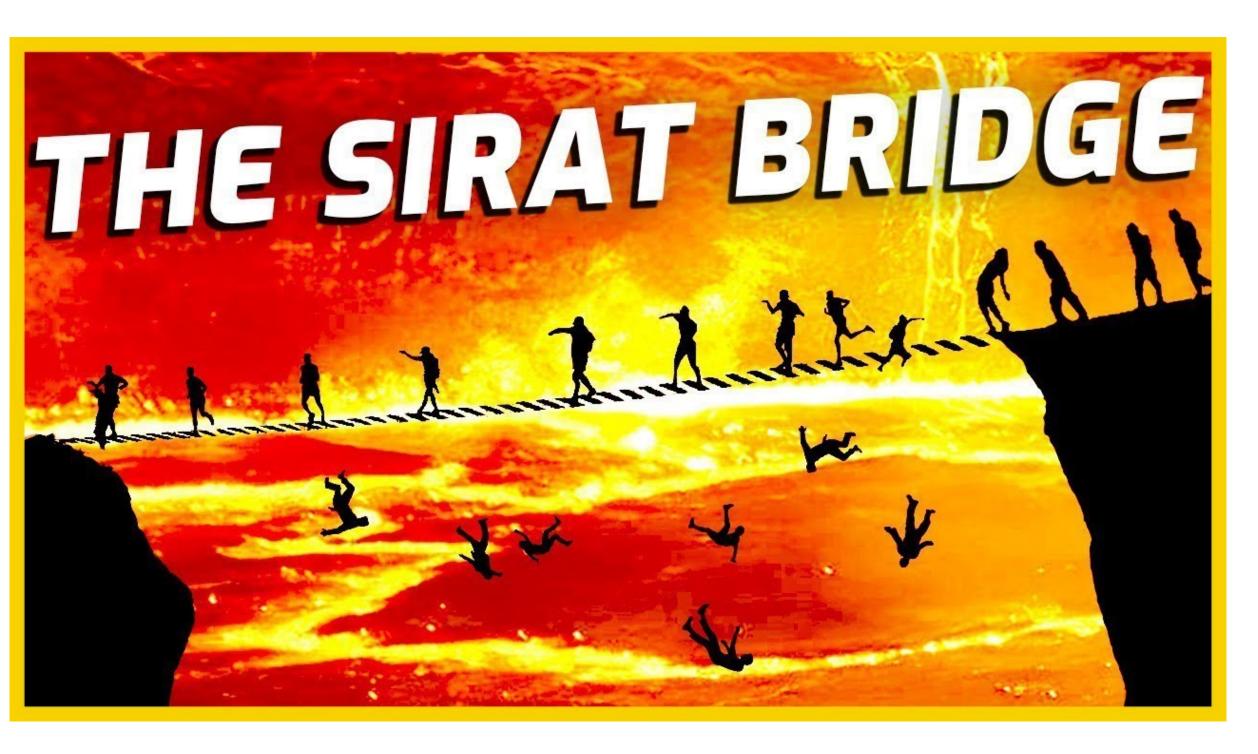




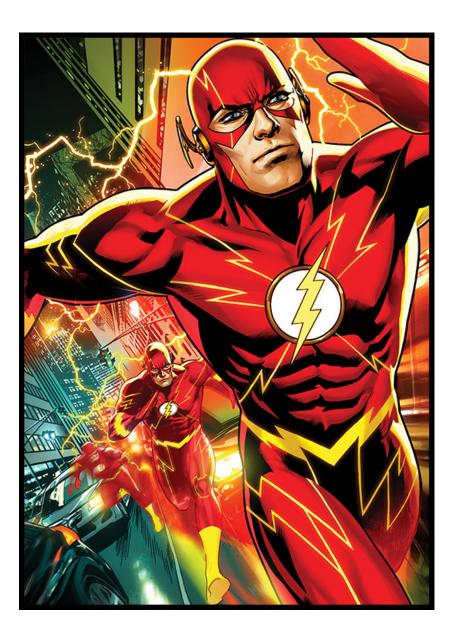
#### Scrum and ML Review

- Evaluation Metric of Sprint.
- You fire some one here.
- Excuses emerge here.
- Be strict on DoDs.







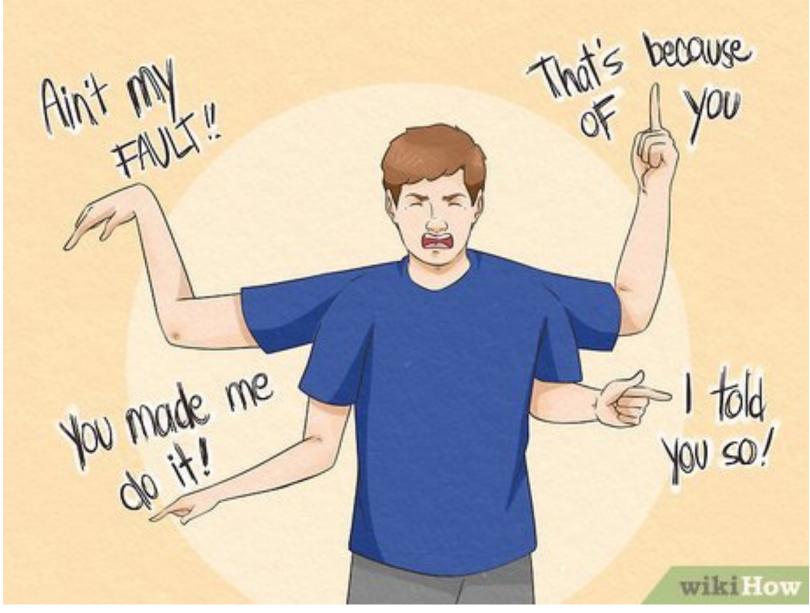


## **Scrum and ML** Retro

- If you do not seek team's progress:
  - Go to therapy.
  - Leave.
- Successful Stories.







#### **How to Success** As a Data Scientist

- Learn fundamentals by your heart.
- Intuitively know what you do. •
- Blend in teams.
- Not too aggressive, Not a potato.
- Be ambitious.
- Be Multi-Dimensional.
- Work Ethics.

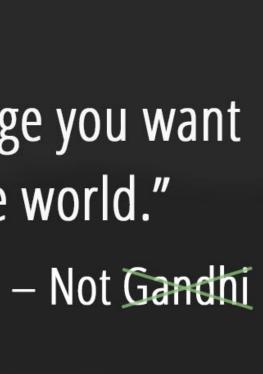


You should become Gholamreza in your field



loseph Ranset

•• Be the change you want to see in the world."





# ANYQUESTONSP IF NOT, JUST GLAP makeameme.org