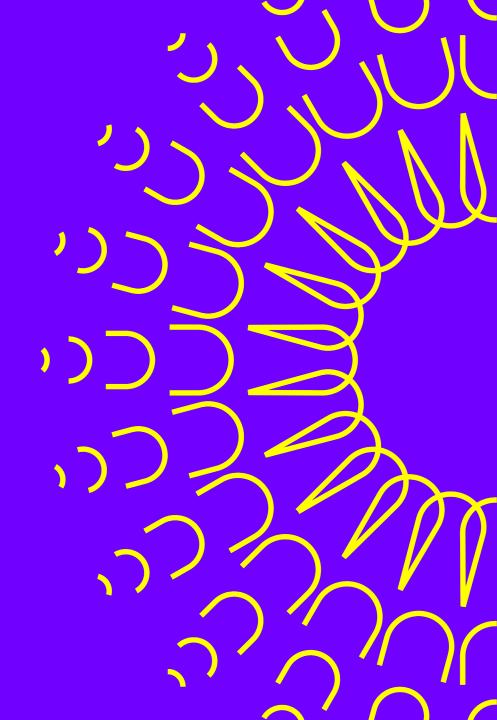
Uzum

# Break, Learn, Refine – The Art of **Hypothesis**-**Driven Development** of ML-Powered Search

Andrey Kulagin

Head of Machine Learning @ Uzum Market Haystack EU 2023

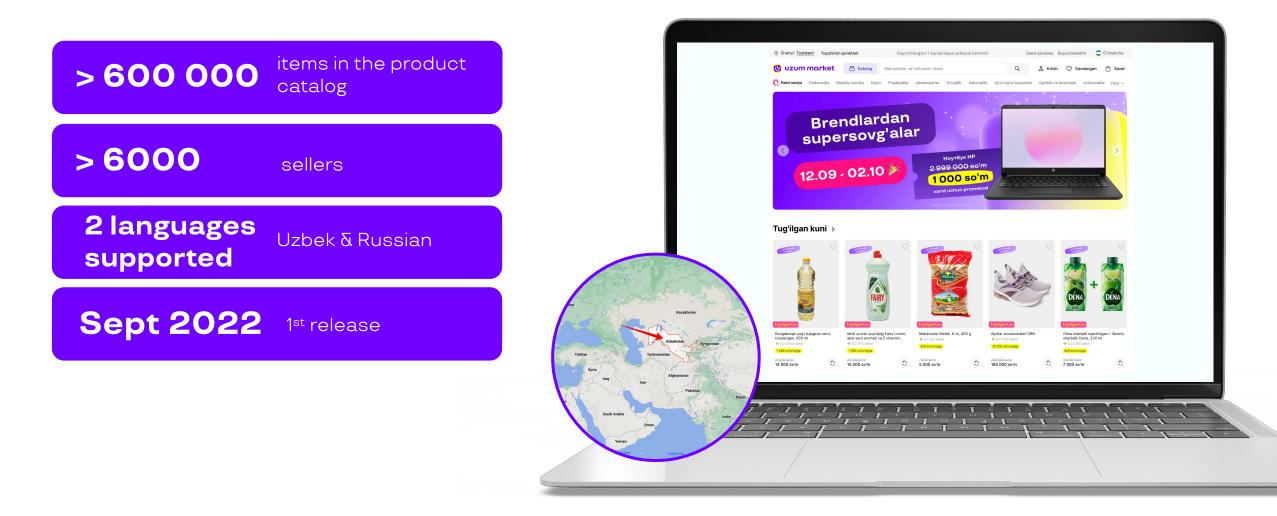


#### Agenda

- Chapter 1: Uzum Market and its Search
- Chapter 2: Search complexity and how to handle it
- Chapter 3: Hypothesis driven development of ML-powered Search
  - 1. Right direction
  - 2. Fast iterations
  - 3. High chances of success
  - 4. Stable results

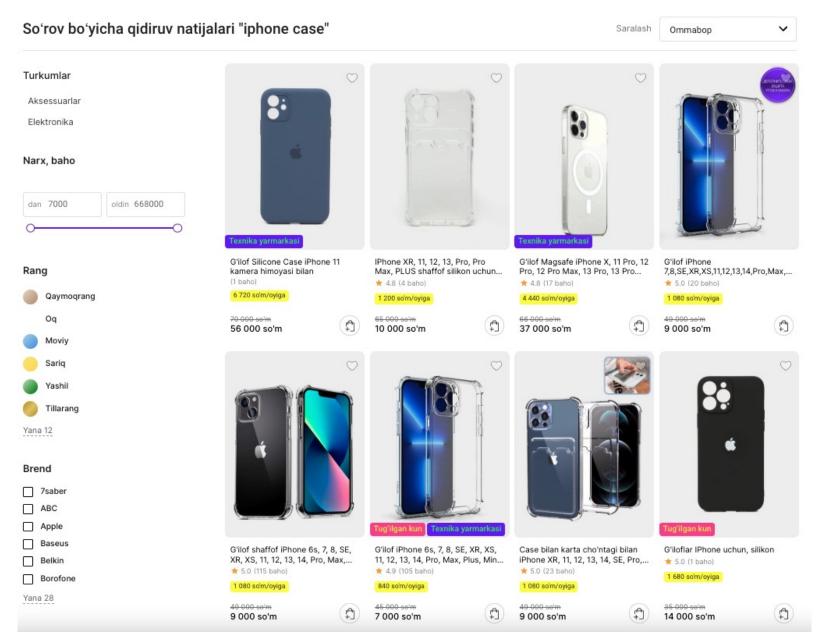
# Chapter1 UZUM MARKET AND ITS SEARCH

#### **UZUM MARKET**





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🕓 Halol nasiya	Elektronika	Maishiy texnika	Kiyim	Poyabzallar	Aksessuarlar	Goʻzallik	Salomatlik	Uy-roʻz	gʻor buyumlari	Qurilish va t	a'mirlash	Avtotovarla	ar Yana ∨
Bosh sahifa / Barcha toifalar													





# WHY SEARCH IS IMPORTANT FOR THE MARKETPLACE? С Искать товары и категории

- The most common marketplace use-case is to purchase a particular item
- A good search system becomes essential for navigation when you have 100k+ of items (1000k+ in the future).



\$



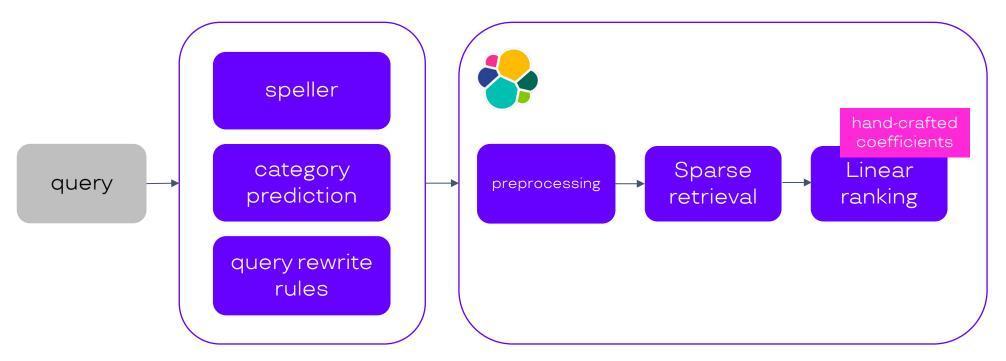




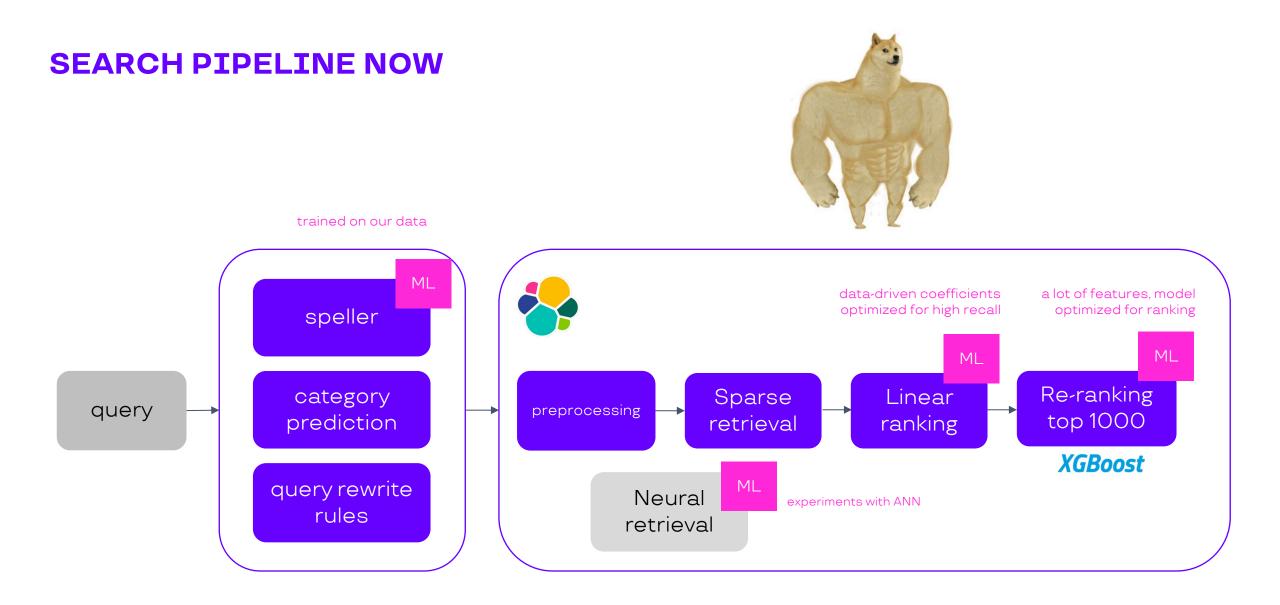


#### SEARCH PIPELINE 1 YEAR AGO

bm25\_score \* A + num\_orders \* B + rating \* C









Chapter 2 SEARCH COMPLEXITY AND HOW TO HANDLE IT

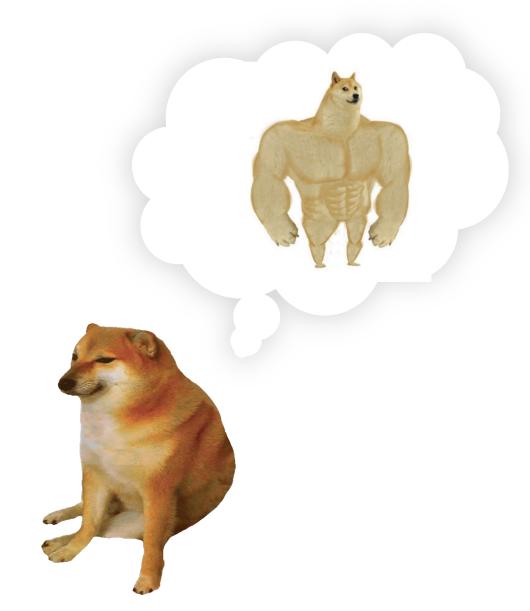
#### THE LONG WAY AHEAD

The search is broken! When will you fix it?

> Just build the search like Google

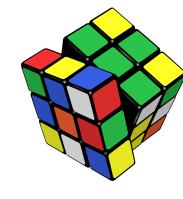
Why product X isn't shown for query Y?

I don't like the results!











#### HARD REALITY

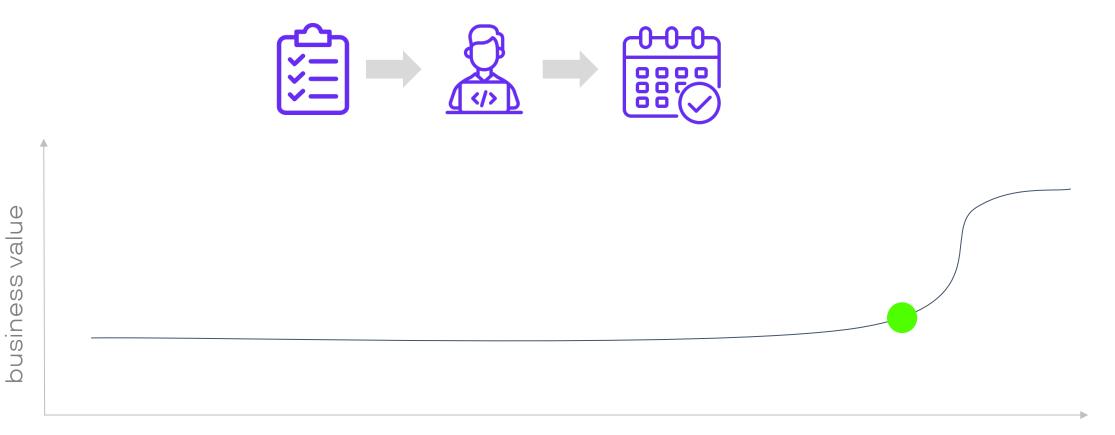
- Extreme complexity and high uncertainty
- Hundreds of new sellers and thousands of new products every day
- Seasonality and new emerging trends  $\square 
  abla$
- Balance between buyers and sellers interests  $rac{44}{2}$
- A lot of RnD 差

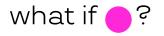






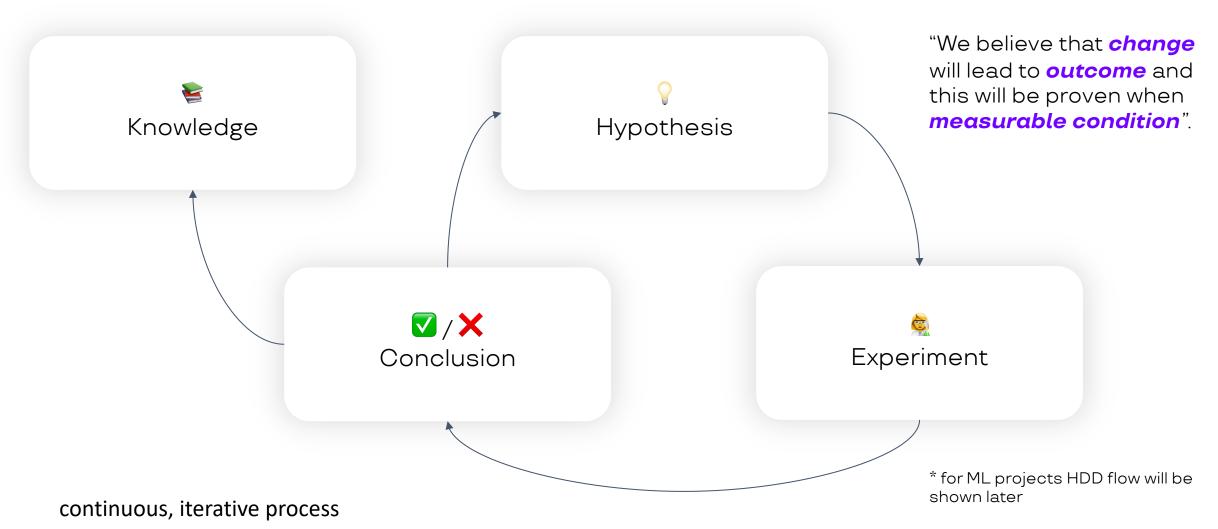
#### **REQUIREMENTS-DRIVEN DEVELOPMENT...** ...is not going to work here







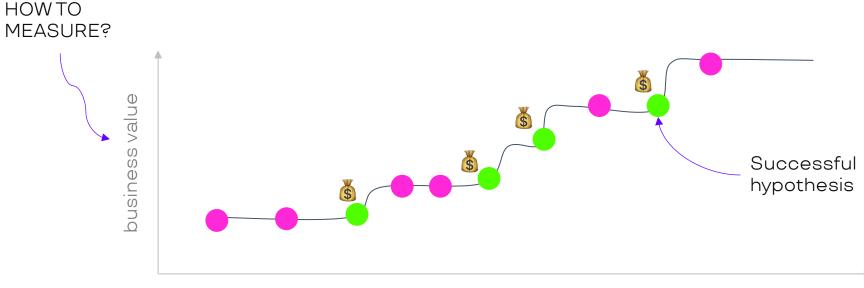
#### **HYPOTHESIS-DRIVEN DEVELOPMENT**





#### **ADVANTAGES OF HDD**

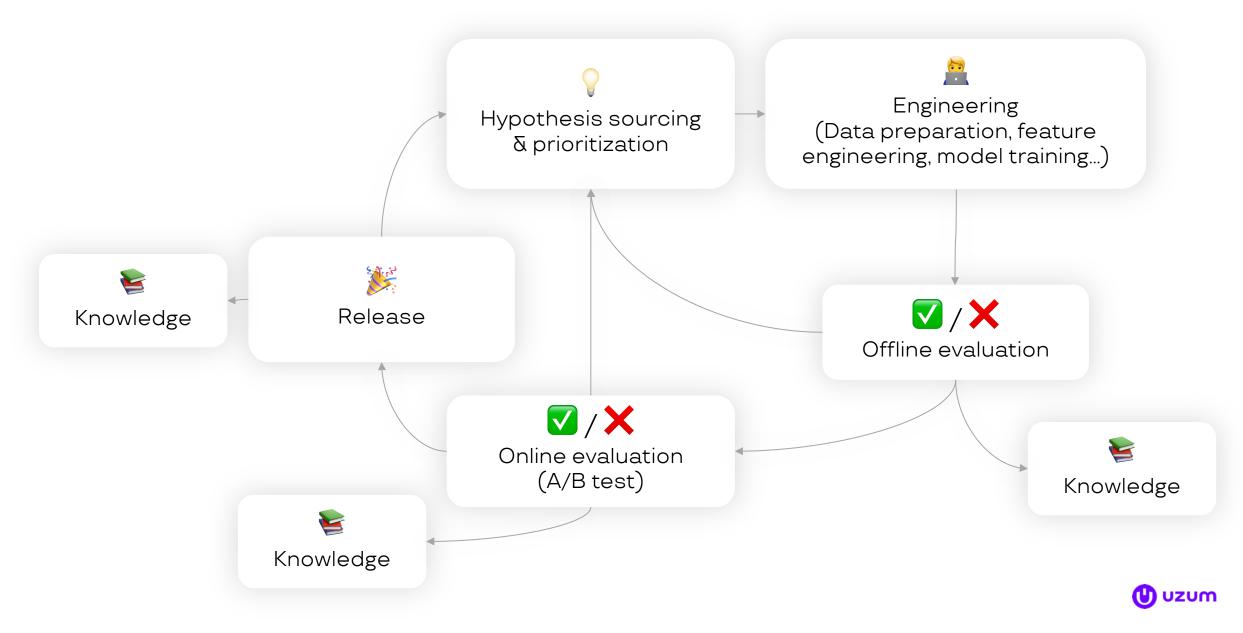
- Regular delivery of additional business value
- Reduced time to market  $\overline{\mathbb{Z}}$
- Decreased delivery risks
- Team management (dopamine) 😃

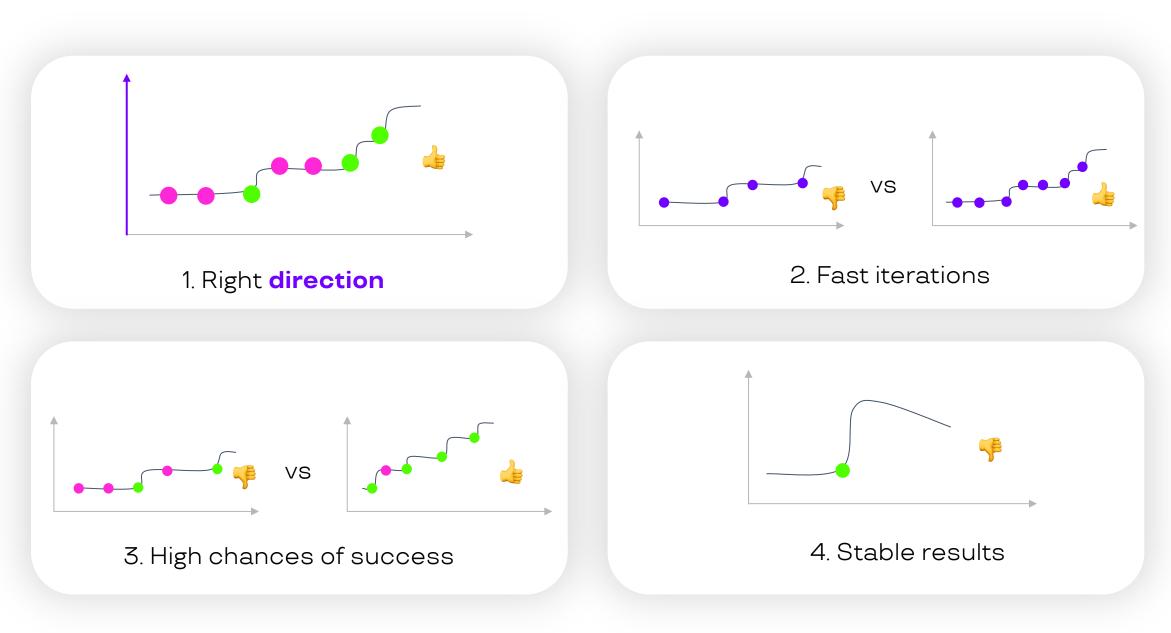




# Chapter 3 HYPOTHESIS DRIVEN DEVELOPMENT OF ML-POWRED SEARCH

#### **HDD FOR ML PROJECTS**







### 1. Right direction: It's all about the correct metrics & evaluation procedures

With the second secon



## WHAT IS THE BEST METRIC FOR SEARCH QUALITY?

# **CTO@k** - how many "bad" results your CTO has found among their *k* random searches

\* where k directly correlates with their free time



#### **BUSINESS METRICS**

- Global
  - ARPU, ARPPU, conversion to purchase, AOV, retention, LTV ...
- Search related
  - conversions from search to click, to add-to-cart, to order, to purchase
  - search ARPU / ARPPU
  - # empty queries
  - search abandonment rate
  - o ...

can be used only as **online metrics** 

we are especially interested in metrics which:

- are connected to the current business strategy
- can be tested during A/B (with adequate MDE / time)



#### **ONLINE METRICS: SEARCH-SESSION-WISE CONVERSIONS**

- 1. CR search2click (= 4/5)
- 2. CR search2atc (= 2/5) fraction of searches which resulted in at least one product from SERP added to cart
- **3. CR search2purchase** (= 1/5) fraction of searches, which resulted in at least one product from SERP purchased <<< requires attribution modelling

	date, user, session_id	query	click	ATC	purchase
all of these are	2022-01-01, Jane, 343g9n	"socks"	$\checkmark$		
	2022-01-01, Jane, 343g9n	"iphone"	$\checkmark$	$\checkmark$	$\checkmark$
ratio-metrics	2022-01-01, Mark, s9g55n	"socks"			
	2022-01-01, Mark, s9g55n	"sunflower oil"	$\checkmark$	$\checkmark$	
	2022-01-01, Mark, s9g55n	"t shirt"	$\checkmark$		



#### ONLINE METRICS: GLOBAL DAILY METRICS, RELATED TO 💰

- 1. ARPU\_daily (ARPDAU) (= \$30 / 5 = \$6) Average Revenue Per Daily Active User
- 2. cr2purchase\_daily (= 3/5) fraction of daily active users who made a purchase
- **3.** ARPPU\_daily (ARPPDAU) (= \$30 / 3 = \$10) Average Revenue Per Paying Daily Active User (usually equal to AOV except cases when several orders are made on the same day)

ARPU_daily = cr2purchase_daily * ARPPU_daily			date, user	made a purchase	spend money
Î	Î		2022-01-01, Jane	$\checkmark$	10 \$
responsible for total	frequency	amount	2022-01-01, Mark		
		anount	2022-01-02, Jane	$\checkmark$	15 \$
GMV			2022-01-02, Bob	$\checkmark$	5\$
			2022-01-03, Alex		

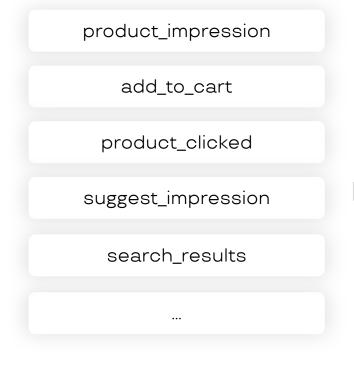


#### ONLINE METRICS: SEARCH DAILY METRICS, RELATED TO 💰

- 1. ARPU\_daily\_search = attributed\_to\_search\_revenue / n\_search\_visitors
- 2. cr2purchase\_daily\_search = n\_search\_buyers / n\_search\_visitors
- 3. ARPPU\_daily\_search = attributed\_to\_search\_revenue / n\_search\_buyers



#### **CLICKSTREAM**





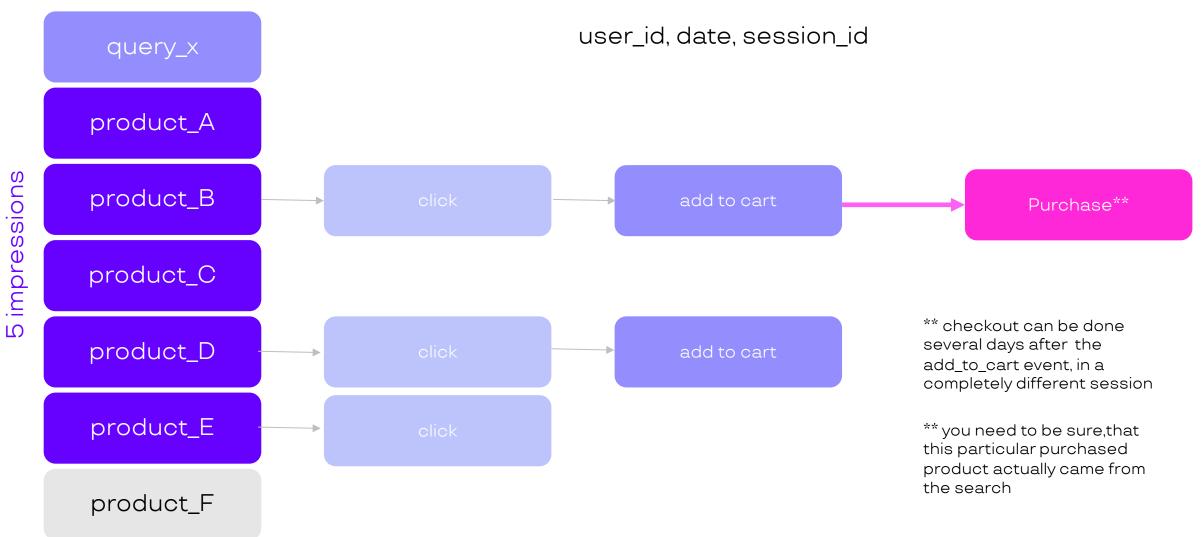
clients generate events

#### **Clickstream service**

- sessionize events
- enrich with user identification data
- clean
- convert to required format

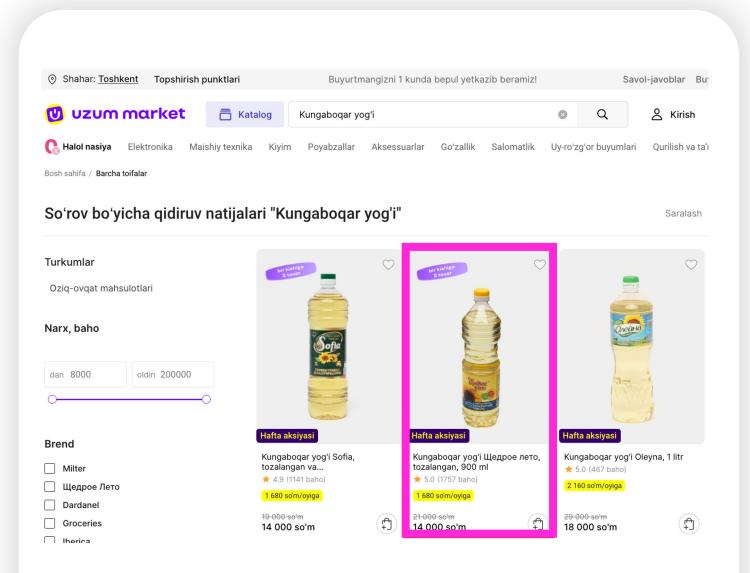


#### **ESSENTIAL DATA**



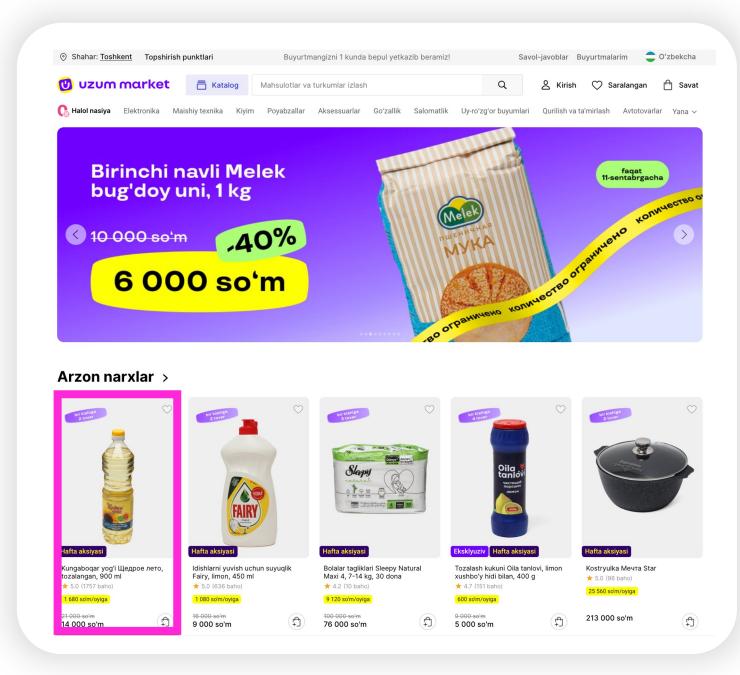


#### Search results, query="Sunflower oil"



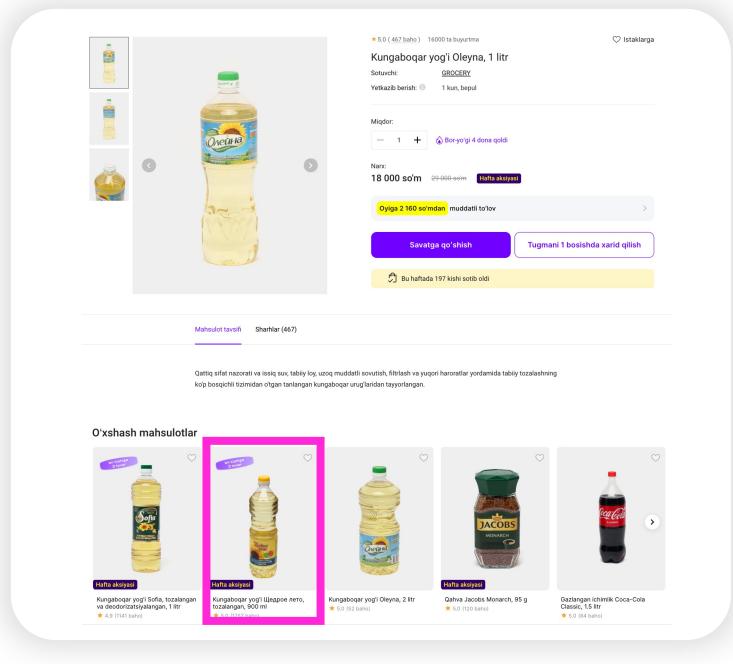


#### Main page, "Sale" collection



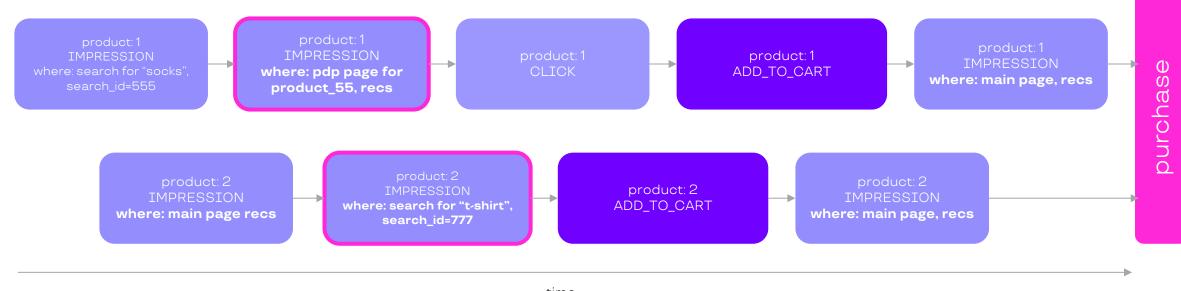


#### Other product's description page (PDP), "Similar products" collection





#### LAST IMPRESSION BEFORE LAST ADD-TO-CARD



time

order_item_id	product	where found	additional_info	session_id
34963	1	recs on pdp	product_id=55	q5g67
34964	2	search	q="t-shirt", search_id=777	f9486



#### **ATTRIBUTION MODELING**



this table is a  $\heartsuit$  and actively used not only in search team



#### **ATTRIBUTION MODELING**

Congrats!

Now we can easily measure metrics like **cr\_search2purchase** and **ARPU\_daily\_search**, **ARPPU\_daily\_search** 

In A/B tests too 🎉



#### MY PERSONAL TOP OF MISTAKES RELATED TO A/B TESTS

- 1. Apply T-test for ratio-metrics
  - Just use Delta-method: <u>Applying the Delta Method in Metric Analytics: A Practical Guide with</u> <u>Novel Ideas</u>
  - Or Linearization: <u>Approximations for Mean and Variance of a Ratio</u>
- 2. Run A/B without proper design
  - $\circ \quad \mbox{Calculate MDE $\delta$ sample-size BEFORE test}$
  - Don't forget to handle multiple comparisons problem
  - Run A/A-test (simulation) for every new metric
- 3. Bugs in metric calculation
- 4. Run an experiment without events logged
- 5. Mess up group labels ("B" and "A") 🙂
- 6. Forget to accurately document A/B results



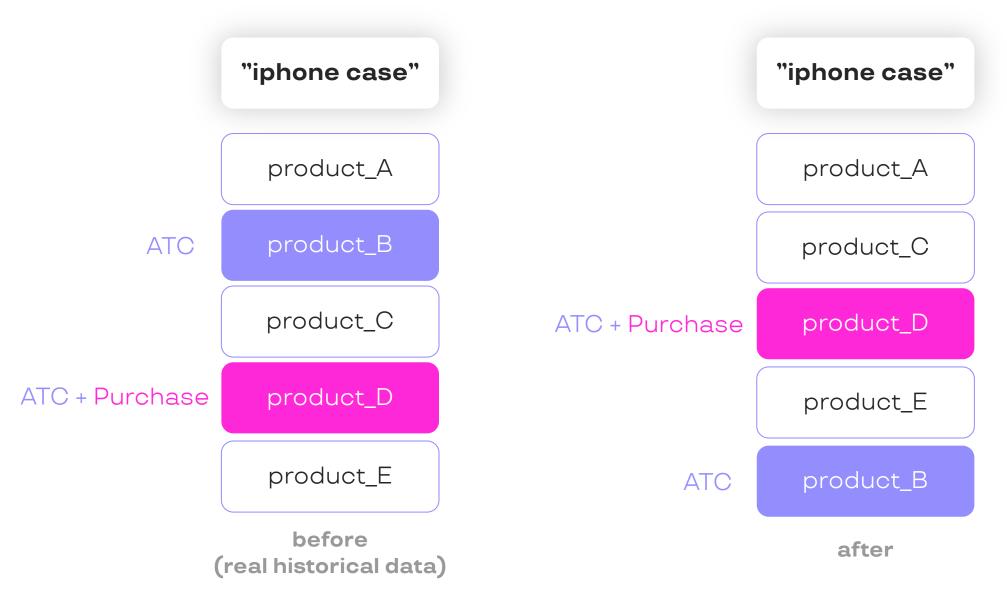
#### A/B TEST IS AN EXPENSIVE PROCEDURE

- Time for a/b itself
- Effort to prepare, conduct and analyse (engineers, analytics)
- User base is limited
- There is always a risk, that group B is worse

A lot of hypothesis can be checked preliminarily on historical data



#### WE NEED PROXY METRICS FOR OFFLINE EVALUATION





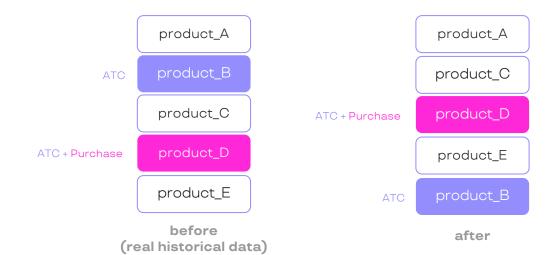
#### **RANKING METRICS**

- NDCG@k Normalized Discounted Cumulative Gain
  - 0 impression, 1 click, 2 atc, 3 order?
  - 0 impression, 1 order?
- MAP@k Mean Average Precision
- MRR Mean Reciprocal Rank
- ERR Expected Reciprocal Rank

Which is the best? How to choose "relevant" signal? How to determine k?



## **RANKING METRICS**



metric	relevance signal	Before	After	Change in %
NDCG	0 - impression, 1 - atc	0.65	0.54	-17%
NDCG	0 - impression, 1 - purchase	0.43	0.5	+16%
NDCG	0 - impression, 1 - atc, 2 - purchase	0.56	0.52	-6%
MRR	0 - impression, 1 - atc	0.50	0.33	-33%
MRR	0 - impression, 1 - purchase	0.25	0.33	+33%
mean first atc pos	0 - impression, 1 - atc	2	3	+50%
mean first order pos	0 - impression, 1 - purchase	4	3	-25%



# How to determine the best offline metric?



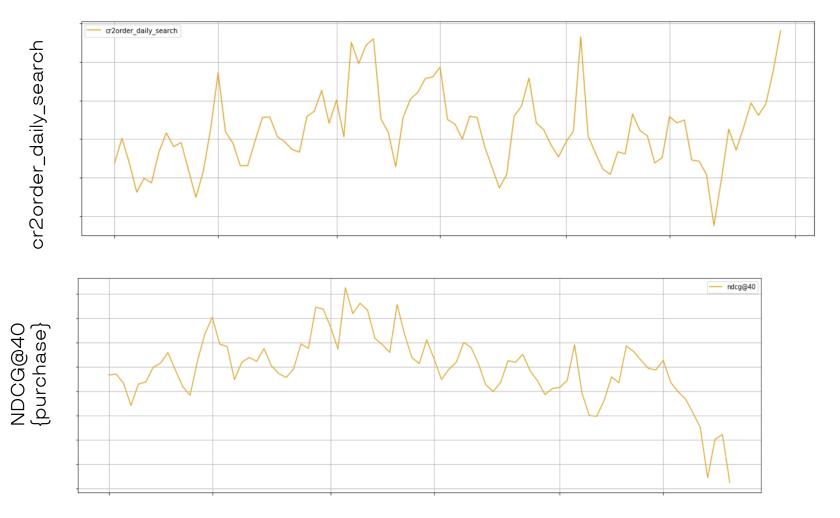
## **THE PROPER APPROACH:** from experiments history

A/B #	Offline metric 1 uplift	Offline metric 2 uplift	Offline metric 3 uplift	Target online metric uplift
001	+23%	+15%	-10%	-5%
002	-10%	-5%	+8%	+4%
003	+5%	-5%	+4%	+8%
004	+3%	-7%	-4%	-10%

Train a model that predicts

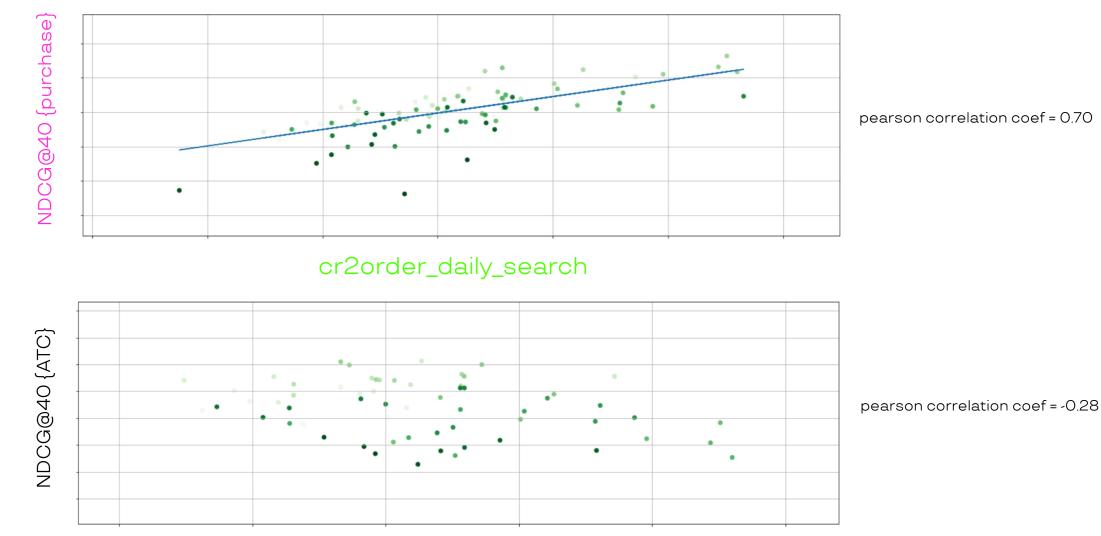




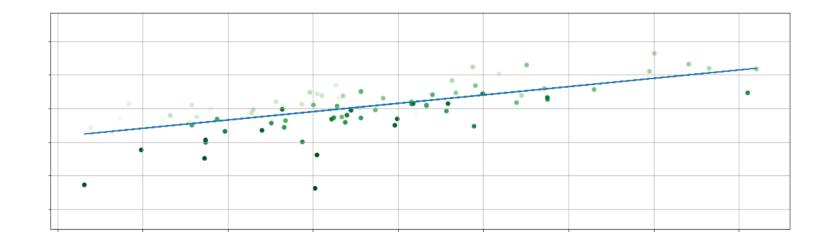




## Offline metric & online metric correlation





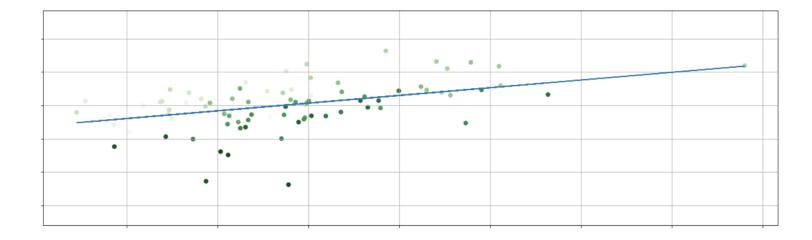


# ARPU daily search & NDCG@40{purchase}

pearson correlation coef = 0.69

#### ARPPU daily search δ NDCG@40{purchase}

pearson correlation coef = 0.49

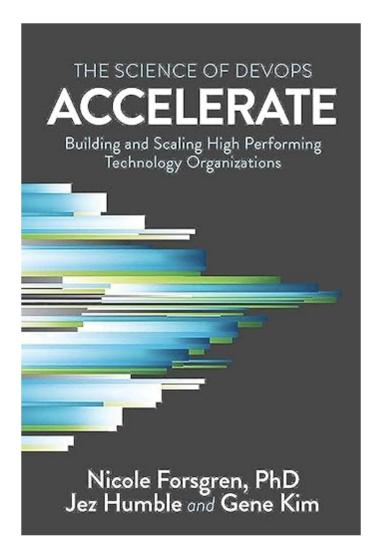




# 2. Fast iterations

Engineering (Data preparation, feature engineering, model training...)







# 3. High chances of success

୍ଚ Hypothesis sourcing & prioritization



## **SOURCES FOR HYPOTHESES**

#### **EXTERNAL**

- Talk to other companies
- Participate in meetups / conferences
- Monitor new publications

#### INTERNAL

- Analyse frequent problems
- Regularly organize brainstorms with your engineers



# ANALYZE FREQUENT PROBLEMS

Every month we collect queries with the **lowest conversion rates** (with potential problems) **among top-10000 most frequent queries**. Then we ask dedicated assessors to specify what is the problem:

- typo
- incorrect keyboard layout
- transliteration
- incomplete query
- synonym
- the request is too specific
- ranking problem
- assortment problem



# ICE – IMPACT, CONFIDENCE, EASE

Feature category	Ease of data collection	Query-time usage in elasticsearch	Can be used in other projects	Expected impact on metrics	Is used by competitors
Price	2	2	1	2	1
Quality	2	1	1	0	0
Popularity	2	1	1	2	1
Popularity	2	1	1	2	1
Popularity	2	2	1	1	1
Tag	2	1	1	2	1
Price	2	1	1	2	1
Price	2	1	1	2	1
Popularity	2	1	1	1	1
Price	2	1	1	1	1
Rating	2	1	1	1	1
Rating	2	1	1	1	1
Rating	2	1	1	1	1
Popularity	1	1	1	2	1
Popularity	2	• 1	1	1	0
Popularity	1	1	1	2	0
Popularity	2	1	1	1	0
Quantity	2	1	1	1	0
Quantity	2	1	1	1	0
Price	2	0	0	2	0
Popularity	2	1	1	0	0
Rating	2	1	1	0	0
Rating	2	1	1	0	0
Quality	2	1	1	0	0
Quality	2	1	1	0	0
Tag	1	1	1	1	0
Tag	2	1	1	0	0
Popularity	2	-2	0	2	1



# 4. Stable results



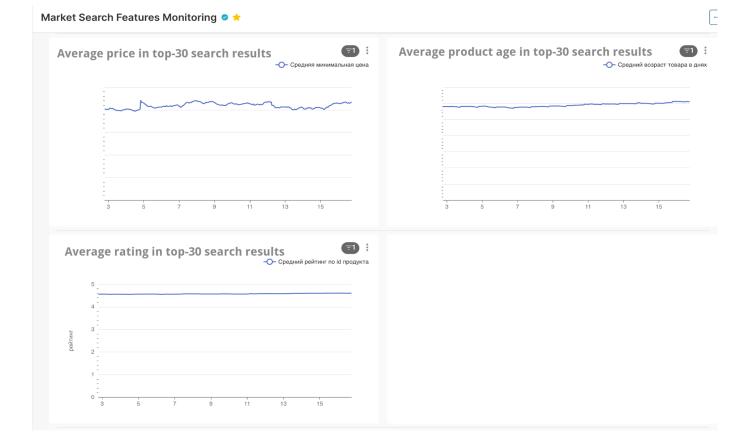


# THE MORE YOU MONITOR, THE BETTER

- Online metrics
- Ranking features distributions
- Clickstream events quality
- Airflow jobs failures









🛱 Cultivate a failure-tolerant culture.

Failure is part of the journey - embrace it.

TExtract knowledge from failures.

 $\bigcirc$  Every hypothesis fuels the growth.





# Q&A time

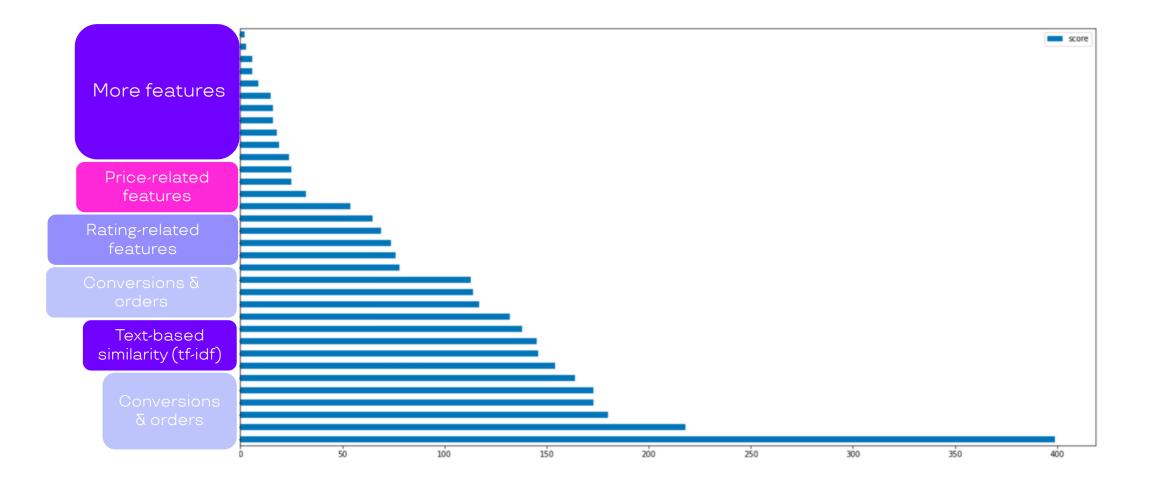
Andrey Kulagin

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tg: @and\_kul

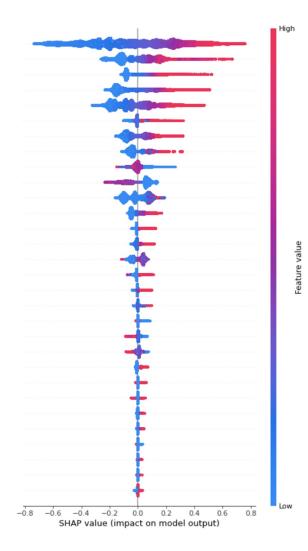


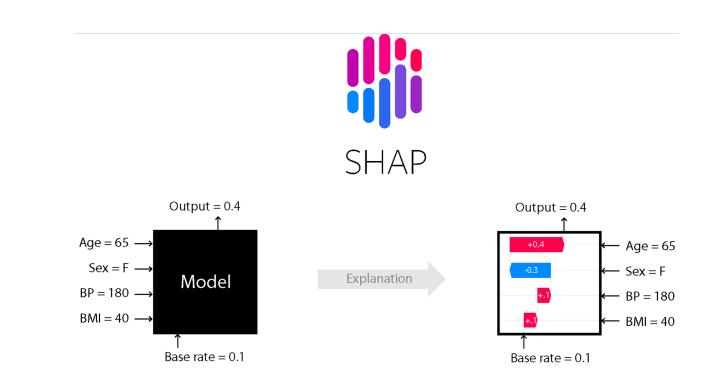
## **FEATURE IMPORTANCES**





## **SHAP VALUES**





https://shap.readthedocs.io/en/latest/

