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Ekstra  
Bladet

# RecSys' 24 Challenge Workshop Sponsor Edition

The 18th ACM Recommender Systems Conference  
Bari, Italy from Oct. 14–18, 2024.

Before we begin...

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# JP|Politikens Hus



# RecSys

[recsys.acm.org/recsys24](https://recsys.acm.org/recsys24)  
[recsys.eb.dk](https://recsys.eb.dk) & [ebnerd-codabench](https://ebnerd-codabench)



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DTU



Michael Riis Andersen  
DTU



Claudio Pomo  
Politecnico di Bari



OpenAI. (2024). Image created by ChatGPT. Retrieved from OpenAI platform.



# Call for Sponsorship RecSys Challenge 2025



The ACM Conference Series on  
**Recommender Systems**

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The [ACM Conference on Recommender Systems \(RecSys\)](#) is the premier **international forum** for presenting new research results, systems, and techniques in the broad field of recommender systems. **RecSys** brings together the major international research groups working on recommender systems and many of the world's leading companies active in e-commerce and other adjacent domains. It has become the most important annual conference for the presentation and discussion of recommender systems research.

**Every year, the RecSys Challenge has been a great success.** Participants from various backgrounds, disciplines, and backgrounds come together to present their innovative research and findings. The challenge provides scholars, researchers, students, and companies a platform to showcase their work, exchange ideas, and engage in meaningful discussions. The event fosters collaboration and networking opportunities, allowing participants to connect with peers who share similar research interests. **The challenge has consistently received positive feedback from presenters and attendees**, highlighting its significance in the academic community. It continues to be a valuable component of the conference, encouraging the advancement of knowledge and the exploration of new ideas.

Deadline: August 31st, 2024

[https://www.recsyschallenge.com/2024/assets/CfS\\_2025.pdf](https://www.recsyschallenge.com/2024/assets/CfS_2025.pdf)

# Agenda

- About Ekstra Bladet & JP/Politiken
  - Motivation for RecSys' 24 Challenge
  - Recommender Systems at Ekstra Bladet
  - Editorial Values and Alignment
- About RecSys' 24 Challenge
  - News Recommendation
  - EB-NeRD
  - Competition
- Open Q&A

# Our motivation for the RecSys 24 Challenge

**WAN/IFRA Study Tour, May 25th 2024**

Kasper Lindskow

Head of Research and Innovation

JP/Politikens Hus - Ekstra Bladet

PM of platform Intelligence in News

Copenhagen Business School – Management, Society and Communication

# Agenda

1. Who are we?
2. Recommender systems at Ekstra Bladet / JP/Politikens Hus
3. Editorial values and alignment



1.

Who are we?

# JP/Politikens Hus

DENMARK

**Ekstra Bladet**

**Jyllands-Posten**

**POLITIKEN**

Jyllands-Postens Lokalaviser Børnecaviser Monitormedier

**MarketWire** **WATCHMEDIER**

MEDICINSK TIDSSKRIFT

**BØRSEN.** **FINANS** Chapter

INFOMEDIA LASSO X **FORUM**

**NORTH AUDIO PUBLISHING** **POLITIKENS FORLAG** **P**

ARNOLDBUSCK.DK MENNE MUSEUM LINDBERG

**saxo** **bold.dk** **dao** BLADKOMPAGNIET

LÆS LYT LEV

UNITED KINGDOM

**de zeen**



NORWAY

INFOMEDIA

**FINANSWATCH** **MEDWATCH**

**EJENDOMSWATCH** **HANDELSWATCH**

KAGGE FORLAG **ADVOKATWATCH**

**ENERGIWATCH**



SWEDEN

BOKFÖRLAGET POLARIS **P** **ROOS TEGNER**

INFOMEDIA



GERMANY

**FINANZBUSINESS**

Herbert Frommes

**VERSICHERUNGSMONITOR**



INTERNATIONAL

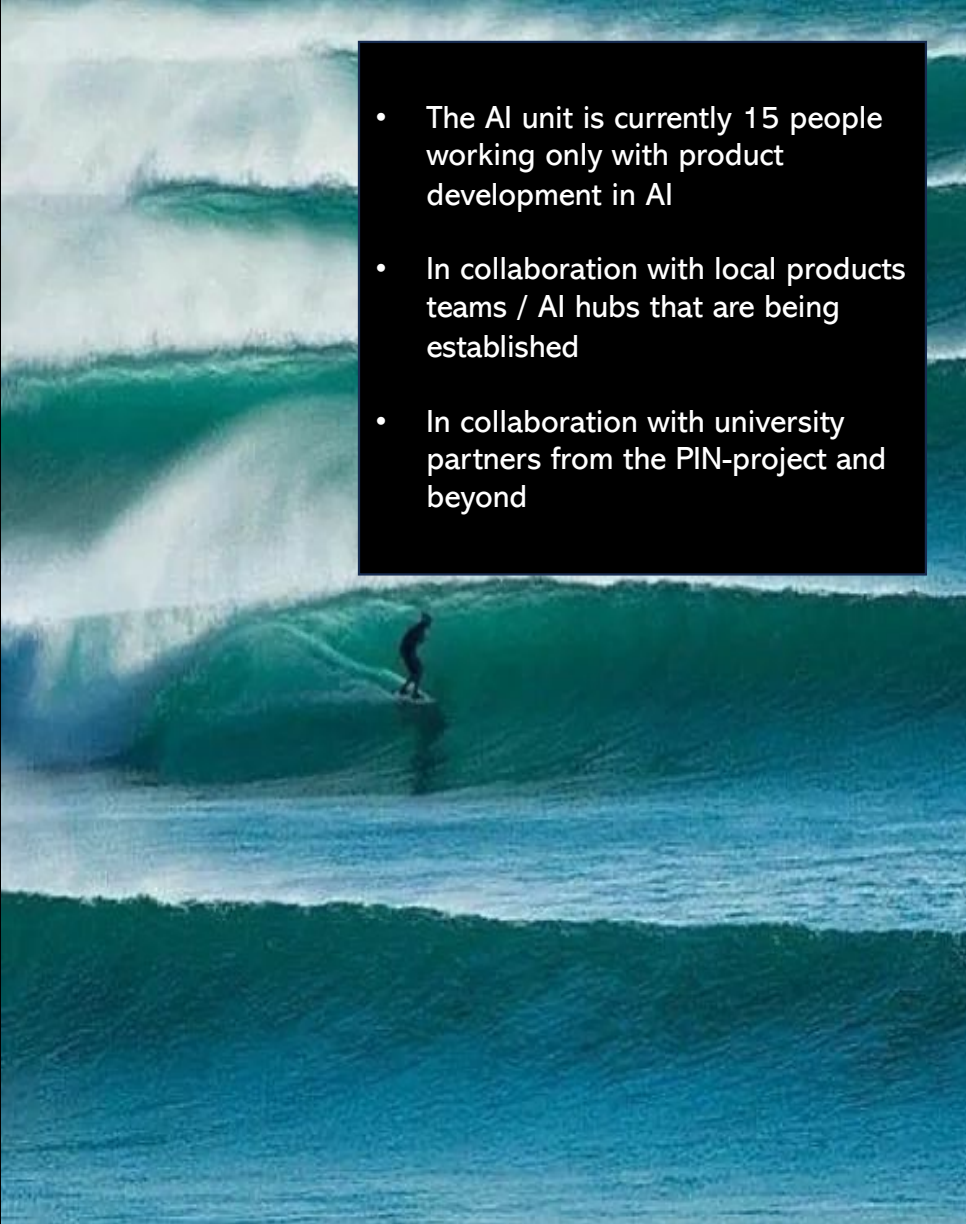
**MEDWATCH** **ENERGIWATCH**

**SHIPPINGWATCH**



# Shared AI unit (2024->)

- We develop and operate **AI products** across JPPOL's media to **enable local value creation**
- We focus on **editorial applications** and the **news experience**.
- We have a clear focus on **value-creating applications** that in some areas are supported by **applied research**.
- We act as **knowledge center** and organizer of **internal and external events** (e.g. the Nordic AI in Media Summit)

- 
- The AI unit is currently 15 people working only with product development in AI
  - In collaboration with local products teams / AI hubs that are being established
  - In collaboration with university partners from the PIN-project and beyond

# The AI unit's **Vision**

*We must drive a **paradigm shift in the way news is produced and consumed** by enabling our brands to effectively and responsibly utilize AI in the editorial workflow and in the news experience in ways **that advance the individual brand's editorial mission and strategy.***

# Our primary product areas

## 1. Generative AI

Editorial tools that assist with **routine tasks** and support **richer news coverage** in text, image, and sound.

Primary product:  
**MAGNA**

## 2. Metadata

Rich metadata about our news content in text, image, and sound, enabling the **activation of content** in new ways and gaining **deep insight** into how it is consumed.

Primary product:  
Name **t.b.d.**

## 3. Recommender systems

**Personalization of the news experience** for the individual in text, image, and sound, which can be **editorially controlled** based on insights into the personalized news streams.

Primary product:  
**Newspick**

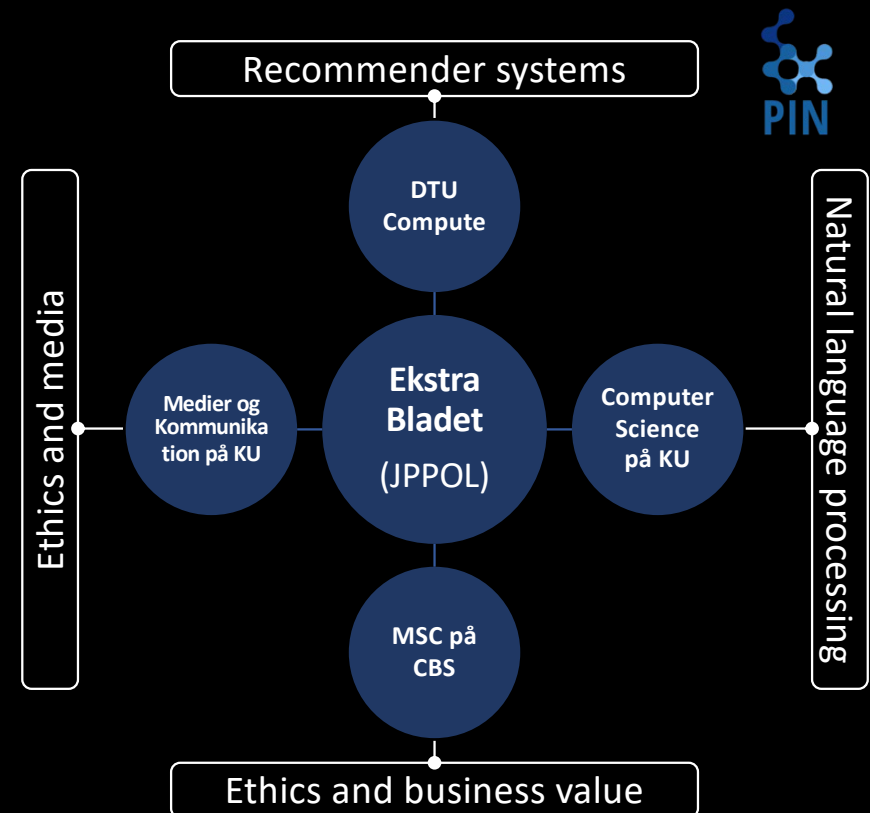
# 2.

Recommender systems at Ekstra Bladet / JP/Politikens Hus



# Goals for AI at Ekstra Bladet (2020)

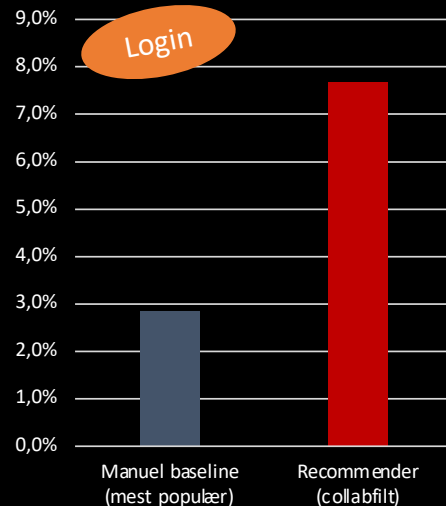
- A more relevant, engaging and information news experience ("wider, deeper, and richer")
- AI systems that are aligned with the values of news publishers and independent of the tech giants
- Contribute to a healthy norm setting for AI in media



# Online A/B tests



Free news (CTR)



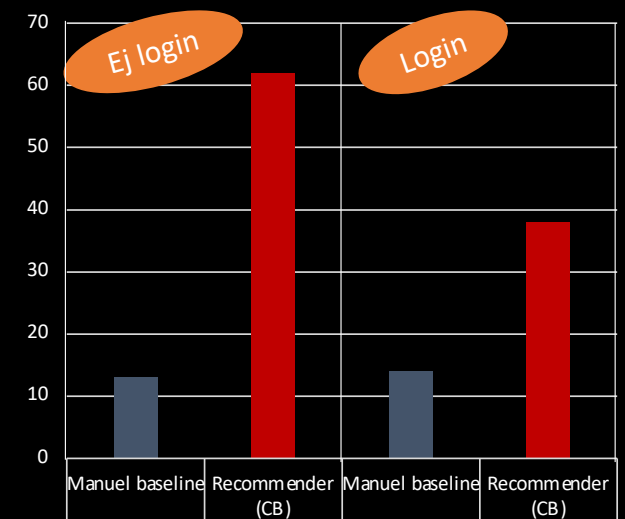
**Faktor 2,7  
bedre med AI**  
(A/B test)

Paid content (CTR)



**Faktor 2,4-3,1  
bedre med AI**  
(A/B test)

Paid content (sales)



**Faktor 2,7-4,8  
bedre med AI**  
(A/B test)

# Implementation on ekstrabladet.dk

- The common news flow is supplemented with recommenders at all horizontal placements.
- Creates a broader news flow because the recommenders supplement with relevant content for the individual.
- The systems have increased free traffic (+110%), use of paid content (+38%), and sales of subs (+35%) at the horizontal placements.



*Different recommender systems suggest personalized articles to various reader segments on the front page of ekstrabladet.dk.*

# We are happy – but we know we can do better

25. 10/10/2023

**Updated:** May 24, 2024.

To be added to the Academic Teams' leaderboard, please fill out the Google Form: [Academic Leaderboard](#).

*Search for multiple teams at once using commas (e.g., "Team1, Team2, ..., TeamN").*

Best AUC per Team

Academic Teams

| Rank | Team            | AUC    | MRR    | NDCG@5 | NDCG@10 |
|------|-----------------|--------|--------|--------|---------|
| 1    | taksai          | 0.8336 | 0.6409 | 0.7072 | 0.7257  |
| 5    | carlosgaravatti | 0.8221 | 0.6195 | 0.6879 | 0.7087  |
| 6    | asato           | 0.8211 | 0.6262 | 0.6907 | 0.7118  |
| 7    | doubleq         | 0.8201 | 0.6178 | 0.6870 | 0.7076  |

3.

Editorial values and alignment

# Why do we have all those beyond accuracy metrics in Codabench? e.g:

*An overview of key insights for Top@5 rankings.*

Overview:

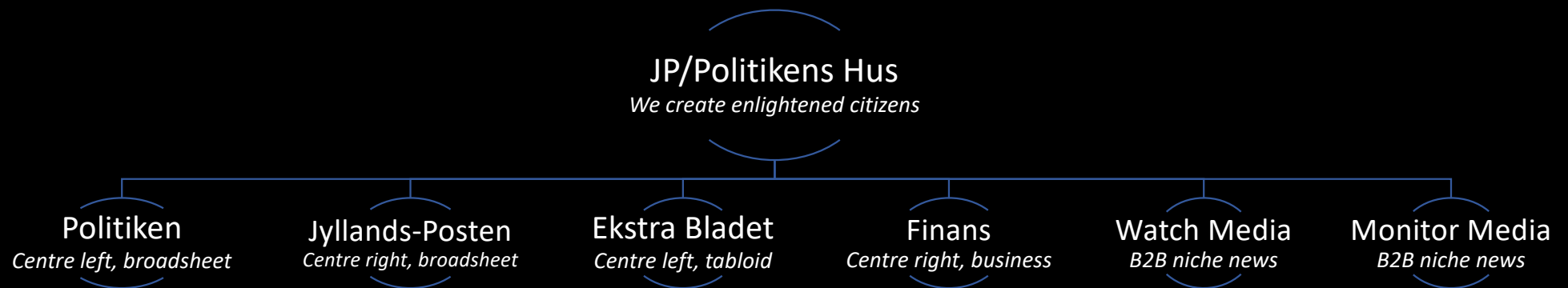
|                     | INTRALIST_DIVERSITY         | NOVELTY                     | SERENDIPITY                 | COVERAGE_FRACTION |
|---------------------|-----------------------------|-----------------------------|-----------------------------|-------------------|
| YOUR_MODEL          | 0.7178 $\hat{A} \pm 0.0841$ | 9.0915 $\hat{A} \pm 0.9842$ | 0.7747 $\hat{A} \pm 0.0416$ | 0.308             |
| TOP-INVIEW-ARTICLES | 0.7905 $\hat{A} \pm 0.0$    | 4.6258 $\hat{A} \pm 0.0$    | 0.7861 $\hat{A} \pm 0.0262$ | 0.02              |
| POPULAR             | 0.8402 $\hat{A} \pm 0.0$    | 3.0699 $\hat{A} \pm 0.0$    | 0.7915 $\hat{A} \pm 0.0263$ | 0.02              |
| RANDOM              | 0.7548 $\hat{A} \pm 0.0902$ | 8.3617 $\hat{A} \pm 1.8496$ | 0.8068 $\hat{A} \pm 0.0337$ | 1.0               |

Overview: distribution\_sentiment:

|                     | NEGATIVE | NEUTRAL | POSITIVE |
|---------------------|----------|---------|----------|
| YOUR_MODEL          | 0.4841   | 0.3988  | 0.1170   |
| TOP-INVIEW-ARTICLES | 0.4000   | 0.0000  | 0.6000   |
| POPULAR             | 0.2000   | 0.4000  | 0.4000   |
| RANDOM              | 0.3962   | 0.2917  | 0.3121   |



# The missions of our news brands

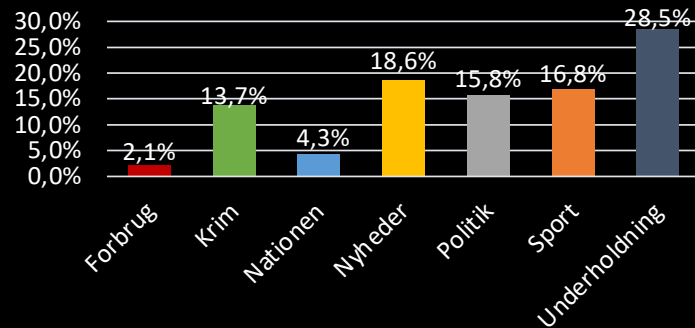


- We need to make sure our recommenders are effective
- We need to make sure our recommenders are aligned with the individual news brands

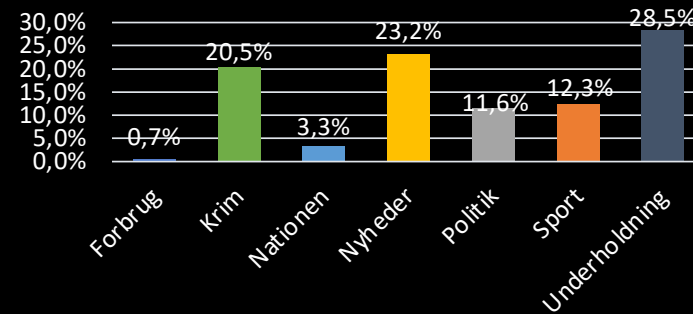
# Recommender system effects on the news flow (I)

Ekstra  
Bladet

**Collaborative filtering (CTR: 6%)**



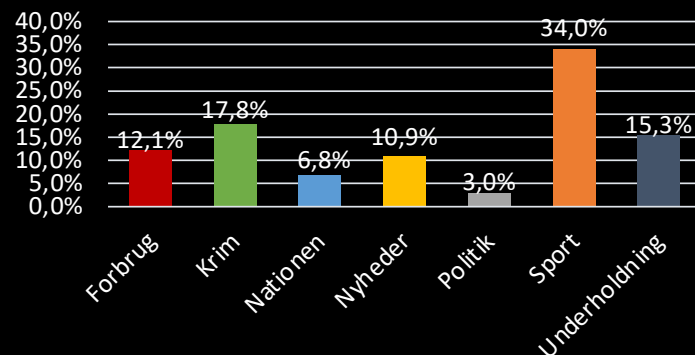
**Most popular (CTR: 2,7%)**



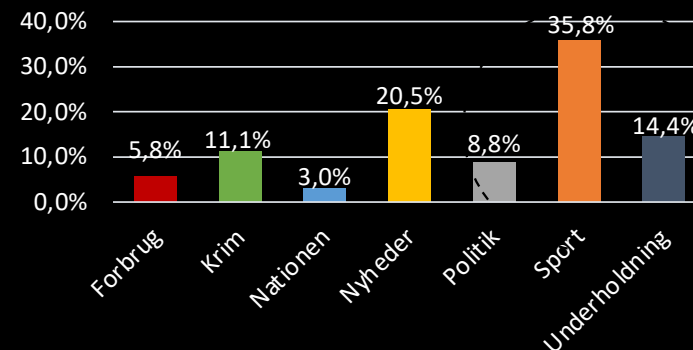
Popularity  
bias

Same  
effect on  
news flow!

**Content filtering (CTR: 6,1%)**



**Randomly selected (CTR: 2,4%)**



lowserendip  
ity

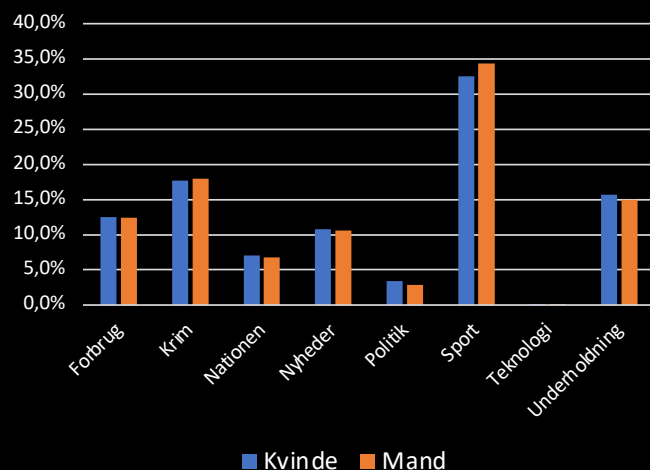
Same  
effect on  
news flow!

Different effect on news flow!

# Recommender system effects on the news flow (II)

## Hybrid recommender

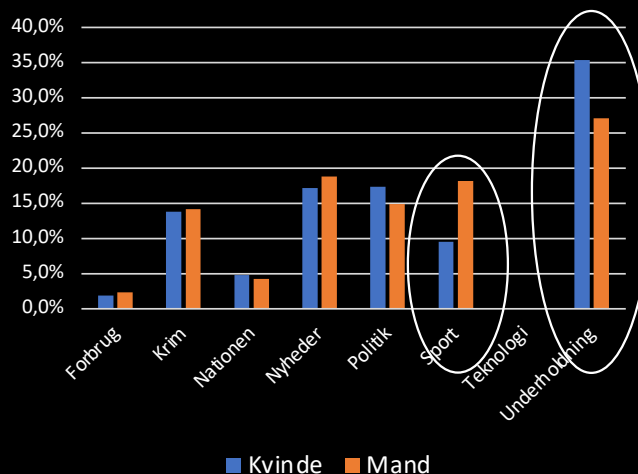
Fordeling af anbefalede artikler



**Ingen kønsforskelle**

## Collaborative recommender

Fordeling af anbefalede artikler

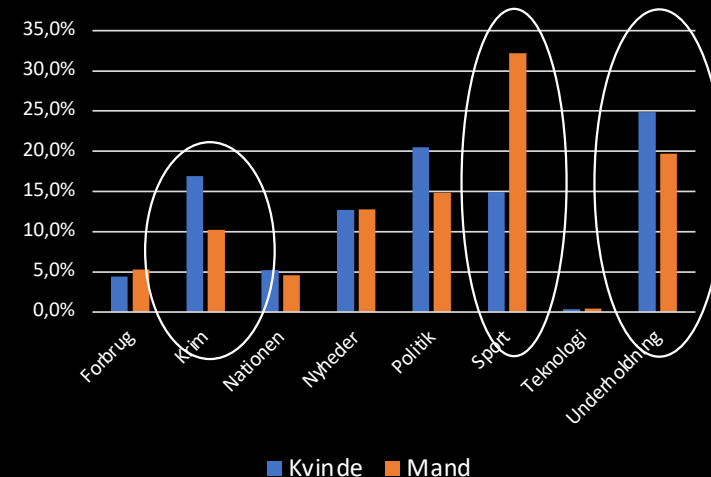


**Små kønsforskelle**

*Mere sport til mænd, mere underholdning til kvinder*

## Content based recommender

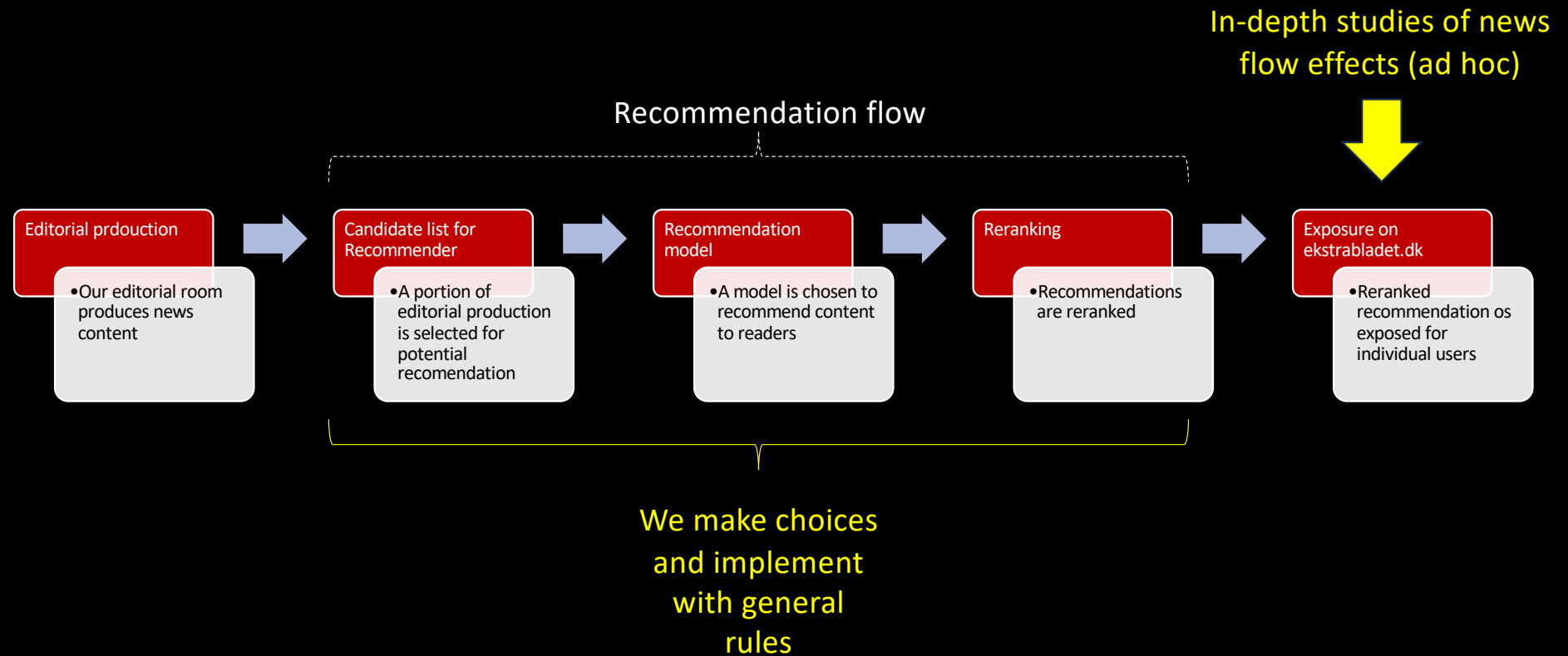
Fordeling af anbefalede artikler



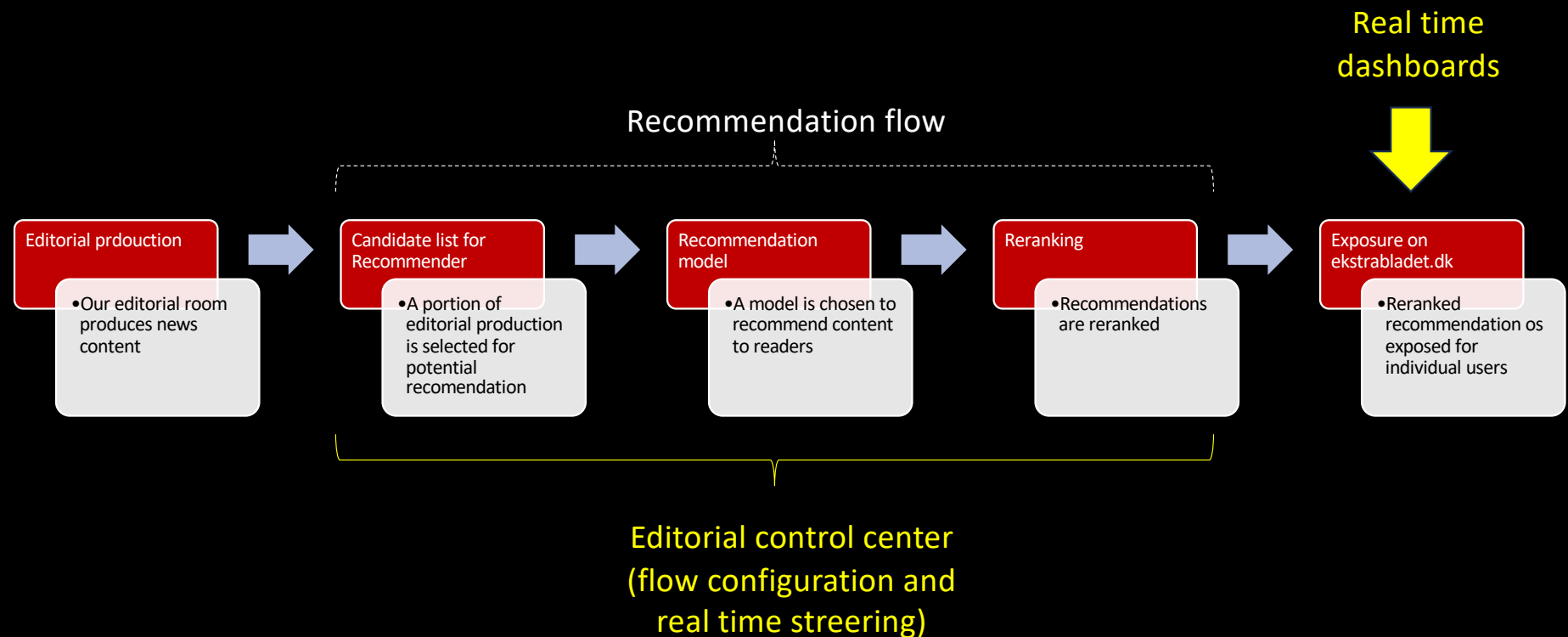
**Store kønsforskelle**

*Langt mere sport til mænd, mere underholdning og kriminalitet til kvinder*

# Editorial alignment of recommender systems (today)



# Editorial alignment of recommender systems (to be)



# We invite you to reflect on beyond accuracy effects of your recommendation methods

*An overview of key insights for Top@5 rankings.*

Overview:

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| RANDOM              | 0.3962   | 0.2917  | 0.3121   |



# Thank you!

Kasper Lindskow

Head of AI

JP/Politikens Media Group

Reach out at [kasper.lindskow@jppol.dk](mailto:kasper.lindskow@jppol.dk) or on LinkedIn (Kasper Lindskow)

# RecSys' 24 Challenge

## Agenda

- News Recommendation
- EB-NeRD: *Ekstra Bladet's News Recommendation Dataset*
- Competition
- Q&A

# News Recommendation

Predict which articles a user clicked on from a list of articles using the user's click history, session details, and personal metadata.



## TECNICAL DIFFICULTIES PLEASE STAND BY

- 1) The **fast pace** of news and **rapid decay** of article relevance
- 2) The **shifts in readers interests** with the news agenda
- 3) The need to model interests based on **implicit feedback**
- 4) Lack of support for **low-resource languages**
- 5) The need to **align AI systems with editorial values (normative thinking)**, incl. avoidance of bias and filter bubbles

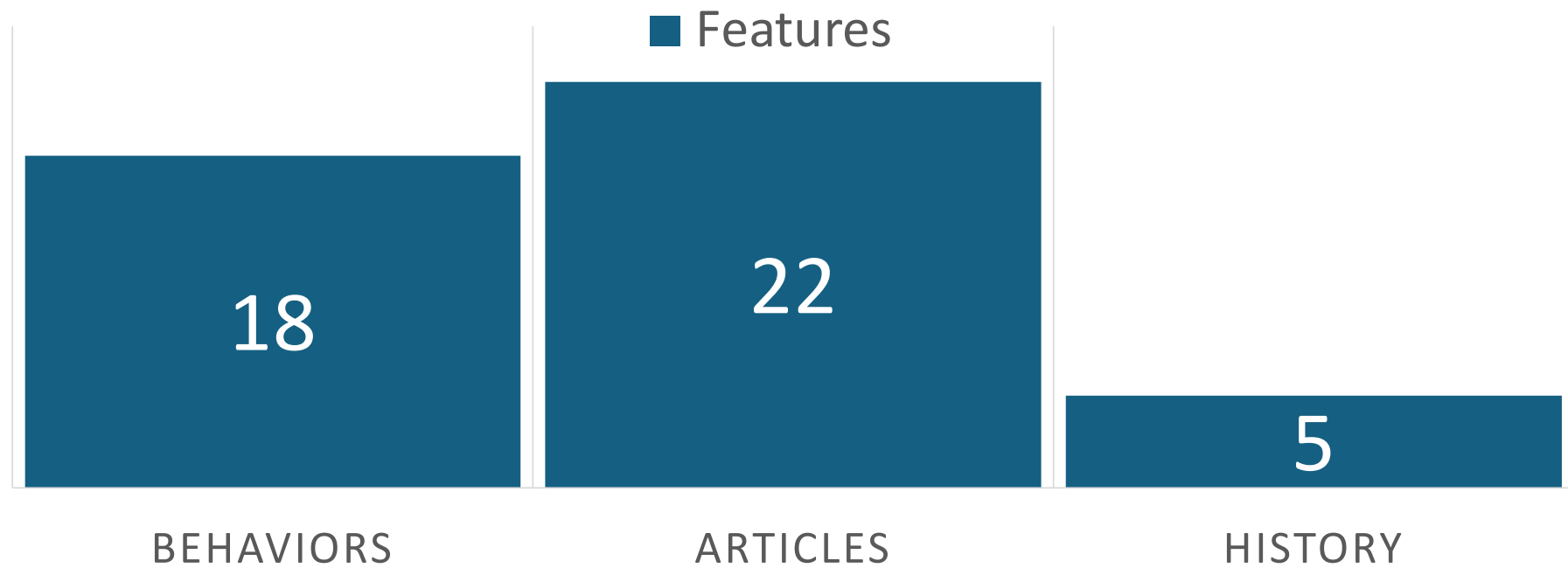
# Ekstra Bladet News Recommendation Dataset

## About

The **Ekstra Bladet News Recommendation Dataset** (EB-NeRD) was created to support advancements in news recommendation research. It was collected from user behavior logs at [Ekstra Bladet](#). We collected behavior logs from active users during the 6 weeks from April 27 to June 8, 2023. This timeframe was selected to avoid major events, e.g., holidays or elections, that could trigger atypical behavior at Ekstra Bladet. The active users were defined as users who had at least 5 and at most 1,000 news click records in a three-week period from May 18 to June 8, 2023. To protect user privacy, every user was delinked from the production system when securely hashed into an anonymized ID using one-time [salt mapping](#). Alongside, we provide Danish news articles published by Ekstra Bladet. Each article is enriched with textual context features such as title, abstract, body, categories, among others. Furthermore, we provide features that have been generated by proprietary models, including topics, named entity recognition (NER), and article embeddings

[recsys.eb.dk/dataset/](https://recsys.eb.dk/dataset/)

## Dataset with Extensive Features

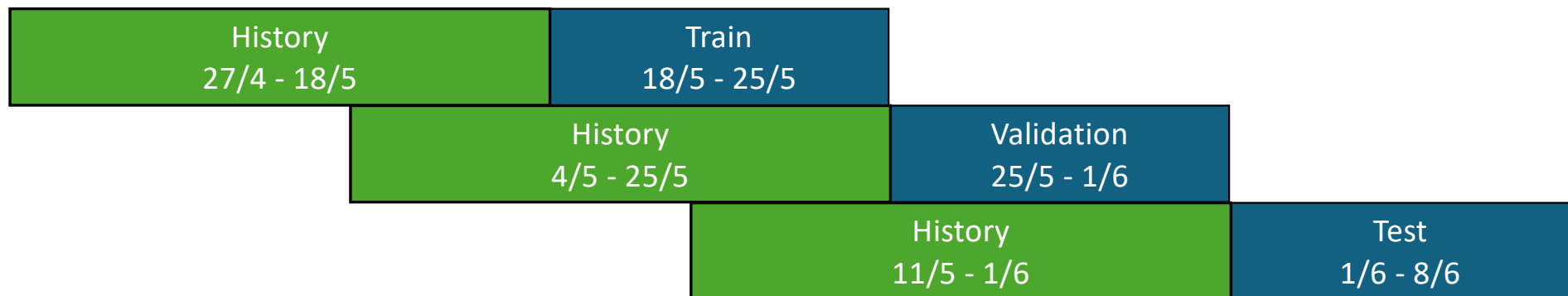


\*Artifacts: document-embeddings and image-embeddings



# EB-NeRD in short

- 👤 +2.3 million users
- 💻 +380 million impression logs from Ekstra Bladet
- 📰 +125,000 news articles
- Collection period: 6 weeks from April 27 to June 8, 2023
- Datasets: train, validation, and test
  - Each split is **7 days of behavior logs** + **21 days of clicks history**



# Competition

- Trainingset, Validationset, and Testset (removed labels, next-readtime, next-scroll percentage)
- Features / dataset utilization
  - **You are allowed to use ALL the open-source data**
  - *Total Inviews & Total Pageviews* (can be estimated even without us provided them)
  - Prohibiting features, preprocessing, etc. is extreme hard to enforce
  - Encourage you to share findings to everyone; interesting for participants, organizers, and future competitions
  - If you use these features or use the dataset in unorthodox ways you have report the results with and without for your paper submission
- **The primary metric for the challenge is Area under the ROC curve (AUC)**

# Evaluation

- **Classification** (AUC), **Ranking** (Mean Reciprocal Rank, Normalized Discounted Cumulative Gain), and **Beyond Accuracy** (Diversity, Coverage, Serendipity, Novelty)
- Beyond-accuracy: 200.000 syntetic samples
  - All have the same article inviews, impression-ID, and timestamp
  - We created a candidate list – you can do this as well
  - Baselines: [make\\_beyond\\_accuracy.ipynb](#)
- **Don't shuffle the testset!**
  - Test-script splits evaluate classification/ranking and beyond-accuracy samples

# Leaderboards

| Date             | AUC* ▼ | MRR*   | NDCG5* | NDCG10* | Detailed Results  |
|------------------|--------|--------|--------|---------|---|
| 2024-04-10 09:11 | 0.7251 | 0.4928 | 0.5586 | 0.602   |  |
| 2024-04-09 01:55 | 0.7246 | 0.4974 | 0.5613 | 0.6041  |  |
| 2024-04-12 02:33 | 0.7237 | 0.4924 | 0.5574 | 0.601   |  |
| 2024-04-09 07:38 | 0.7077 | 0.4755 | 0.5343 | 0.5802  |  |
| 2024-04-08 11:28 | 0.7064 | 0.4722 | 0.5322 | 0.5783  |  |
| 2024-04-07 07:23 | 0.6977 | 0.4591 | 0.5205 | 0.5679  |  |

\*: 50% of the testset

## An overview of key insights for Top@5 rankings.

### Overview:

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Updated: May 24, 2024.

To be added to the Academic Teams' leaderboard, please fill out the Google Form: [Academic Leaderboard](#).

Search for multiple teams at once using commas (e.g., "Team1, Team2, ..., TeamN").

| Rank | Team            | AUC    | MRR    | NDCG@5 | NDCG@10 |
|------|-----------------|--------|--------|--------|---------|
| 1    | taksai          | 0.8336 | 0.6409 | 0.7072 | 0.7257  |
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| 6    | asato           | 0.8211 | 0.6262 | 0.6907 | 0.7118  |
| 7    | doubleq         | 0.8201 | 0.6178 | 0.6870 | 0.7076  |

# Academic leaderboard



**\$3,500, \$2,500, \$1,500 + \$2,500**  
(Best Academic Team)

Updated: May 24, 2024.

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Search for multiple teams at once using commas (e.g., "Team1, Team2, ..., TeamN").

Search by Team names

Best AUC per Team

Academic Teams

| Rank | Team            | AUC    | MRR    | NDCG@5 | NDCG@10 |
|------|-----------------|--------|--------|--------|---------|
| 5    | carlosgaravatti | 0.8221 | 0.6195 | 0.6879 | 0.7087  |
| 13   | andrealari      | 0.8153 | 0.6076 | 0.6779 | 0.6998  |
| 235  | testuser123     | 0.5970 | 0.3774 | 0.4237 | 0.4965  |
| 295  | imane           | 0.4998 | 0.3157 | 0.3490 | 0.4339  |

[Academic Leaderboard Form](#)

## Academic Leaderboard

To be added to the academic leaderboard, please add yourself here.

**Important: You need to use the email and name on Codabench.**

johanneskrse@gmail.com [Switch account](#)



\* Indicates required question

Email \*

Your email

Name on the Codabench leaderboard \*

Your answer

Submit

Clear form

# Call for papers (TBA)

The topics of interest include, but are not limited to:

- Benchmarking and evaluation of recommender systems on EB-NeRD
- Novel model architectures for news recommendation
- Dataset analyses and preprocessing techniques
- Contributions focused on beyond accuracy, such as fairness, diversity, coverage, etc.
- Scalability and efficiency of recommendation algorithms
- Cross-domain and multi-modal recommendations





Q&A

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