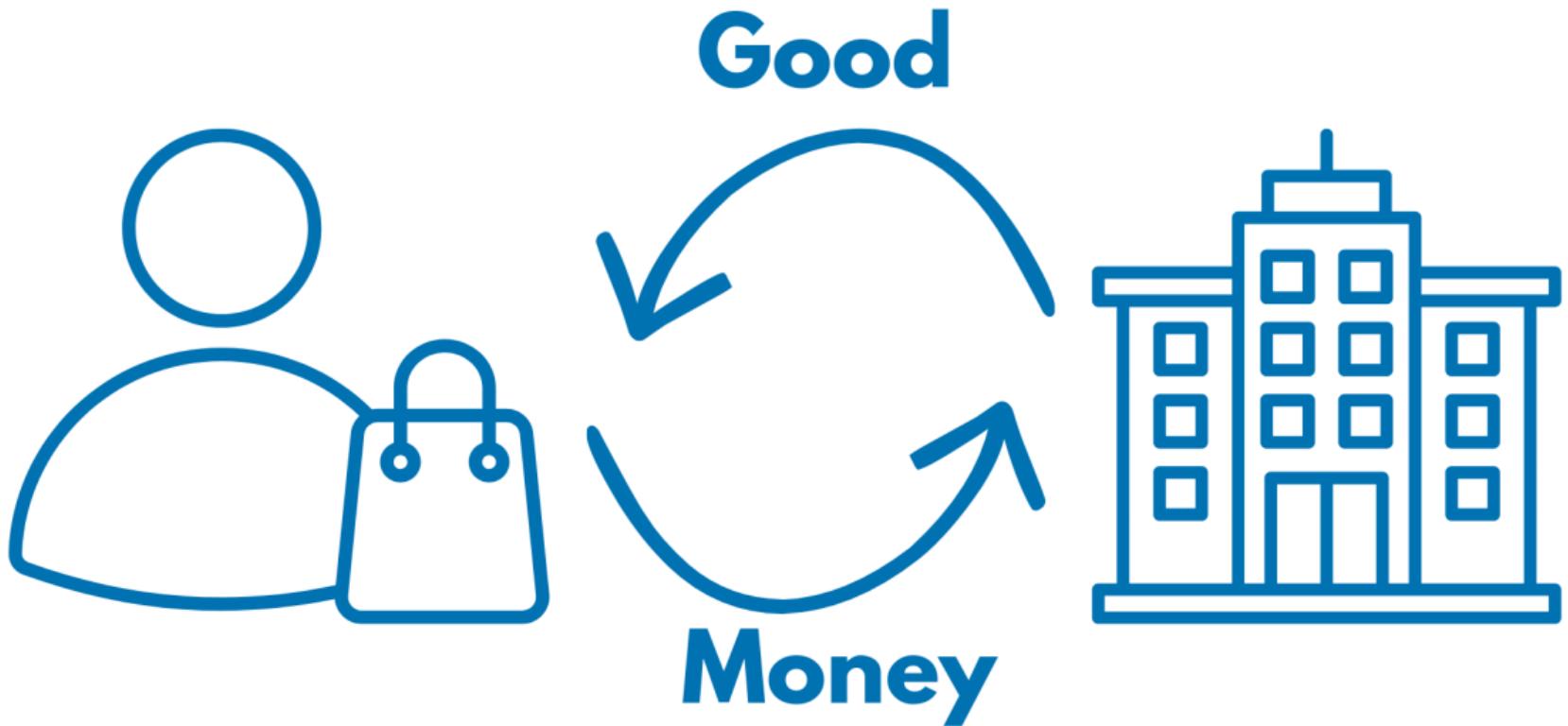
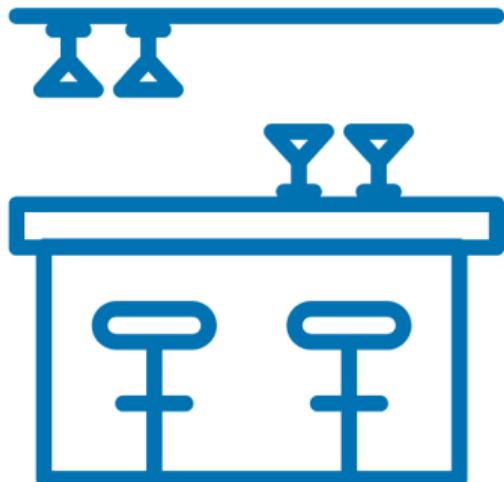
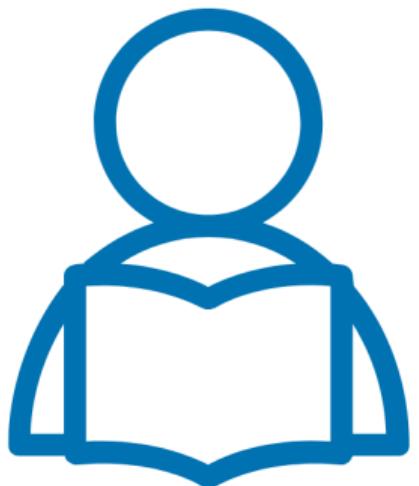


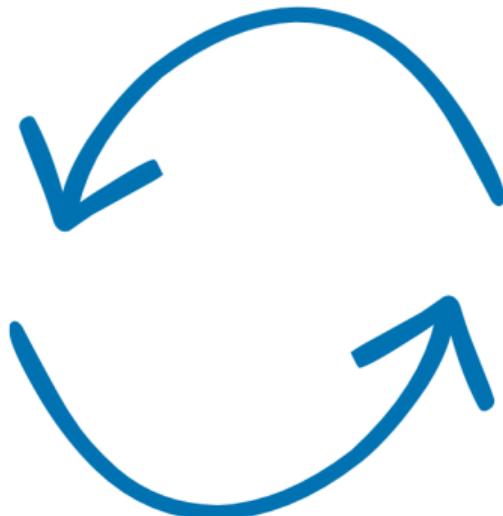
Will AI change the internet?



Beer



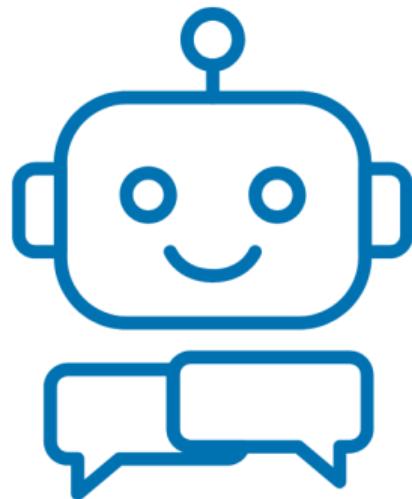
Information



Attention

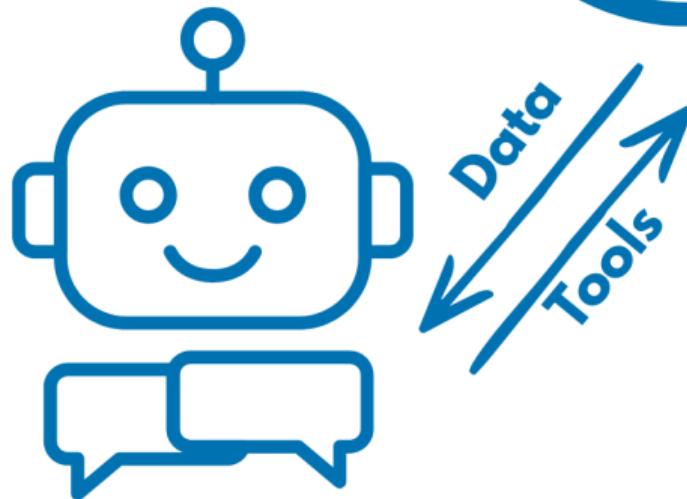


Information
↔
Attention

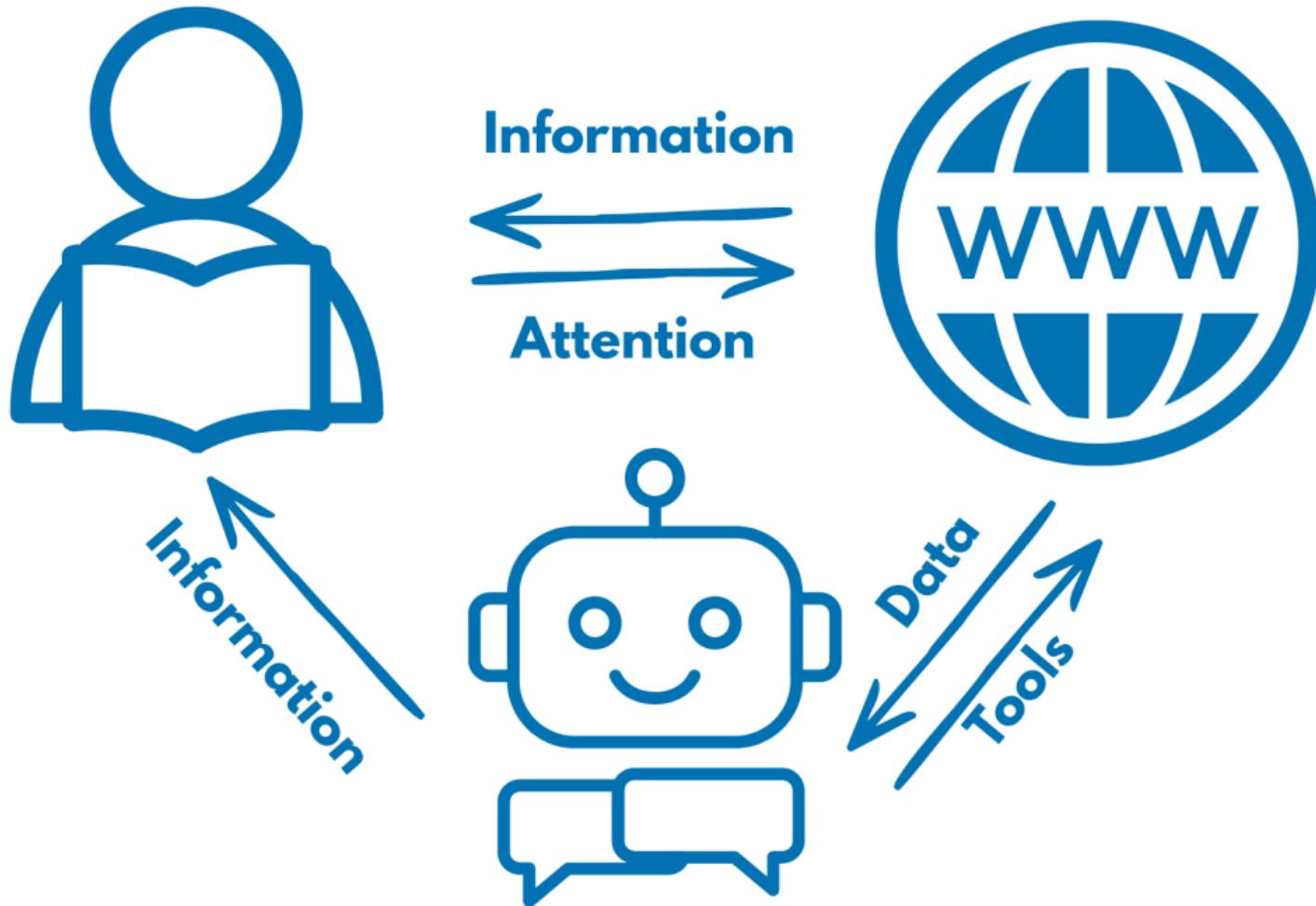


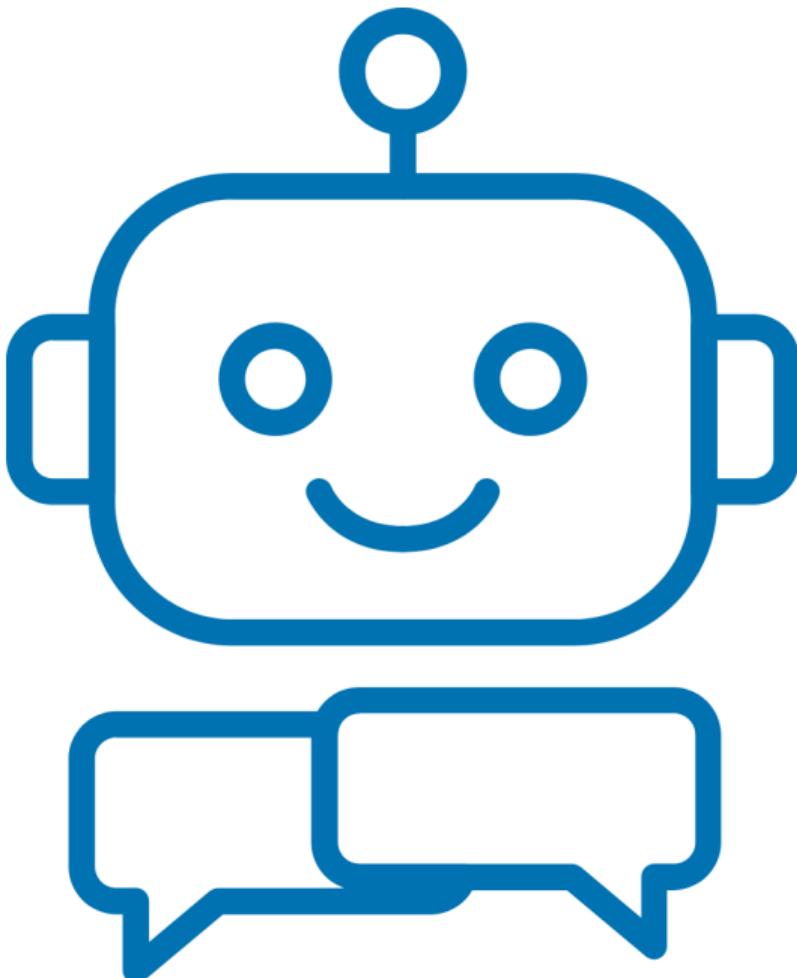


Information
↔
Attention



Data
Tools





Effect 1: The AI steals customers, thus **less** incentive to produce **content**.

Effect 2: The AI makes it easier to produce, thus **more** **content**.

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- ▶ Preview: Gen AI makes things worse but not always.

► Gen AI's Impact Online

Burtch, Lee, and Chen 2024; Rio-Chanona, Laurentsyeva, and Wachs 2023; Reeves, Yin, and Simperl 2025; Shan and Qiu 2025; Lyu et al. 2025; Zhao and Berman 2025; Koren et al. 2026

Contribution: Structured, comprehensive and credible evidence of Gen AI's impact both on supply and demand.

► Discrete Choice & Monopolistic competition

Melitz 2003; Krugman et al. 1980; Train 2009 and many many many more

Contribution: Novel literature application to a new field: Online market with interest of AI.

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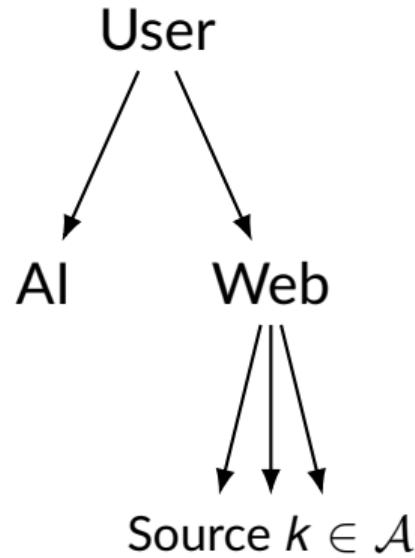
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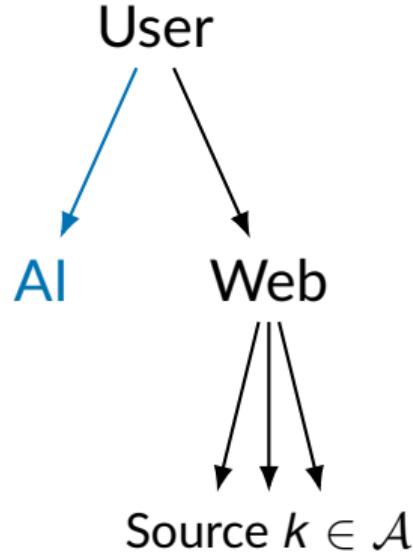
Timing of the model:

1. Providers draw quality of their information.
2. Given the quality decide whether to enter the market or not.
3. Users pick the sources among the providers that entered and GenAI

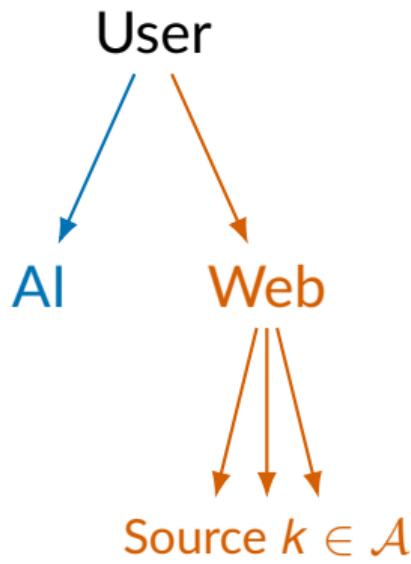


► Utility:

$$U_{iA} = \ln Q_A + \epsilon_{iA}$$



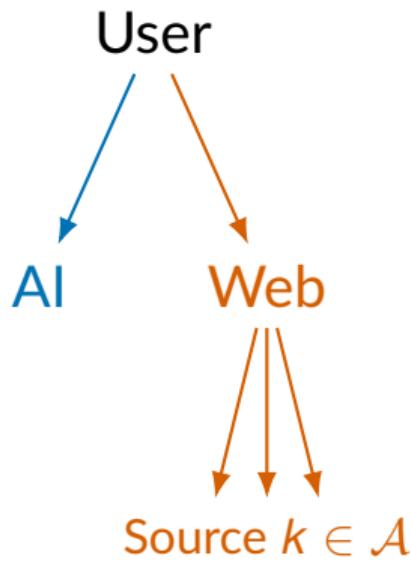
► Utility:



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$$\epsilon_i \sim GEV(\theta)$$



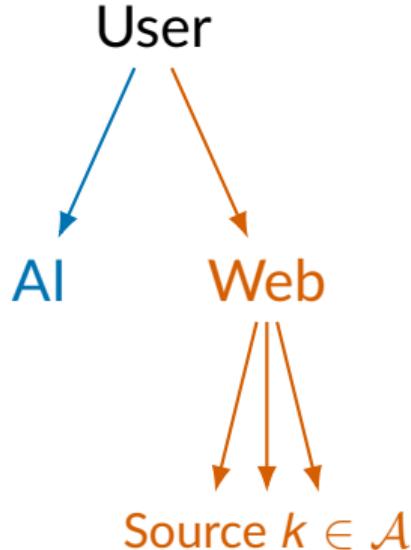
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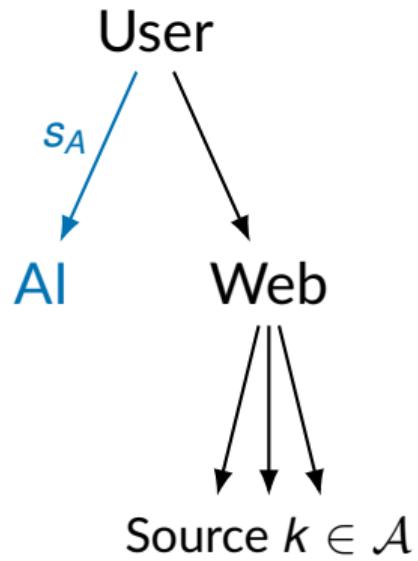
$$U_{ik} = \ln Q_k + \ln \delta_W + \epsilon_{ik}.$$

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- ▶ AI aggregates the information available on the web, with efficiency ϕ

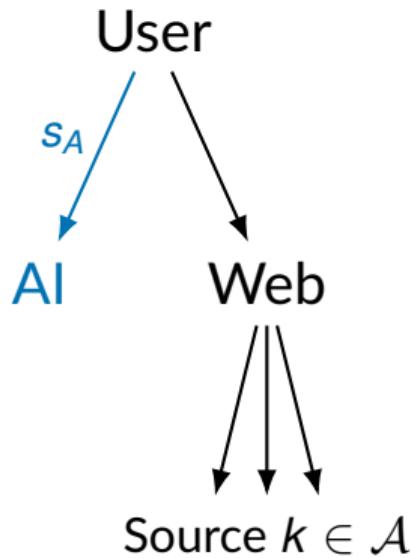
$$Q_A = \phi \left(\sum_k Q_k^{1/\theta} \right)^\theta$$

- ▶ The shares of users that go to:
 - ▶ AI



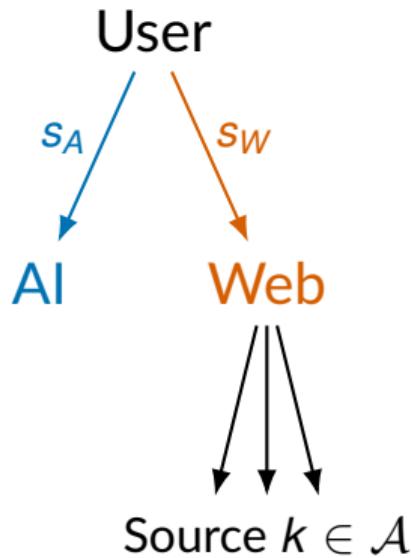
$$s_A = \frac{\phi}{\delta_w + \phi}$$

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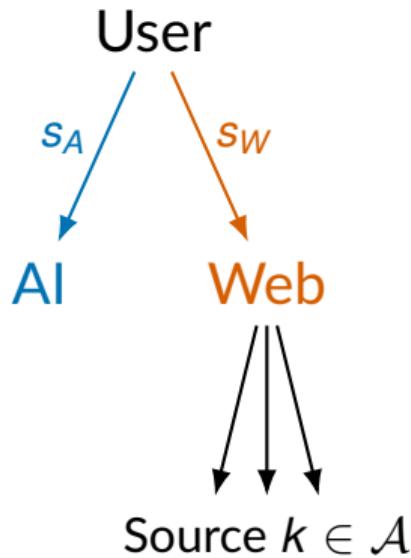


$$s_A = \frac{\phi}{\delta_w + \phi}, \quad \uparrow \phi \text{ and } \downarrow \delta_w.$$

- ▶ Web

$$s_W = \frac{\delta_w}{\delta_w + \phi}$$

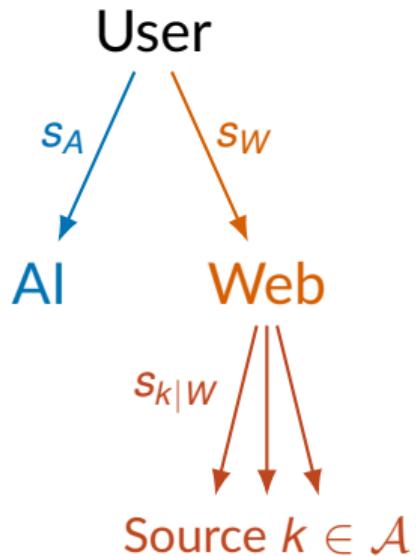
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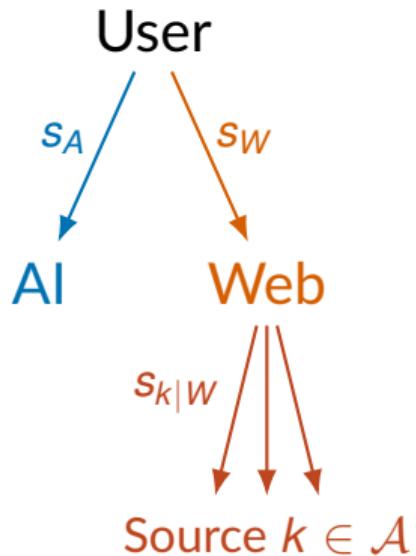
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$$s_{k|W} = \frac{Q_k^{1/\theta}}{\sum_j Q_j^{1/\theta}}$$



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Supply of information:

- ▶ **increases**↑ because AI decreases the cost of provision (η).
- ▶ **decreases**↓ because AI steals the consumers, thus there is less incentive to produce (ϕ).

- We care about the quality that is consumed by the users:

$$\tilde{Q} = s_A Q_A + \sum_{j \in A} s_j Q_j$$

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$$\begin{aligned}\tilde{Q} &= s_A Q_A + \sum_{j \in A} s_j Q_j \\ &= \frac{Q_0}{\delta_W + \phi} \left[\frac{\delta_W(\gamma - 1/\theta)}{\gamma - 1/\theta - 1} + \phi^2 \left(\frac{r\delta_W}{F(1-\eta)(\delta_W + \phi)} \right)^\theta \right]\end{aligned}$$

Issues

- ▶ DATA
 - ▶ GESIS Panel.dbd Prerelease
 - ▶ 6k german users, >1mil datapoints.
 - ▶ Captures user, time, URL, and HTML snapshots of visited pages.
 - ▶ Socio-economics stuff available.
- ▶ Production side without a production function.

Conclusions

1. How does Generative AI (GenAI) affect online content supply and quality?
2. Ideally Structural model, but for now theory.
3. When we reduce the cost of information provision, there is more information quality and supply.

Thank you!

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