

# Recommendation of content to mitigate the echo chamber effect

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## CONTRIBUTION

**Goal:** mitigate the echo chamber effect.

**Objective metric:** content diversity on newsfeeds.

**Method:** content recommendation.

**Application:** political Twitter dataset.

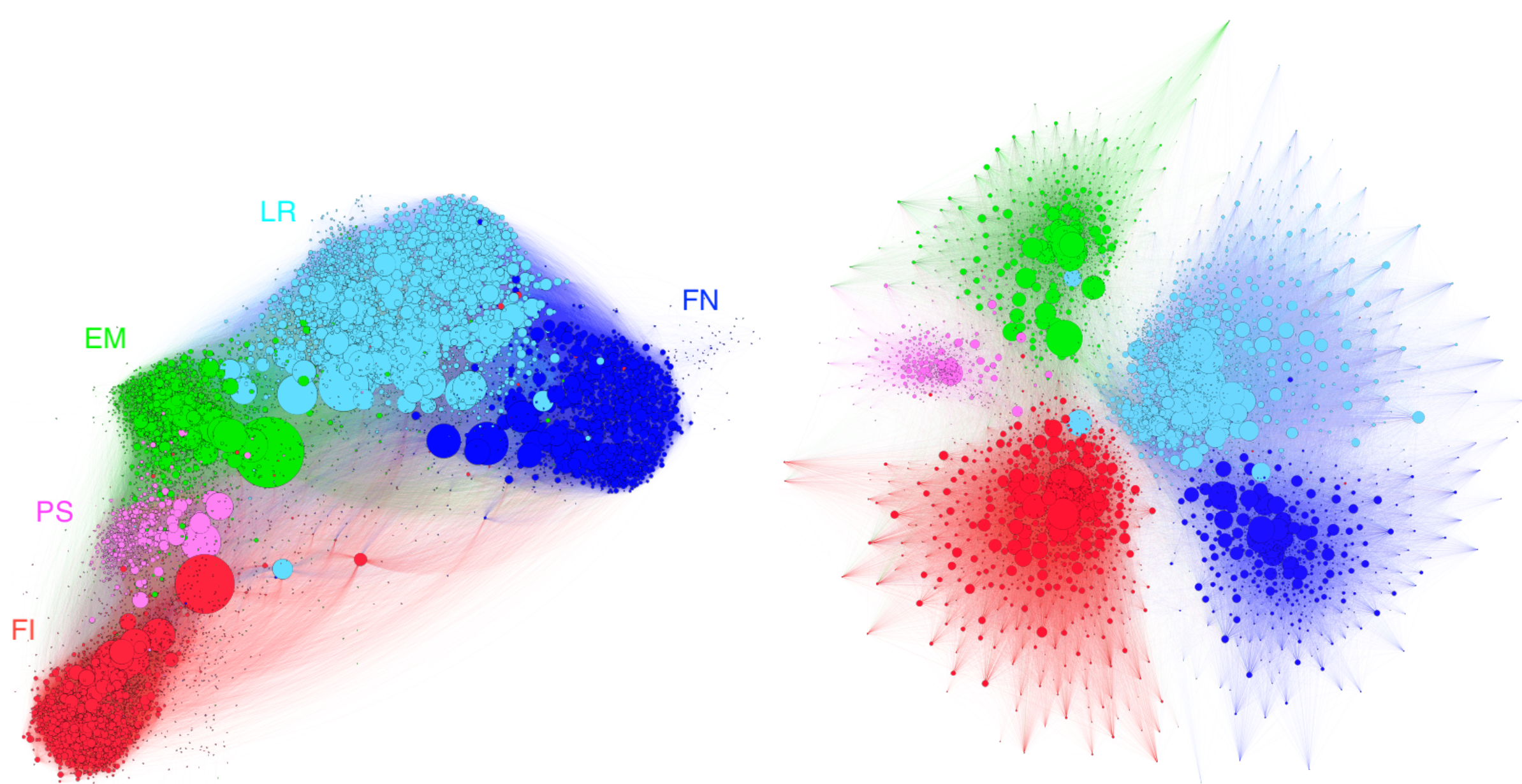
**Result:** significant improvement in content diversity.

**Side product:** extension of an existing diffusion model to describe opinion dynamics.

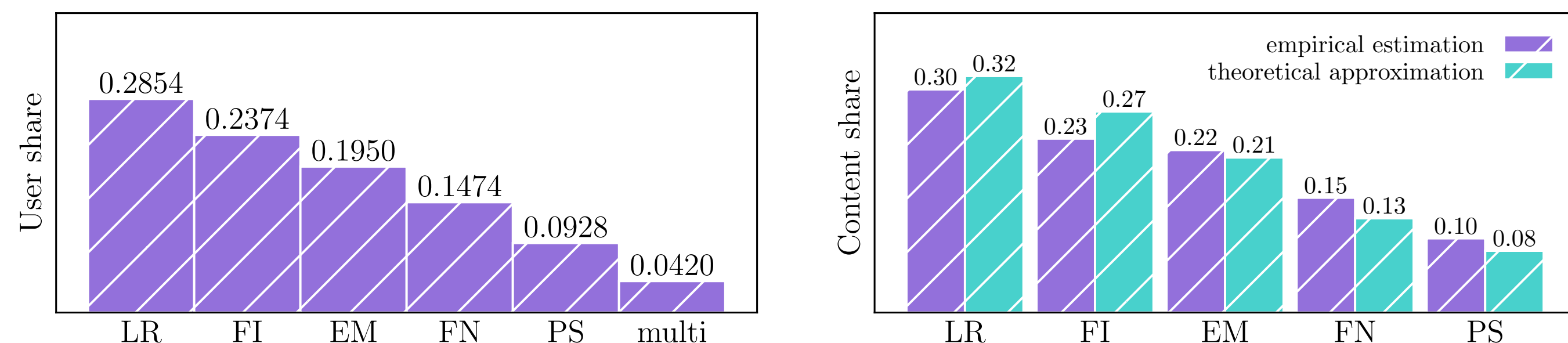
## #Elysee2017fr dataset

Twitter dataset re. 2017 French presidential elections.

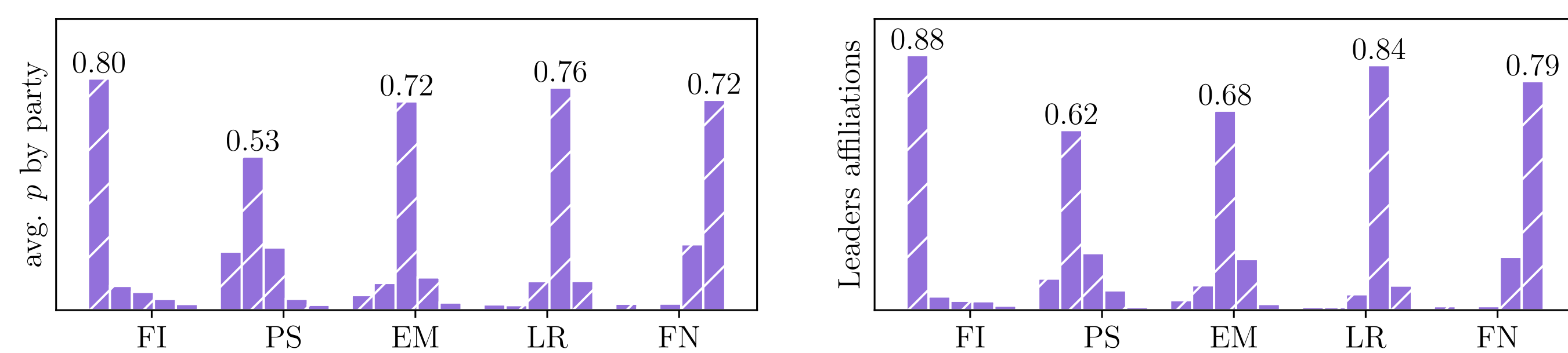
- 10 million tweets.
- $N = 8,277$  users,  $975K$  edges.
- $S = 5$  parties.
- Political affiliations of users: FI, PS, EM, LR, FN.



Follow (left) and Retweet (right) graph.



Share of users and content for each party.



Echo chambers and homophily.

## Model framework

- Strongly connected network of  $N$  users.
- User  $n$  creates posts supporting party  $s$  at rate  $\lambda_s^{(n)}$ .
- User  $n$  reposts from their newsfeed at rate  $\mu^{(n)}$ .

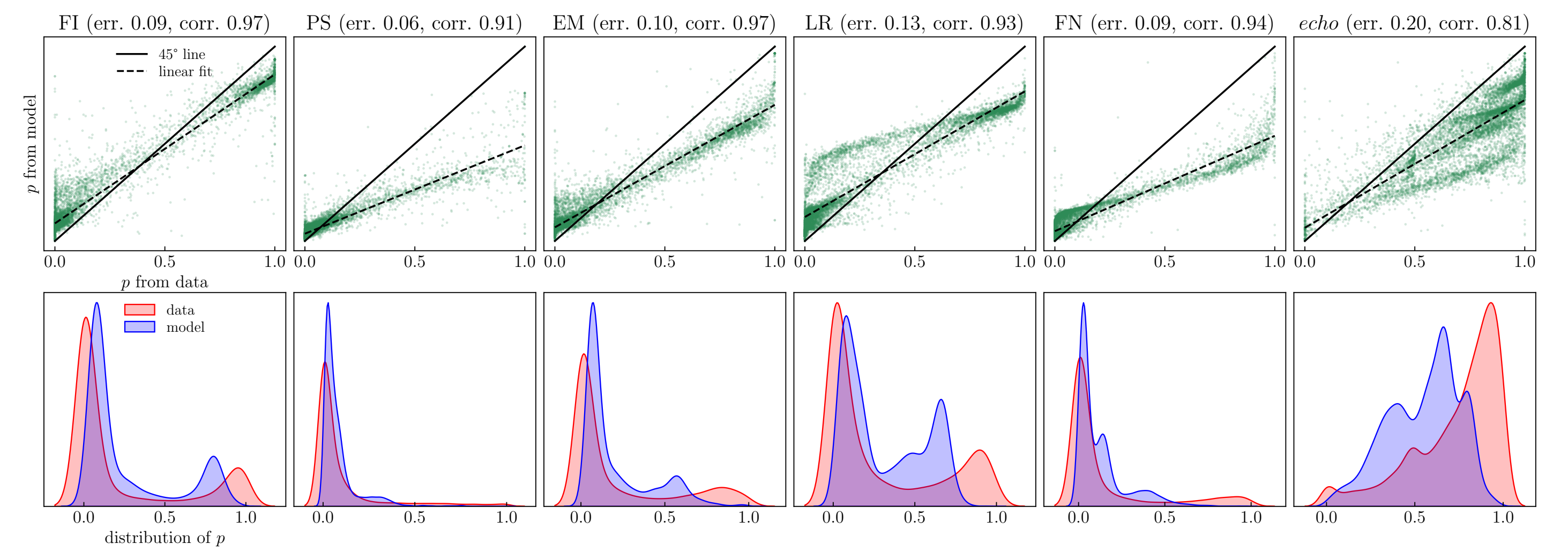
### Balance of opinions on newsfeeds

$$p_s^{(n)} \sum_{k \in \mathcal{L}^{(n)}} (\lambda^{(k)} + \mu^{(k)}) = \sum_{k \in \mathcal{L}^{(n)}} (\lambda_s^{(k)} + \mu^{(k)} p_s^{(k)}). \quad (1)$$

### Content diversity

$$\Phi_n = \frac{S}{S-1} \sum_{s=1}^S p_s^{(n)} (1 - p_s^{(n)}). \quad (2)$$

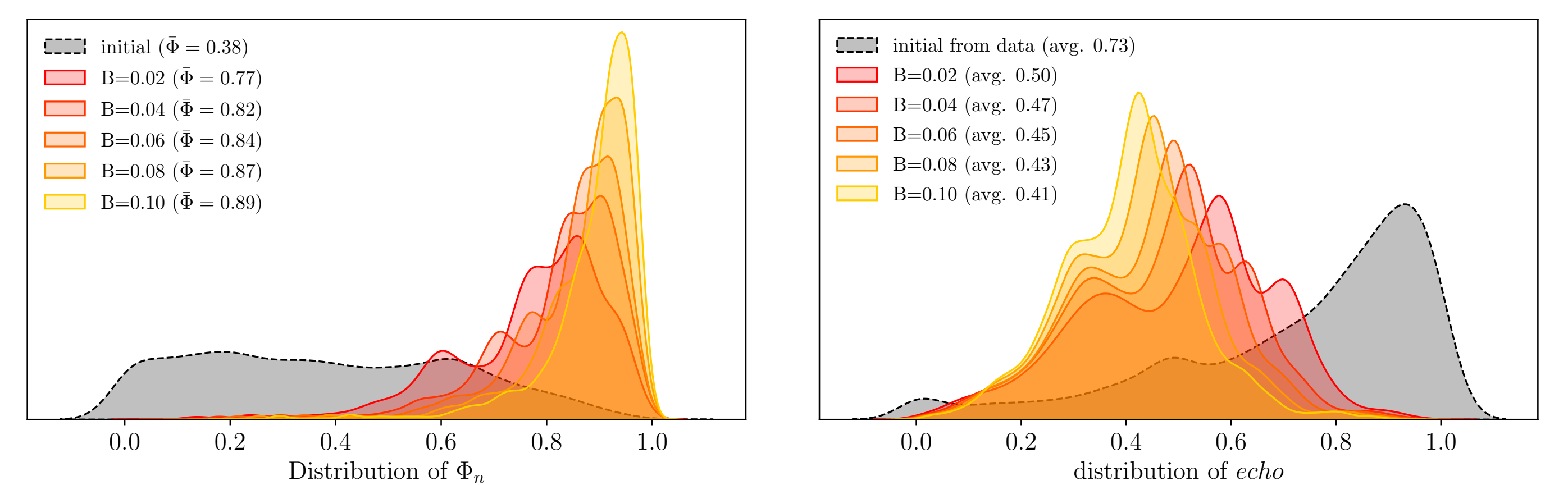
$p_s^{(n)}$ : avg. prop. of content from party  $s$  on the newsfeed of  $n$ .



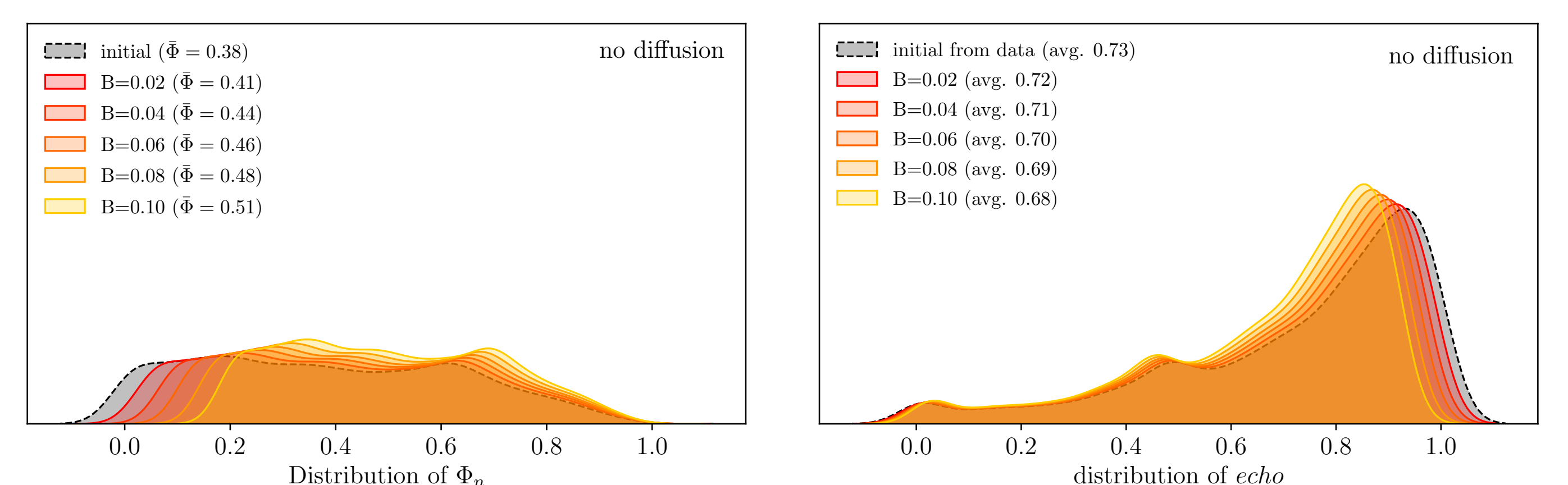
Empirical estimates of  $p$  versus theoretical values.

## Optimisation

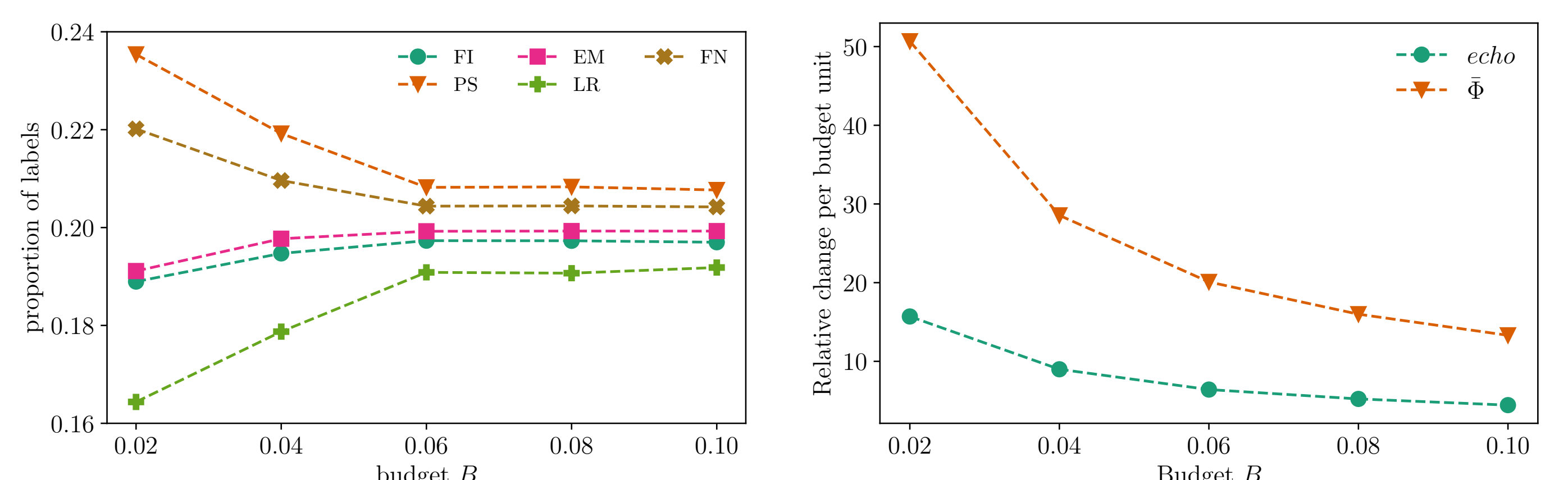
$$\begin{aligned} \underset{x, p}{\operatorname{argmax}} \quad & \frac{1}{N} \sum_n \Phi_n \\ \text{s.t.} \quad & \text{for all } n, s : \\ & \underbrace{\frac{p_s^{(n)}}{1-B} \sum_{k \in \mathcal{L}^{(n)}} (\lambda^{(k)} + \mu^{(k)}) = x_s^{(n)} + \sum_{k \in \mathcal{L}^{(n)}} (\lambda_s^{(k)} + \mu^{(k)} p_s^{(k)})}_{\text{model equation}}, \\ & \underbrace{\sum_s x_s^{(n)} = \frac{B}{1-B} \sum_{k \in \mathcal{L}^{(n)}} (\lambda^{(k)} + \mu^{(k)})}_{\text{budget constraint}}, \\ & x_s^{(n)}, p_s^{(n)} \geq 0. \end{aligned}$$



Left: Increase of diversity for various budgets. Right: decrease of exposure to like-minded content.



Same but without diffusion.



Left: proportion of content circulating through the network for each party. Right: impact of each budget unit.

## Contact

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