



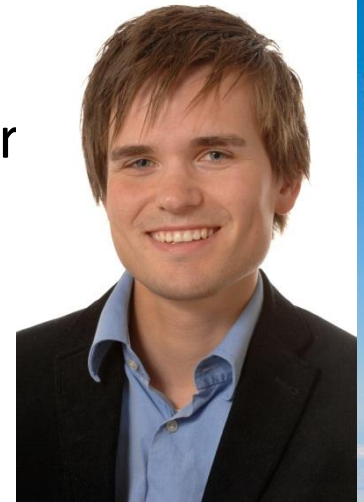
# An Empirical Study of Valuation and User Behavior in Social Networking Services (SNS)



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## → Motivation

- Is each network connection of equal value?
- Does content productivity increase with network size?
- What generates value in a network and how do you model it?



# Agenda

## I. Introduction

### A. Network Effects

### B. Network Laws:

- Sarnoff's law

- Metcalfe's law

- Zipf's law

## II. Our Study

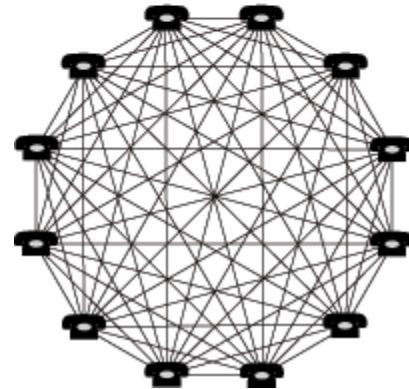
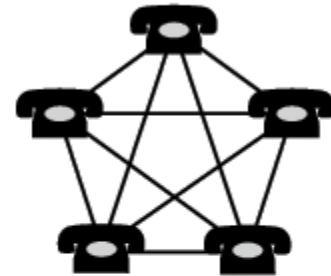
### A. Content popularity: SNS and Zipf's Law

### B. Correlation between productivity and network size in SNS

### C. Proposed model for SNS valuation

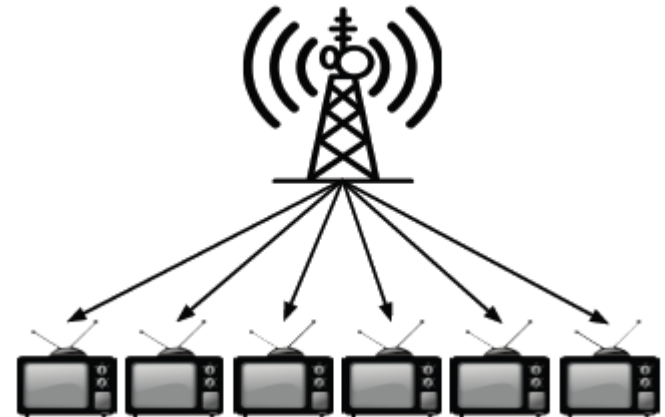
## → Network Effects

- Utility of consumption is affected by the number of other users using the same or compatible products
- Examples include telephone networks and social networking services



## → Sarnoff's law

- Applicable to broadcast networks
- Value proportional to number of subscribers
- $S(n) = n$



# → Metcalfe's law

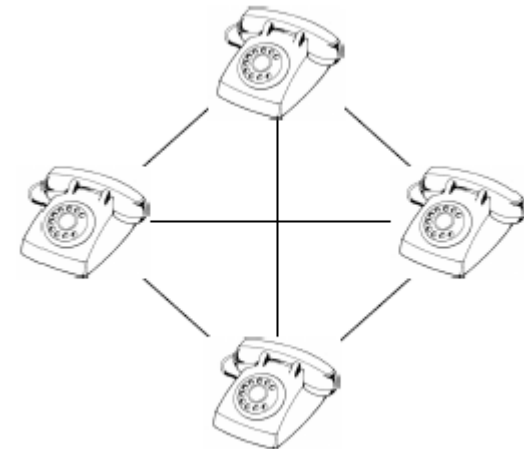
→ Applicable to communication networks

- Telephone
- Fax
- Email

→ Value proportional to number of members squared

$$\rightarrow M(n) = \frac{(n-1)n}{2}$$

$$\rightarrow M(n) \approx n^2$$



## → Zipf's law

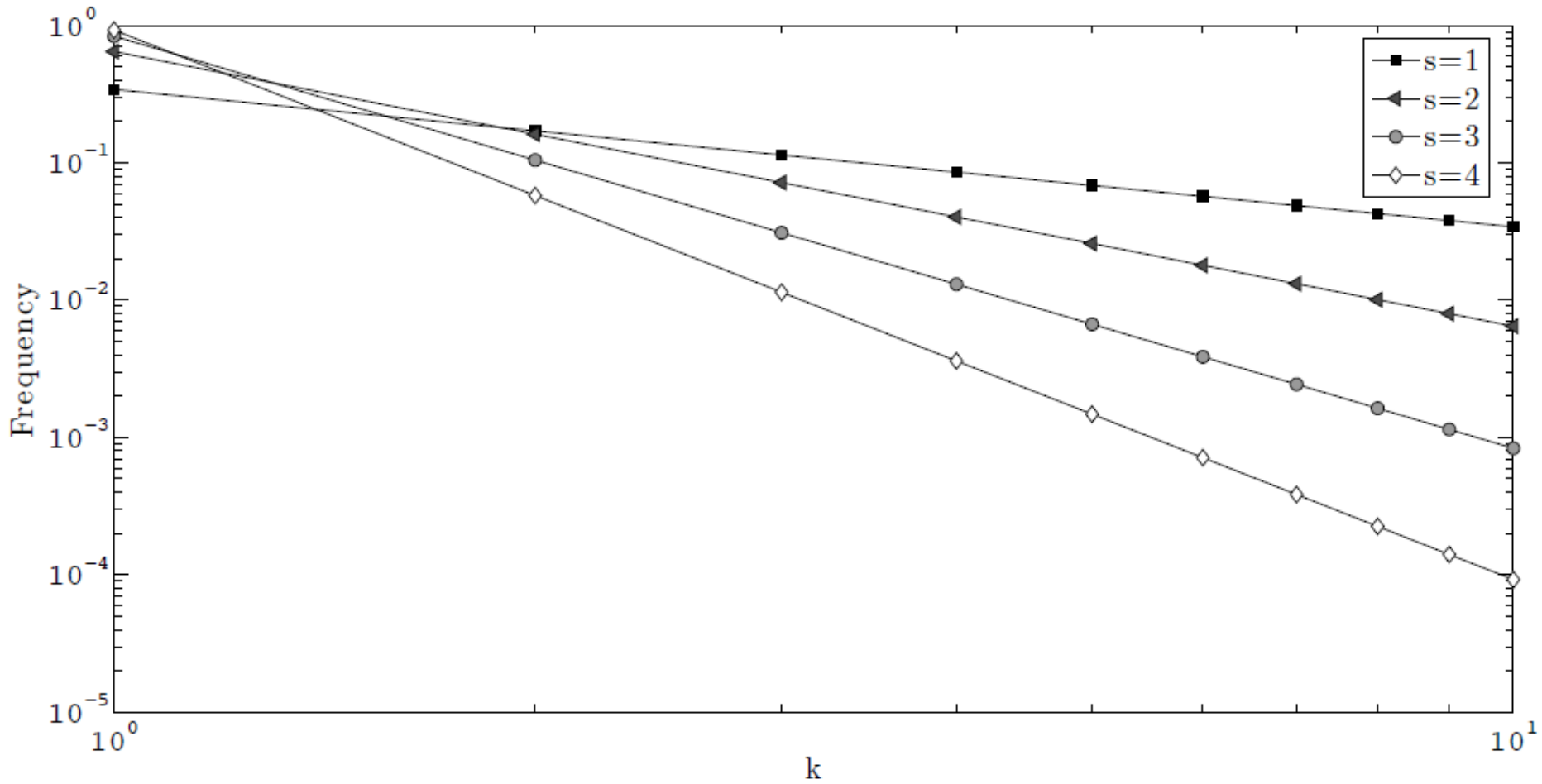
- Several types of data follow a Zipfian distribution.
- Zipf's law has proven to be very accurate for modeling popularity of data, such as words in the English language and sizes of large cities

$$\rightarrow f(k, s, N) = \frac{1/k^s}{\sum_{n=1}^N 1/n^s}, s > 0 \in \mathbb{R}, n \in \mathbb{I}$$

- $N$  is the number of elements
- $k$  is their popularity rank
- $s$  is the value of the exponent characterizing the distribution



# → Zipf's law





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### A. Content popularity: SNS and Zipf's Law

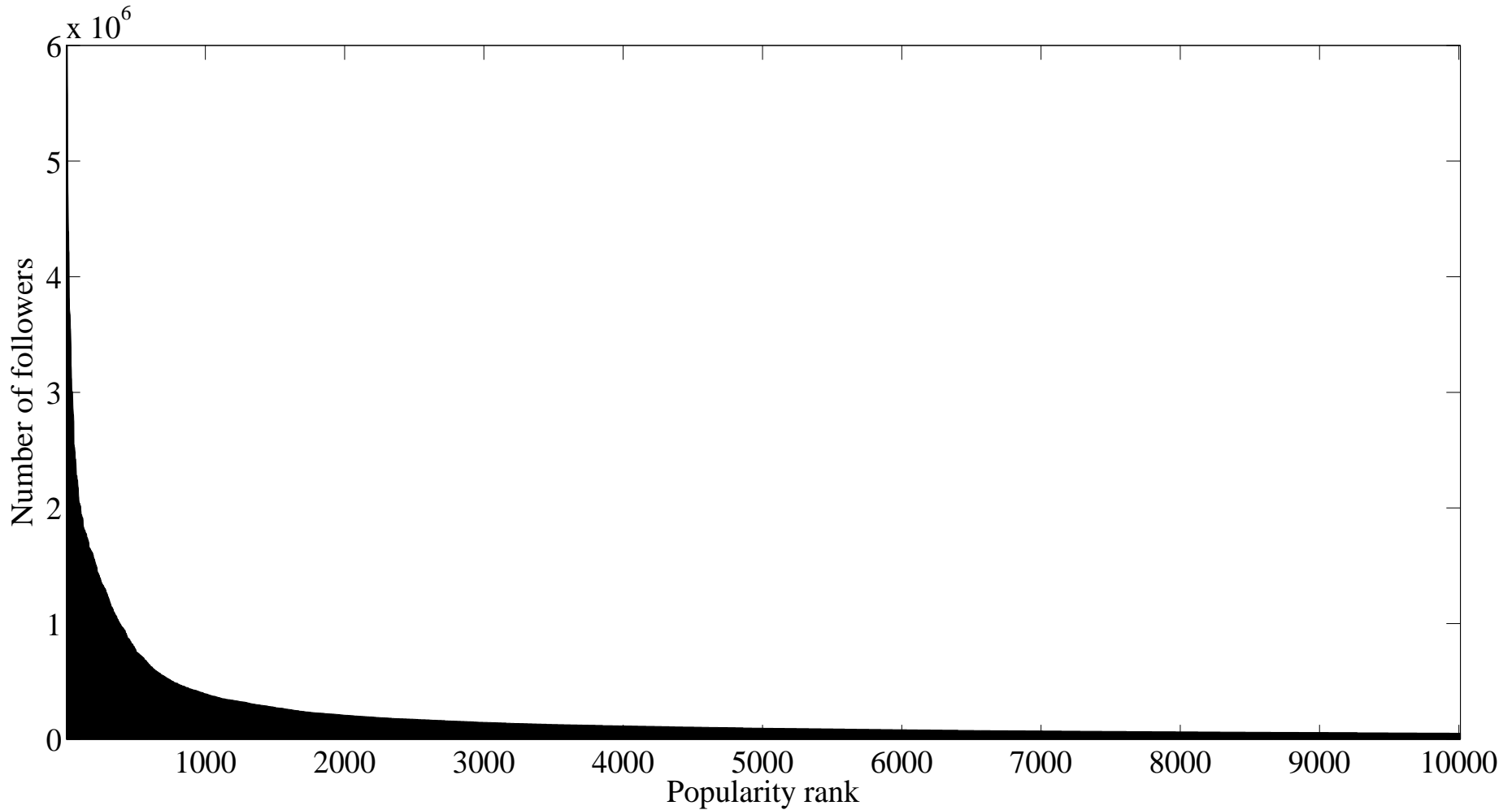
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### C. Proposed model for SNS valuation

## → Content popularity - Twitter and Zipf's Law

- Zipf's law was used to model popularity of Twitter users
- An Internet page containing statistics for the 10 020 most popular Twitter users was used as data basis
- The value of the exponent,  $s$ , in Zipf's law was optimized to find the best-fit Zipf probability mass function

# → Raw data from Twitter



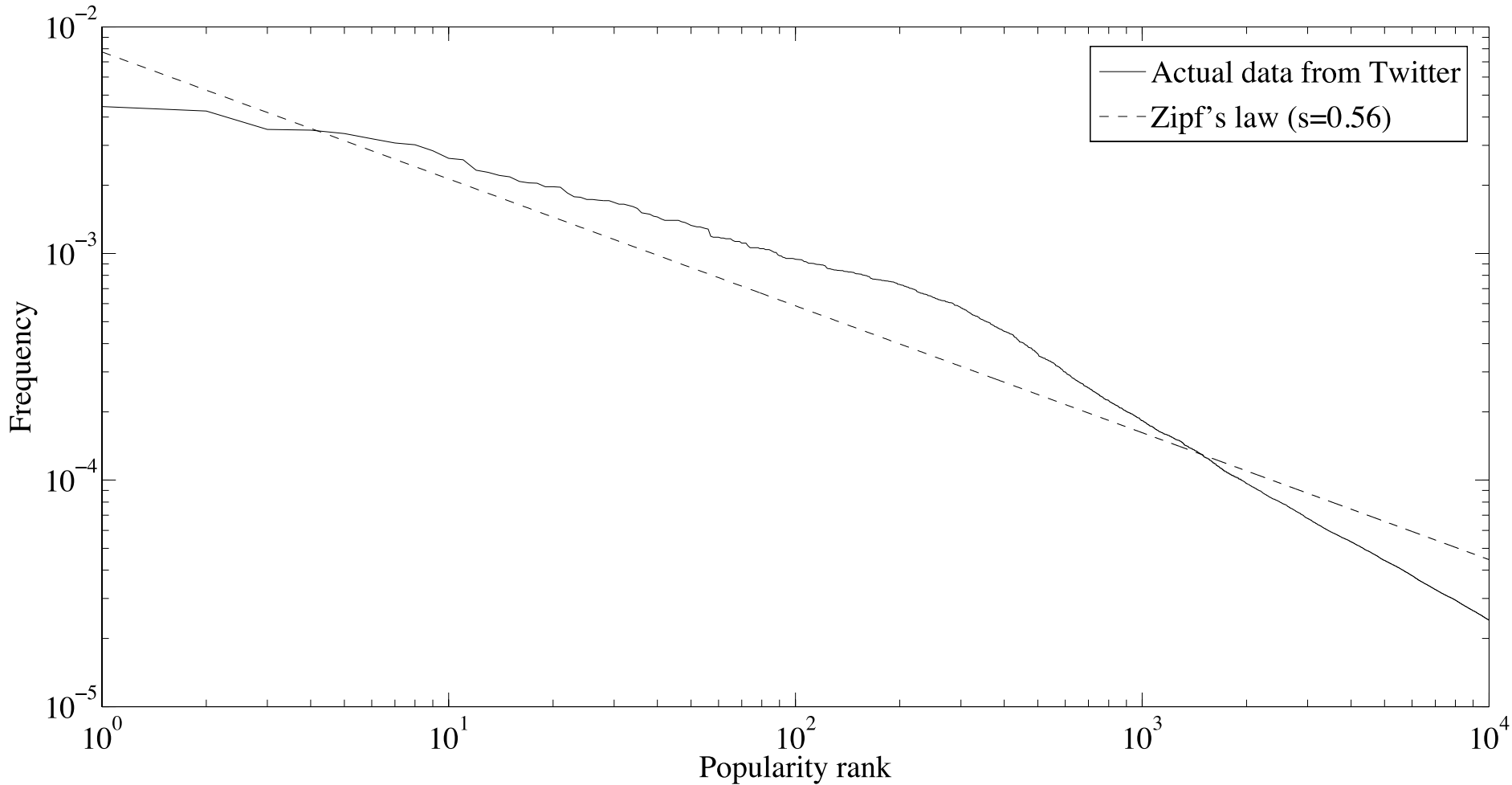
## → Optimization of the exponent, s (Twitter)

→ A minimum sum of squares problem was solved to find the optimal value of the exponent, s:

→ 
$$\min \sum_{k=1}^{10011} \left( f_k - \frac{1/k^s}{\sum_{n=1}^{10011} 1/n^s} \right)^2 \text{ subject to } s > 0$$

→ The optimal value of s was 0.56

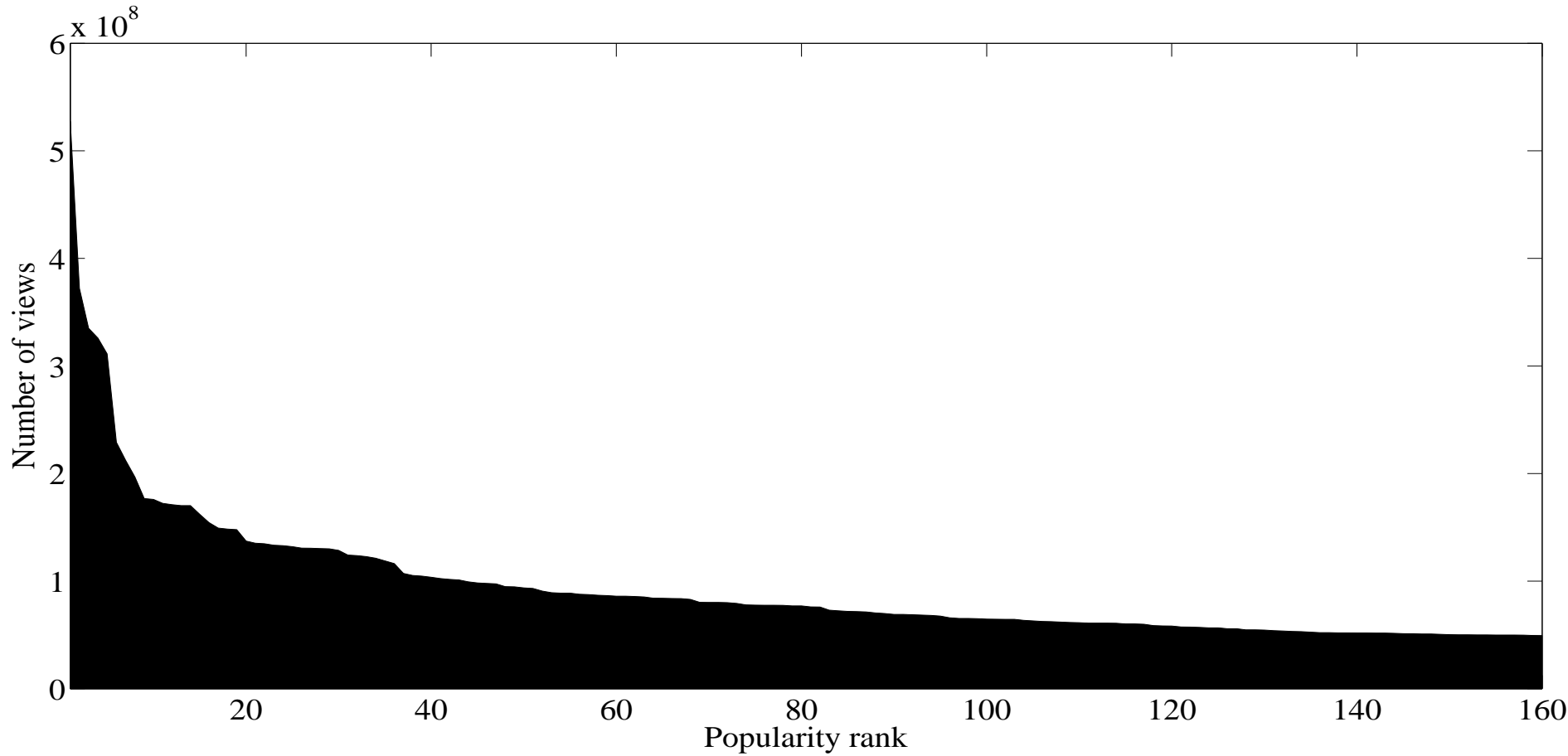
# → Twitter compared with Zipf's law



## → Content popularity - Youtube and Zipf's Law

- Zipf's law was used to model popularity of Youtube videos
- Number of views for the 160 most popular Youtube videos was retrieved
- The value of the exponent,  $s$ , in Zipf's law was optimized to find the best-fit Zipf probability mass function

# → Raw data from Youtube

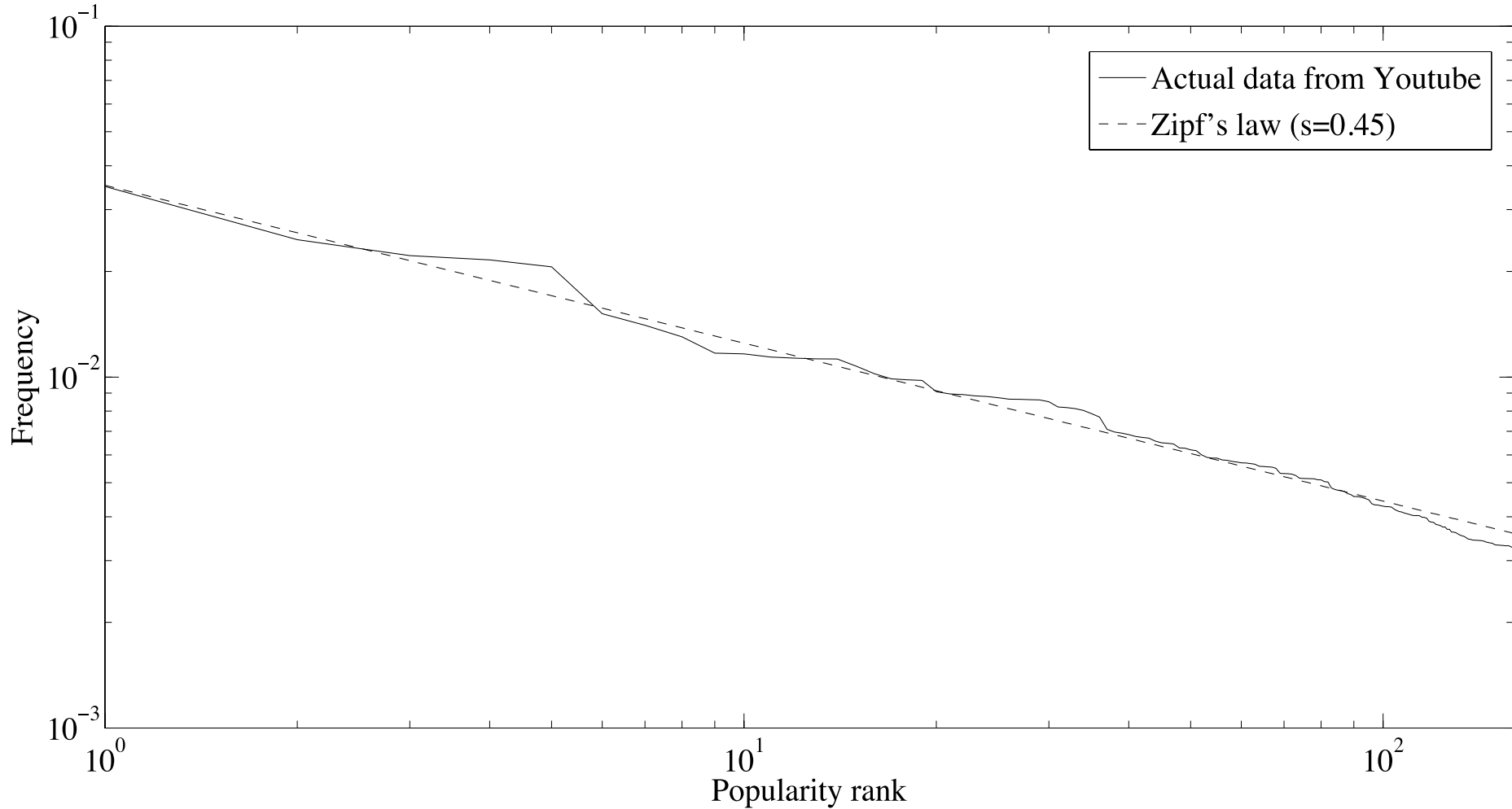




## → Optimization of the exponent, $s$ (Youtube)

- The procedure performed was the same as with the fitting of Zipf's law with Twitter.
- The optimal value of the exponent this time,  $s$ , was 0.45

# → Youtube compared with Zipf's law



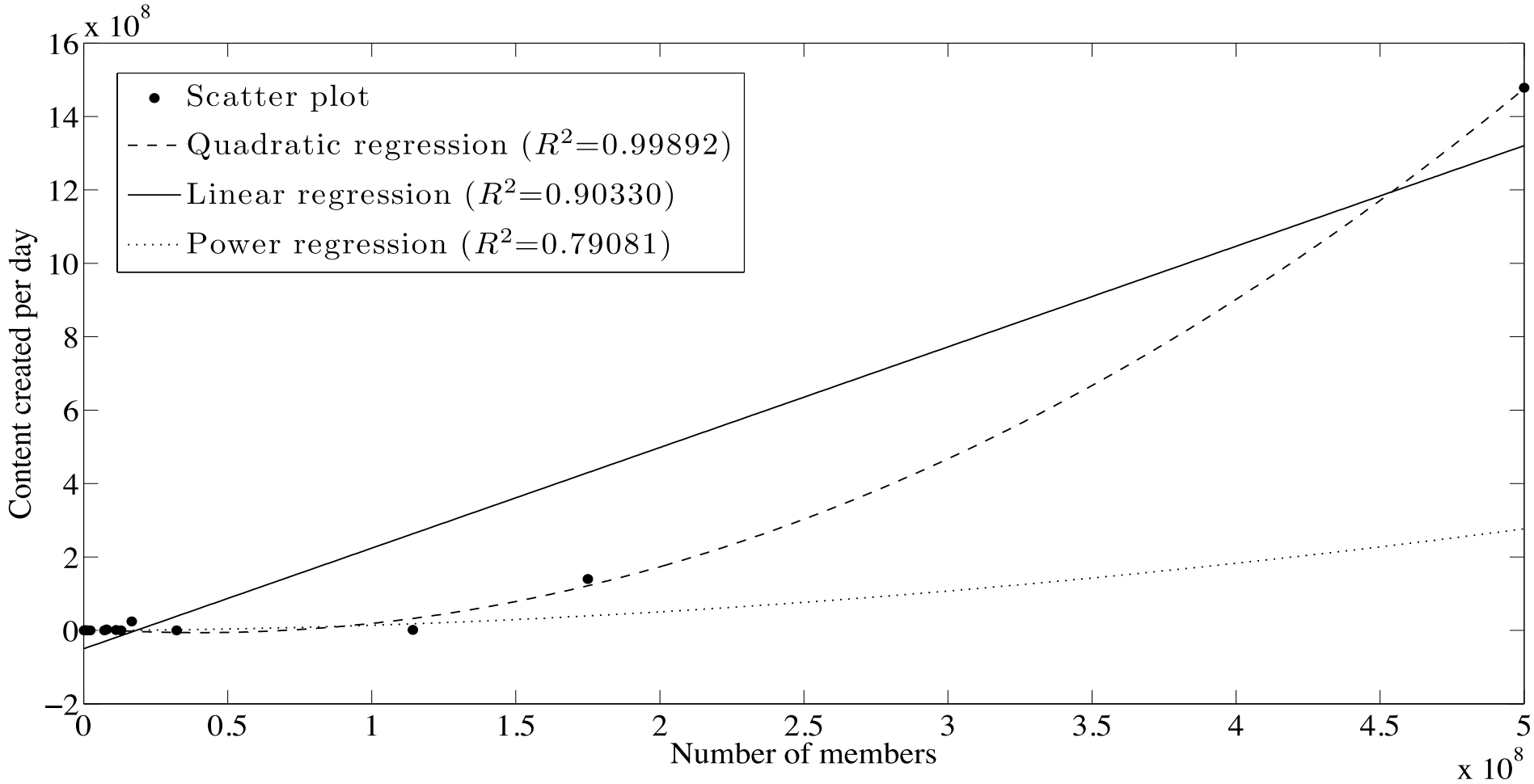


# Correlation between productivity and network size in SNS

- The relationship between network size and content created in SNS was studied to see whether content productivity increases with network size
- 15 social networking services provided information about network size and content productivity
- Various best-fit functions were calculated and tested



# Correlation between productivity and network size in SNS





# Correlation between productivity and network size in SNS

- The quadratic model fitted the data significantly better than the linear model
- Consequently, average productivity increased with network size for SNS studied

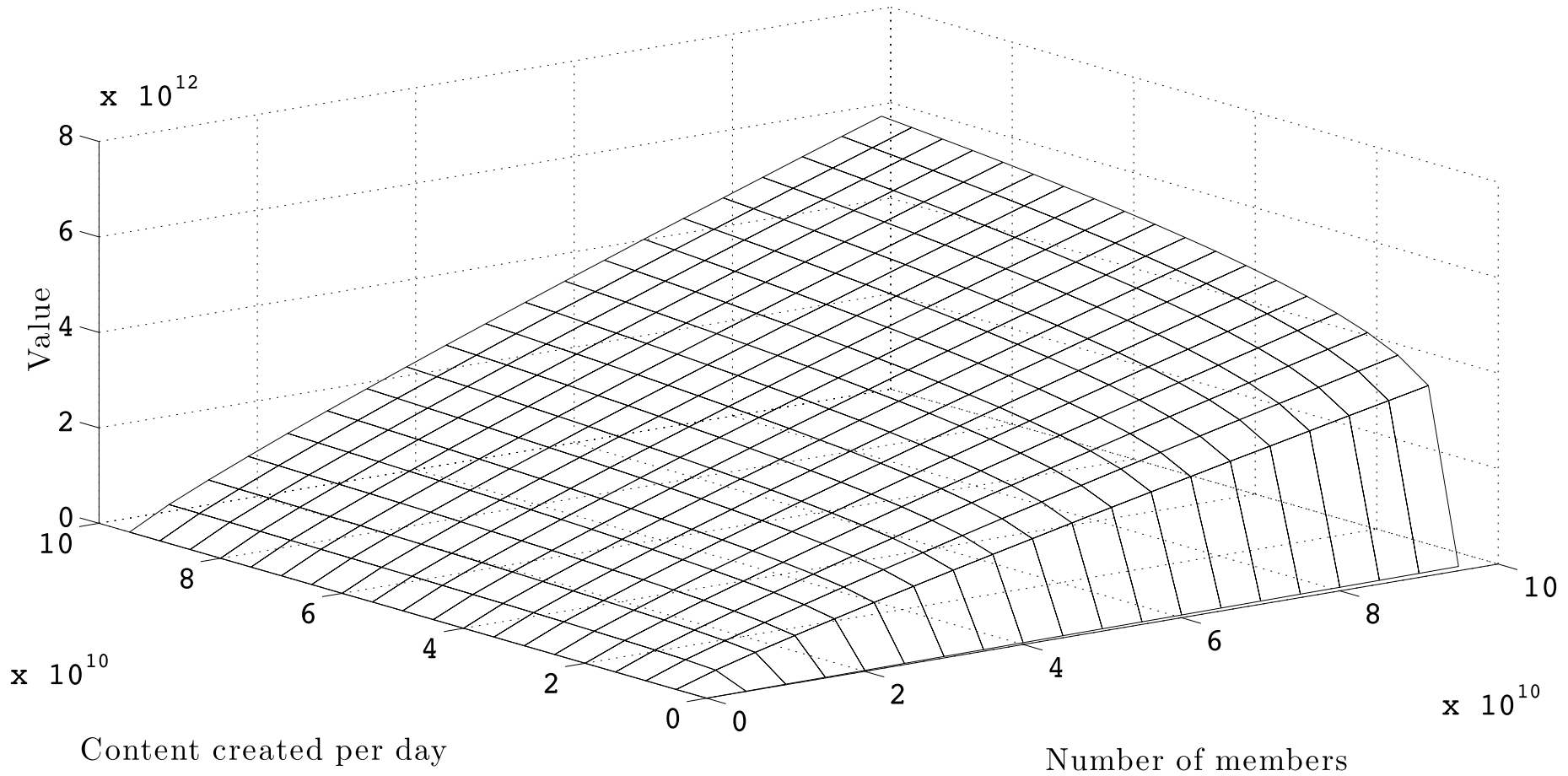
# → Estimated value of Social Networking Services

- Three alternative response surface models for valuation of SNS were based on network size, average content created per day and actual market value in United States dollar
- Only five social networks were able to provide the information needed for our valuation model
- The software Mathematica 8 was used to calculate a best-fit linear, quadratic and power response surface

# → Estimated value of Social Networking Services

- The estimated best-fit linear and quadratic response surfaces had both undesirable properties. These models were therefore considered inappropriate
- The best-fit power response surface was:
- $V_{prs}(n, c) = 14.1514n^{0.892437}c^{0.167022}$ 
  - where  $n$  is the network size and  $c$  average content created per day

# → Estimated value of Social Networking Services





## → Conclusions

- Zipf's law with  $s=0.56$  was not an accurate describer of popularity of Twitter members
- Zipf's law with  $s=0.45$  was a good describer of popularity of Youtube videos
- Content productivity increases with network size for SNS studied
- An empirical model for SNS valuation was proposed based on two variables: network size ( $n$ ) and average content created per day ( $c$ ). The best-fit response surface was the following power function:
  - $V_{prs}(n, c) = 14.1514n^{0.892437}c^{0.167022}$



# Questions?



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