

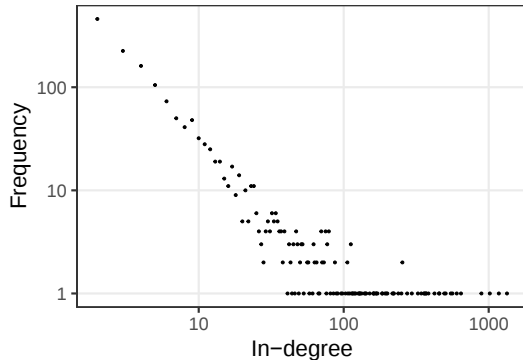
Evidencing preferential attachment in dependency network evolution

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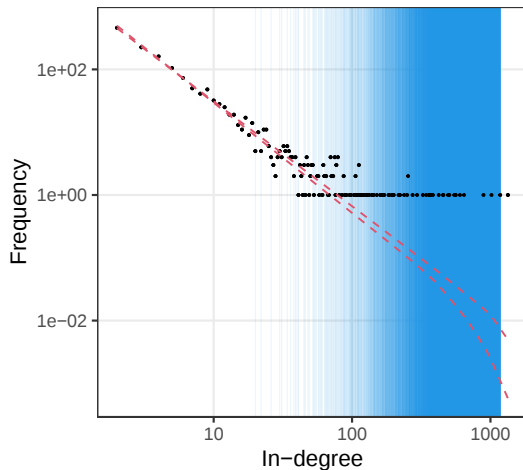
Introduction

- ▶ Interest in degree distribution
- ▶ Power law $\rightarrow$ Scale-freeness
- ▶ Associated with preferential attachment
 1. New node joins & brings new edges
 2. Node i selected with weight $g(d_i)$
 3. Barabási and Albert (1999):
 $g(d_i) = d_i$



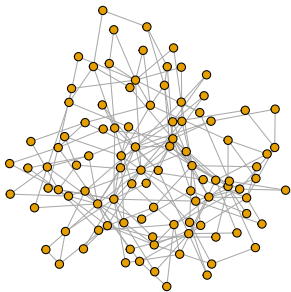
A few issues

1. General case $\nrightarrow$ power law
 - ▶ $g(d_i) = d_i^\alpha$ (Krapivsky and Redner 2001)
2. Other models $\rightarrow$ scale-freeness
 - ▶ Generalised random graphs (Hofstad 2016)
3. Debate on ubiquity
 - ▶ Broido and Clauset (2019): rare in real networks
 - ▶ Voitalov et al. (2019): rebuttal
 - ▶ Lee, Eastoe, and Farrell (2024): partial scale-freeness
4. Main cause: only snapshots available

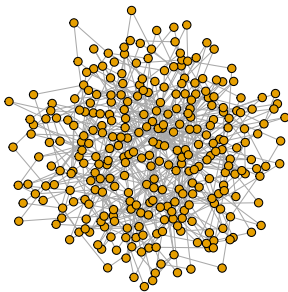


Take a step back

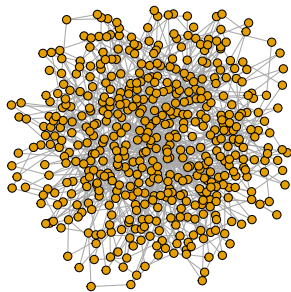
$t-1$



t



$t+1$



Goals

If we have the evolution data:

1. Can we evidence the preferential attachment?
2. How do we model the data in a probabilistic / statistical way?
3. What is the precise form of the weight function?
 - ▶ $g(d_i) = d_i^\alpha$ or a more general form?

Daily increments of R packages on CRAN

##		from	to	add
## 1		glmmsr	methods	FALSE
## 2		GpGp	FNN	TRUE
## 3		pmxTools	stats	FALSE
## 4		ceterisParibus	knitr	TRUE
## 5		pmxTools	xpose	FALSE
## 6		mRchmadness	shiny	FALSE
## 7		glmmsr	lme4	FALSE
## 8		mRchmadness	rvest	FALSE
## 9		xpose	dplyr	FALSE
## 10		xpose	utils	FALSE

⋮

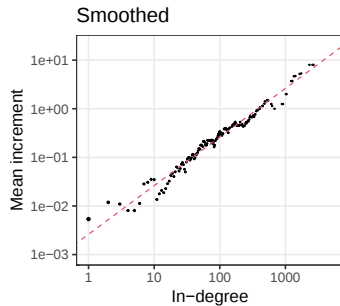
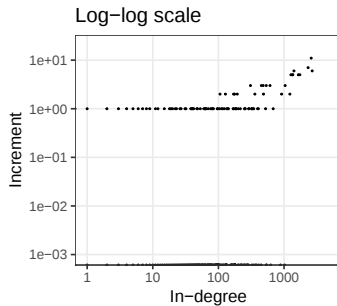
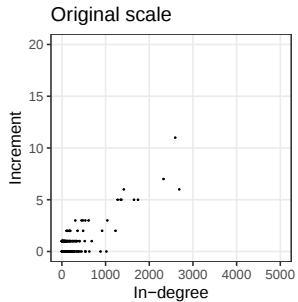
- ▶ From 2019-01-29 to 2024-12-31
- ▶ Obtained by
`crandep::get_dep_all_packages()`
- ▶ ... which uses
`tools::CRAN_package_db()`
- ▶ Change in Imports for one day
- ▶ Some new packages, some existing ones

Aggregating

##	previous date	indegree	increment	count
## 1	2019-01-29	0	0	9315
## 2	2019-01-29	0	1	1
## 3	2019-01-29	1	0	1306
## 4	2019-01-29	2	0	459
## 5	2019-01-29	3	0	226
## 6	2019-01-29	4	0	161
## 7	2019-01-29	5	0	103
## 8	2019-01-29	6	0	75
## 9	2019-01-29	7	0	49
## 10	2019-01-29	8	0	40

⋮

Scatter plot

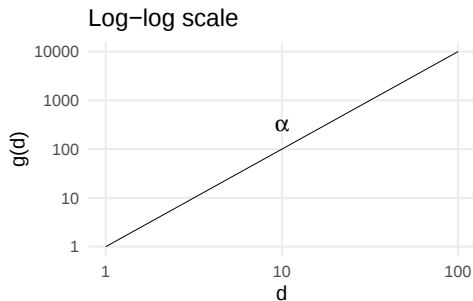
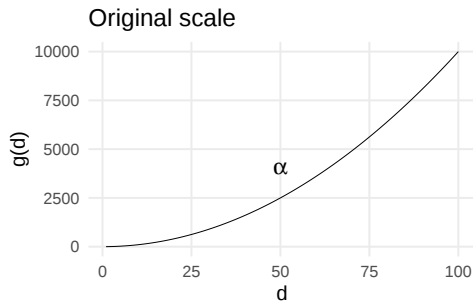


Regress increment on in-degree

- ▶ The **statistical** model based on preferential attachment
- ▶ Y_i : number of new edges / increments for node i
- ▶ $\{Y_i\} \sim$ Multinomial with prob. $\left\{ \frac{d_i^\alpha}{\sum_j d_j^\alpha} \right\}$, 1 new edge
- ▶ $\{Y_i\} | M \sim$ Multinomial with prob. $\left\{ \frac{d_i^\alpha}{\sum_j d_j^\alpha} \right\}$, $M \sim \text{Poisson}(\mu)$ new edges
 - ▶ $\Rightarrow \{Y_i\} \sim$ Independent Poissons with mean $\left\{ \frac{\mu d_i^\alpha}{\sum_j d_j^\alpha} \right\}$
 - ▶ $\log E(Y_i) = \alpha \log d_i + \log \mu - \underbrace{\log \left(\sum_j d_j^\alpha \right)}_{\text{constant}}$

Revisiting weight function

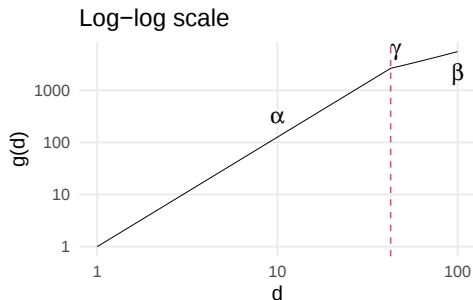
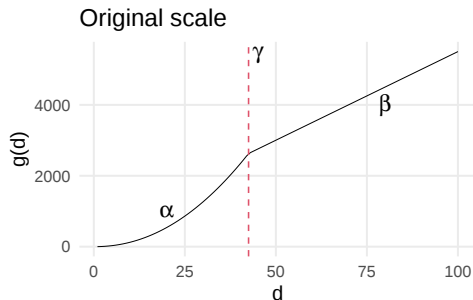
- ▶ $g(d) = d^\alpha$
 - ▶ Inflexible tail behaviour of degree distribution (Oliveira and Spencer 2005; Rudas, Tóth, and Valkó 2007)
- ▶ Real data displays subtle tail behaviour
 - ▶ Heavy-tailed, but not as heavy as power law (Lee, Eastoe, and Farrell 2024)



A more general one

►
$$g(d) = \begin{cases} d^\alpha, & d \leq \gamma, \\ \gamma^\alpha + \beta(d - \gamma), & d \geq \gamma \end{cases}$$

- Flexible heavy-tail behaviour (Boughen, Lee, and Palacios Ramirez 2025)



Modelling recipe

1. Regress increment on in-degree

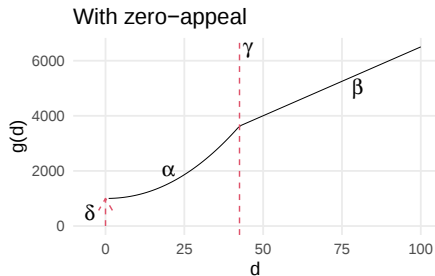
► $Y_i \sim \text{Poisson} \left(\frac{\mu g(d_i)}{\sum_j g(d_j)} \right)$

2. Choice of $g(d)$

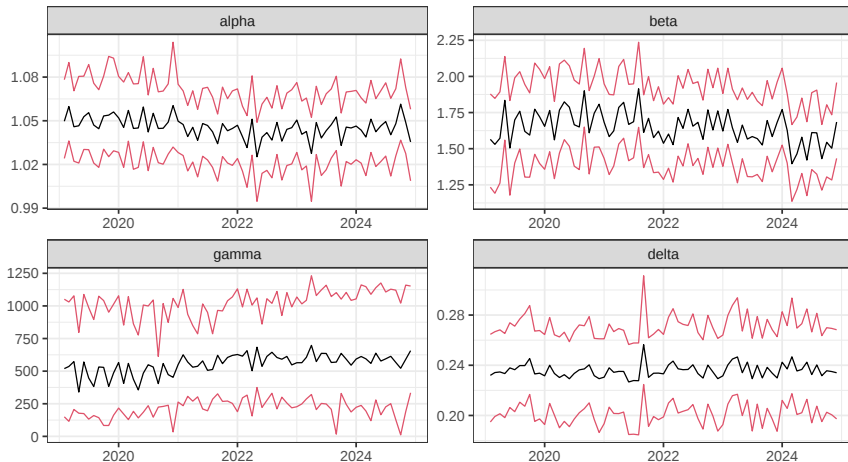
- Piecewise function (previous slide)
- Power function for benchmark:
 $g(d) = d^\alpha$
- Include parameter for zero-appeal δ

3. Bayesian inference for $(\alpha, \beta, \gamma, \delta, \mu)$

- A set of parameters for each month of increments
- Results of α most important

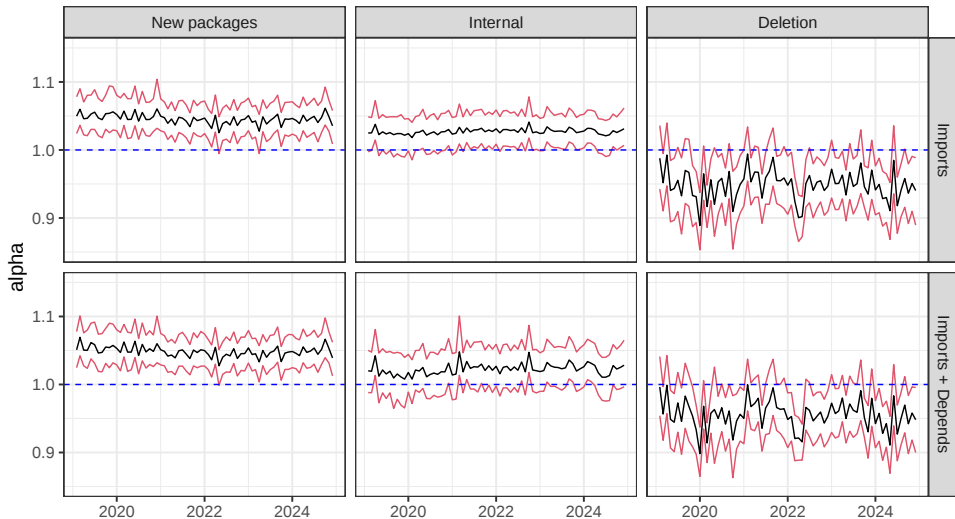


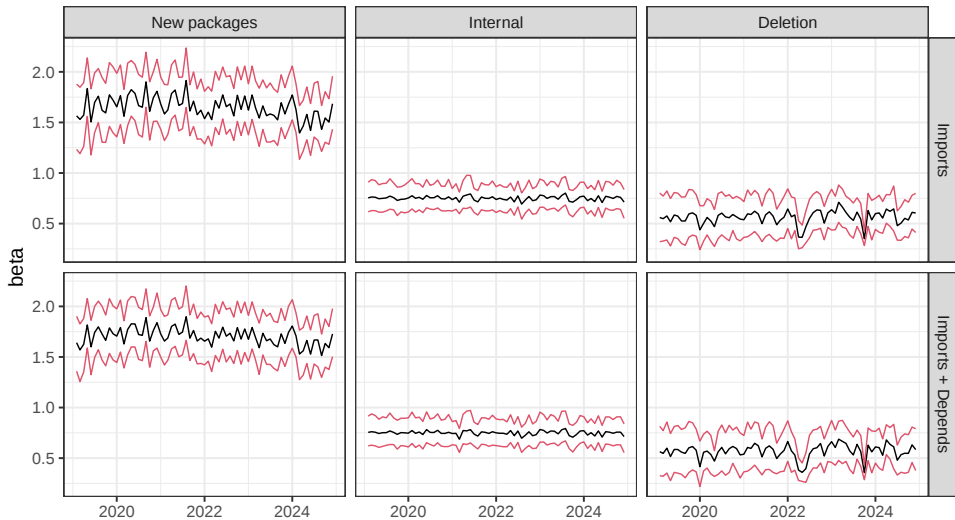
Piecewise function, Imports, new packages



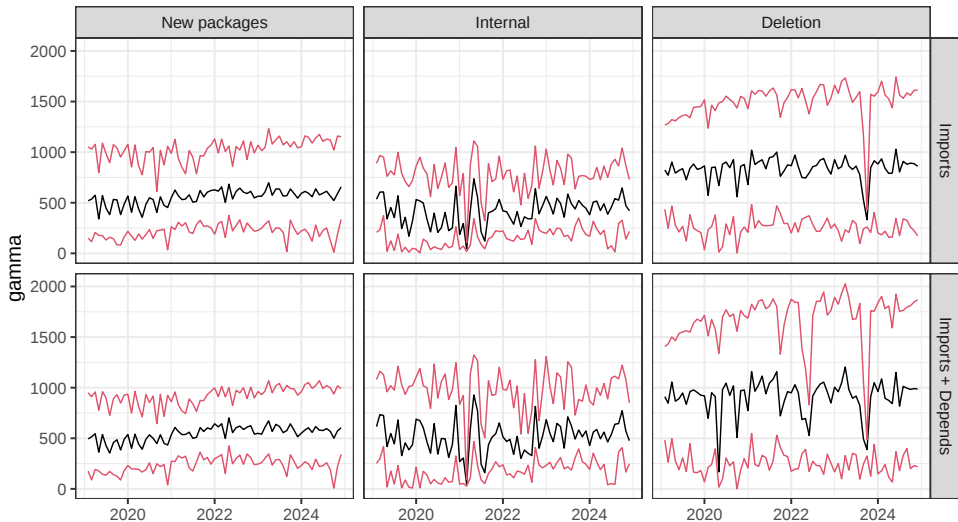
► Very similar results for α & δ using power function

α (super-/sub-linear?)

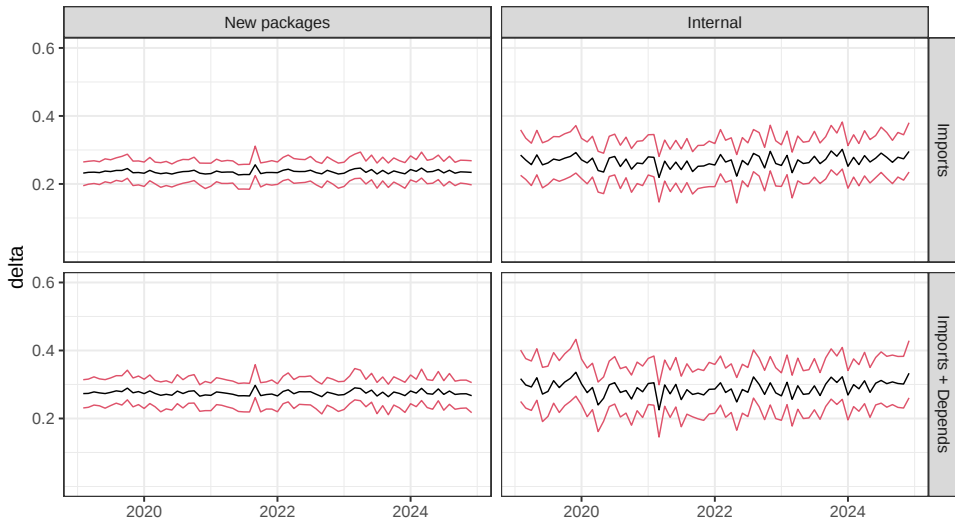


β 

γ



δ



Summary

- ▶ Preferential attachment evidenced from increments of network evolution
 - ▶ increment vs in-degree on log-log scale
 - ▶ Poisson regression model
- ▶ Piecewise weight function of in-degree
 - ▶ allows partial super-/sub-linear preferential attachment
 - ▶ accommodates flexible heavy tail (of the degree distribution)
- ▶ Model applied to CRAN package dependencies
 - ▶ parameters stable over the observation period
 - ▶ adding edges: **Partial superlinear preferential attachment**
 - ▶ deleting edges: **Sublinear preferential detachment without zero-appeal**

Thank you

- ▶ This presentation's slides: <https://bit.ly/sunbelt2025>
- ▶ On partial scale-freeness: <https://doi.org/10.1111/stan.12355>
- ▶ On degree tail flexibility: <https://doi.org/10.48550/arXiv.2506.18726>



Bibliography

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