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**BATH**

# PageRank: Ranking chemicals based on heterogeneous data

Amelie Klein and Uziel Gonzalez



# Introduction

Google's PageRank method was developed to evaluate the importance of web-pages via their link structure. The mathematics of PageRank, however, are entirely general and apply to any graph or network in any domain.

It's even used for systems analysis of road networks, as well as biology, chemistry, neuroscience, and physics.



# Data

When you are seeking a herbicidal, many chemicals should be tested. Those can be applied to multiple plants of multiple species at various doses and measuring the responses in many screens.



# Data

| Compound | Rate | Method   | Specie 1 | Specie 2 | Specie 3 | Specie 4 | Specie 5 | Specie 6 | Specie 7 |
|----------|------|----------|----------|----------|----------|----------|----------|----------|----------|
| 1        | 1000 | EPS POST | 60       | 90       | 70       | 90       | 0        | 70       | 100      |
| 2        | 1000 | EPS POST | 0        | 30       | 0        | 0        | 0        | 0        | 0        |
| 3        | 1000 | EPS POST | 100      | 100      | 80       | 100      | 0        | 90       | 100      |
| 3        | 1000 | EPS POST | 100      | 100      | 90       | 100      | 0        | 90       | 100      |
| 4        | 1000 | EPS POST | 0        | 30       | 0        | 10       | 0        | 0        | 30       |
| 4        | 1000 | EPS POST | 20       | 100      | 20       | 60       | 0        | 10       | 70       |
| 5        | 1000 | EPS POST | 0        | 60       | 20       | 30       | 0        | 10       | 50       |
| 6        | 1000 | EPS POST | 20       | 60       | 30       | 30       | 0        | 20       | 60       |
| 7        | 1000 | EPS POST | 40       | 80       | 90       | 40       | 0        | 90       | 90       |
| 7        | 1000 | EPS POST | 20       | 80       | 90       | 70       | 0        | 80       | 80       |
| 8        | 1000 | EPS POST | 90       | 100      | 90       | 100      | 0        | 90       | 90       |
| 9        | 1000 | EPS POST | 10       | 90       | 80       | 100      | 0        | 80       | 90       |
| 10       | 1000 | EPS POST | 90       | 80       | 90       | 80       | 0        | 90       | 90       |

# Toy example

## Transition probability

| Dosis | Compound | Specie1 | Specie2 | Specie3 |
|-------|----------|---------|---------|---------|
| 50    | 1        | 60      | 90      | 0       |
| 50    | 2        | 30      | 10      | 20      |
| 50    | 3        | 50      | 40      | 70      |

|       |    |          |
|-------|----|----------|
| 0     | 20 | 70       |
| 60+90 | 0  | 50+40+50 |
| 60+90 | 0  | 0        |



From our graph, we construct a transition matrix,  
and calculate its left eigenvector.

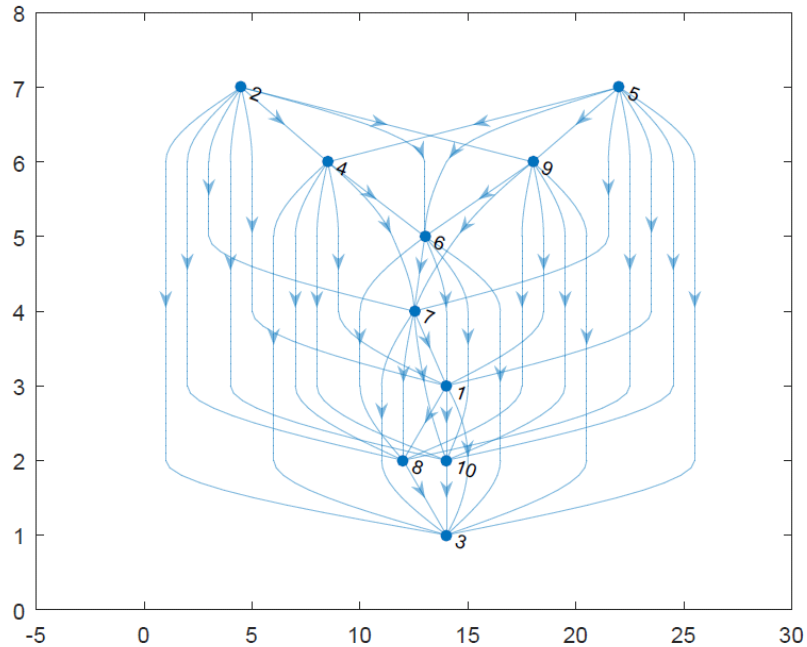
This eigenvector gives us a score for each  
chemical.

This scores give us a ranking,

# Analysis of Data

| Compound | Rate | Method   | Specie 1 | Specie 2 | Specie 3 | Specie 4 | Specie 5 | Specie 6 | Specie 7 |
|----------|------|----------|----------|----------|----------|----------|----------|----------|----------|
| 1        | 1000 | EPS POST | 60       | 90       | 70       | 90       | 0        | 70       | 100      |
| 2        | 1000 | EPS POST | 0        | 30       | 0        | 0        | 0        | 0        | 0        |
| 3        | 1000 | EPS POST | 100      | 100      | 80       | 100      | 0        | 90       | 100      |
| 3        | 1000 | EPS POST | 100      | 100      | 90       | 100      | 0        | 90       | 100      |
| 4        | 1000 | EPS POST | 0        | 30       | 0        | 10       | 0        | 0        | 30       |
| 4        | 1000 | EPS POST | 20       | 100      | 20       | 60       | 0        | 10       | 70       |
| 5        | 1000 | EPS POST | 0        | 60       | 20       | 30       | 0        | 10       | 50       |
| 6        | 1000 | EPS POST | 20       | 60       | 30       | 30       | 0        | 20       | 60       |
| 7        | 1000 | EPS POST | 40       | 80       | 90       | 40       | 0        | 90       | 90       |
| 7        | 1000 | EPS POST | 20       | 80       | 90       | 70       | 0        | 80       | 80       |
| 8        | 1000 | EPS POST | 90       | 100      | 90       | 100      | 0        | 90       | 90       |
| 9        | 1000 | EPS POST | 10       | 90       | 80       | 100      | 0        | 80       | 90       |
| 10       | 1000 | EPS POST | 90       | 80       | 90       | 80       | 0        | 90       | 90       |

## A first approach



- 1] 0.1787
- 2] 0.0868
- 3] 0.8669
- 4] 0.0910
- 5] 0.0868
- 6] 0.1046
- 7] 0.1220
- 8] 0.2820
- 9] 0.0910
- 10] 0.2820





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# Future work

# References

**PageRank beyond the Web by  
David F. Gleich**

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**Graph Theory and Linear Algebra of Google's Pagerank**  
(<http://chipjacks.com>)

**The predictive power of ranking systems in association football**  
Jan Lasek