



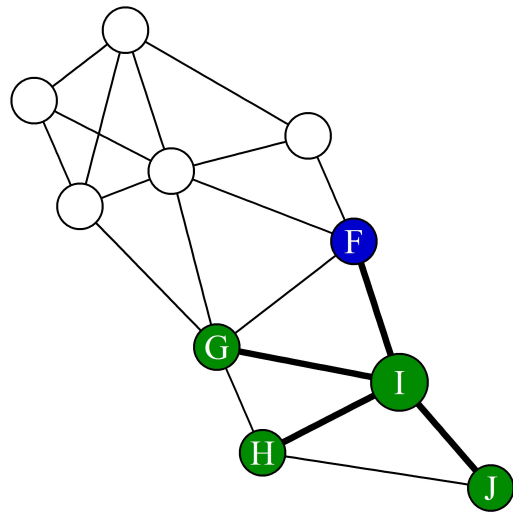
PREDICTIVE ANALYTICS USING NETWORKED DATA IN R

# Basic Network features

Bart Baesens, Ph.D.

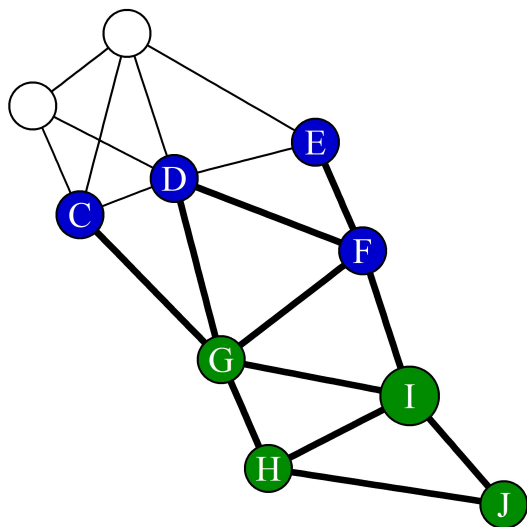
Professor of Data Science, KU Leuven and University of Southampton

# Neighborhood features



- First order degree
  - Number of connected nodes

```
degree(g)
A B C D E F G H I J
4 3 4 6 3 4 5 3 4 2
>
```

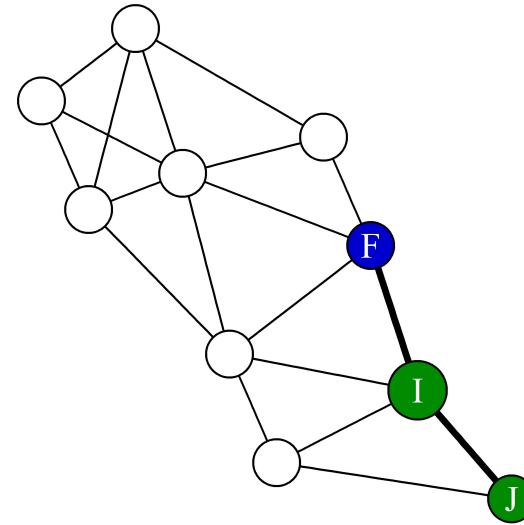
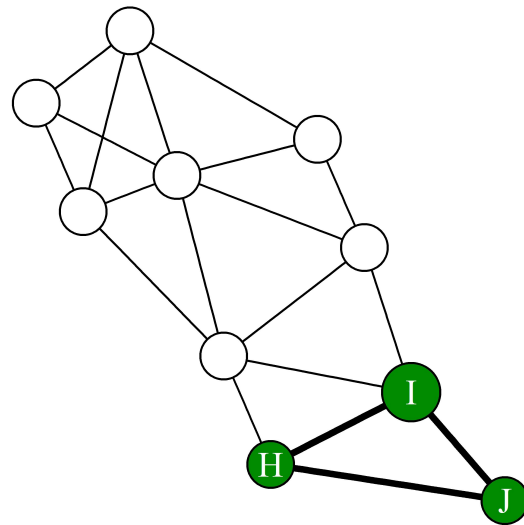


- Second order degree
  - Number of connected nodes that are two or less edges away

```
neighborhood.size(g, order = 2)
[1] 7 7 9 9 8 10 10 7 8 5
```



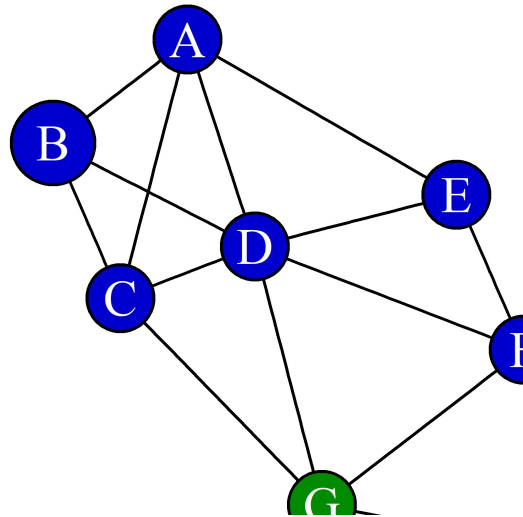
# Neighborhood features - triangles



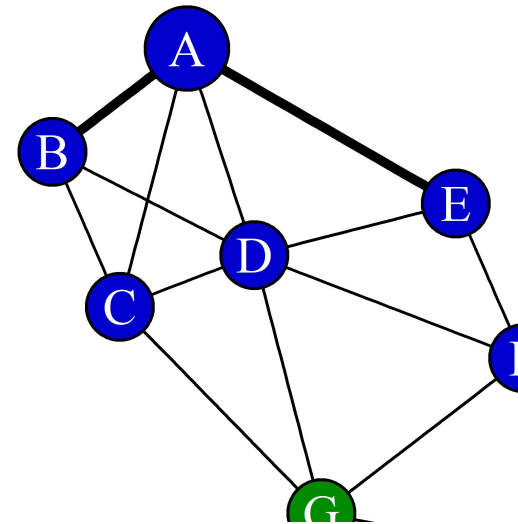
```
count_triangles(g)
[1] 4 3 4 7 2 3 4 2 3 1
```

# Centrality Features

- Betweenness



- Closeness



```
betweenness(g)
```

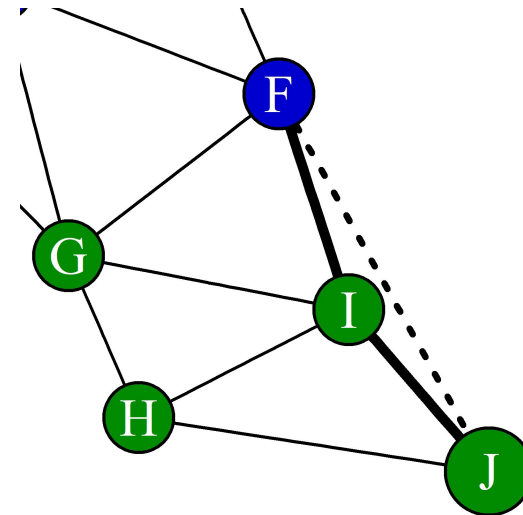
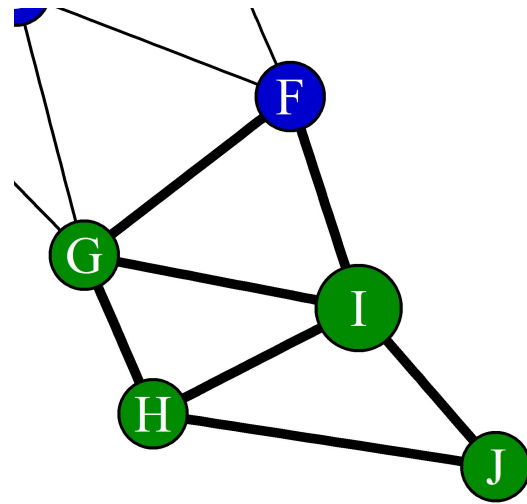
A	B	C	D	E	F	G	H	I	J
1.00	0.00	3.32	8.10	0.92	5.37	11.47	2.07	5.77	0.00

```
closeness(g)
```

A	B	C	D	E	F	G	H	I	J
0.06	0.05	0.07	0.08	0.06	0.07	0.08	0.06	0.06	0.04



# Transitivity



```
transitivity(g,type = 'local')  
[1] 0.67 1.00 0.67 0.47 0.67 0.50 0.40 0.67 0.50 1.00
```



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**Let's practice!**



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# Link Based Features

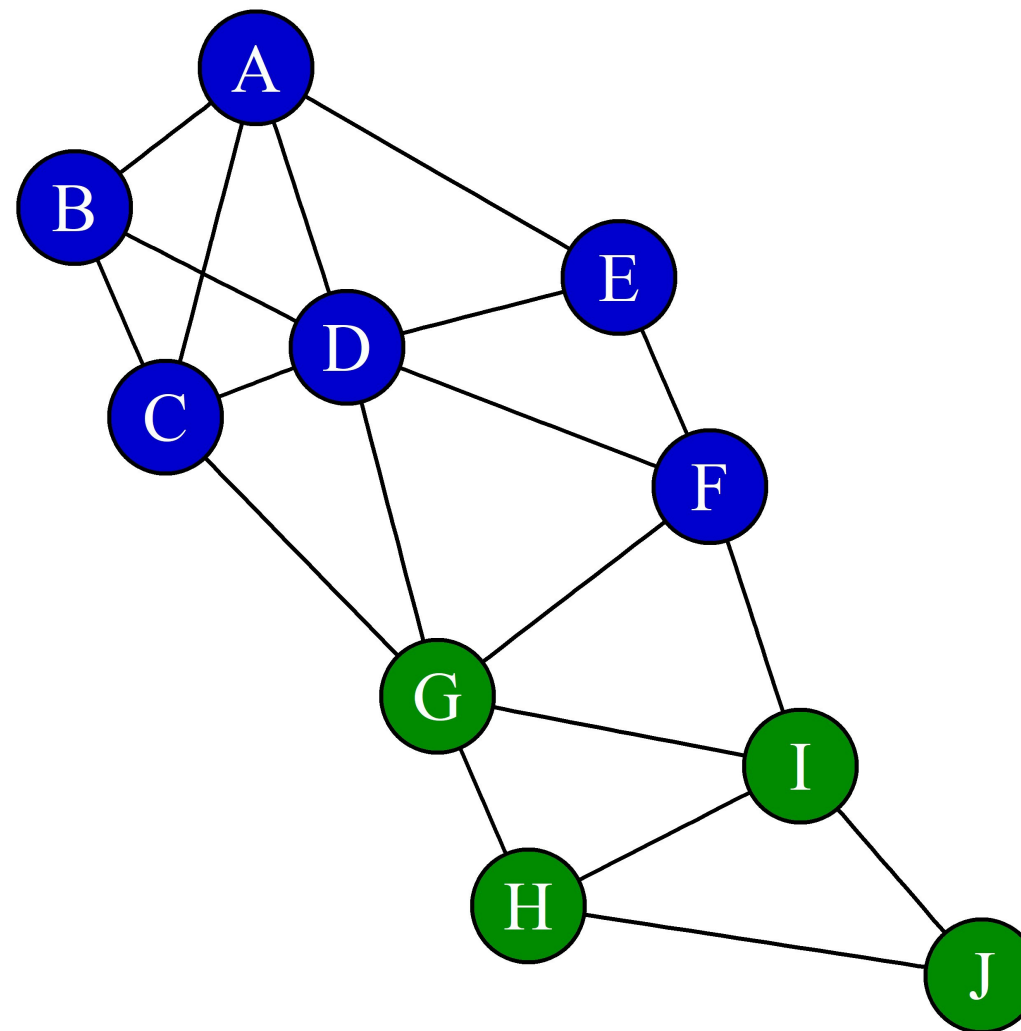
María Óskarsdóttir, Ph.D.  
Post-doctoral researcher



# Adjacency Matrices

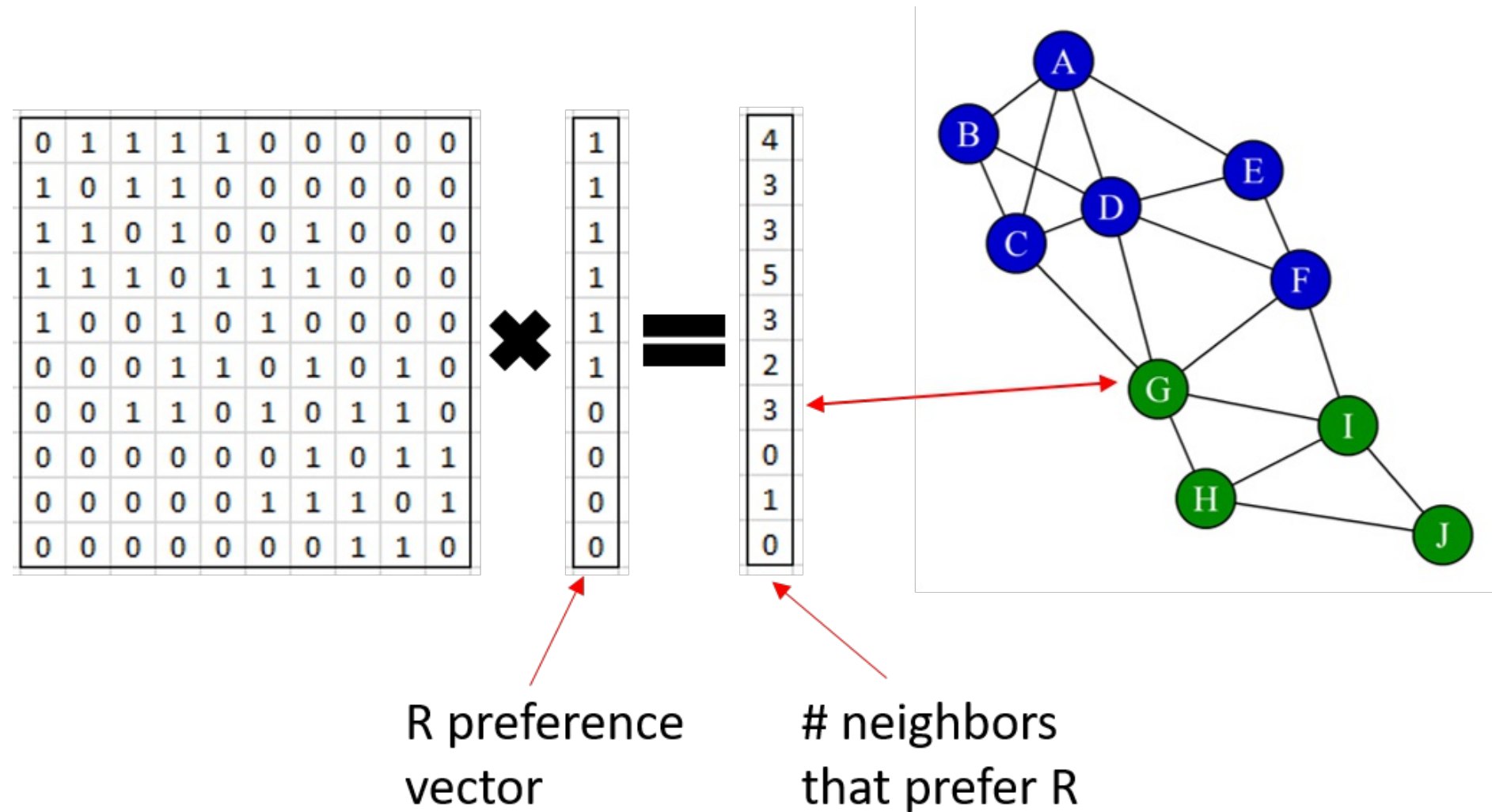
	A	B	C	D	E	F	G	H	I	J
A	0	1	1	1	1	0	0	0	0	0
B	1	0	1	1	0	0	0	0	0	0
C	1	1	0	1	0	0	1	0	0	0
D	1	1	1	0	1	1	1	0	0	0
E	1	0	0	1	0	1	0	0	0	0
F	0	0	0	1	1	0	1	0	1	0
G	0	0	1	1	0	1	0	1	1	0
H	0	0	0	0	0	0	1	0	1	1
I	0	0	0	0	0	1	1	1	0	1
J	0	0	0	0	0	0	0	1	1	0

```
A <- get.adjacency(g)
```





# Link based features

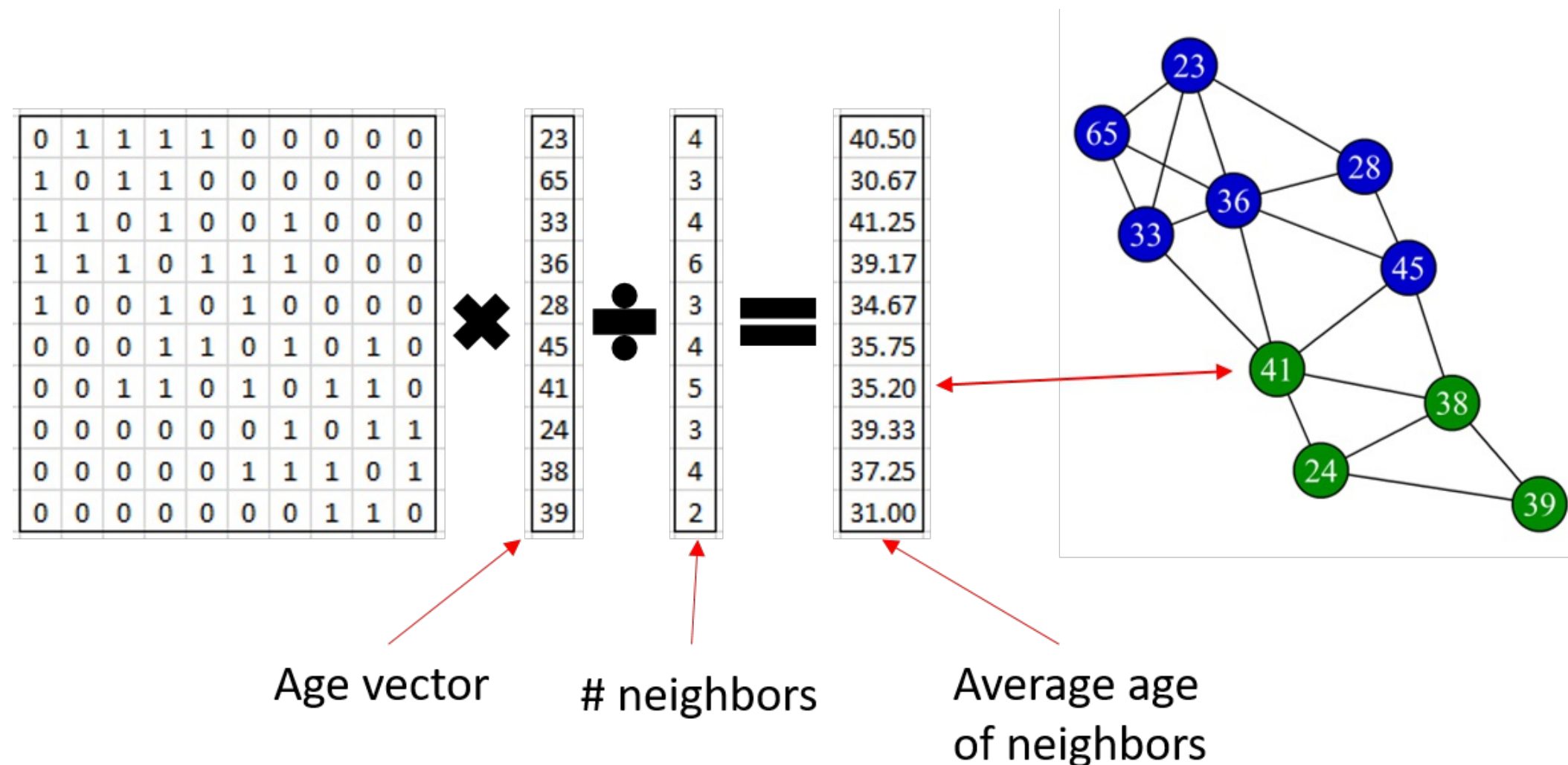


```

preference <- c(1,1,1,1,1,1,0,0,0,0)
rNeighbors <- A %*% preference
as.vector(rNeighbors)
[1] 4 3 3 5 3 2 3 0 1 0

```

# Neighborhood features



```
age <- c(23,65,33,36,28,45,41,24,38,39)
degree <- degree(g)
averageAge <- A %*% age / degree
```



## PREDICTIVE ANALYTICS USING NETWORKED DATA IN R

**Let's practice!**



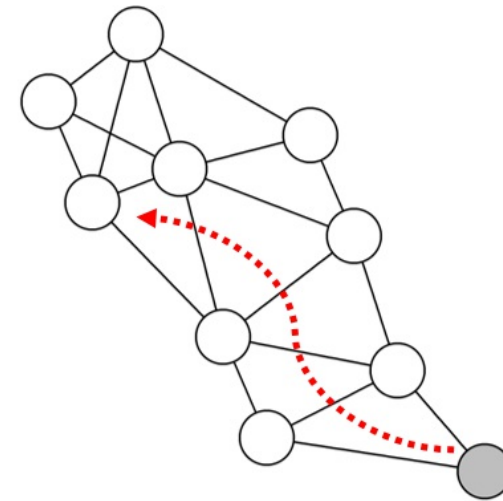
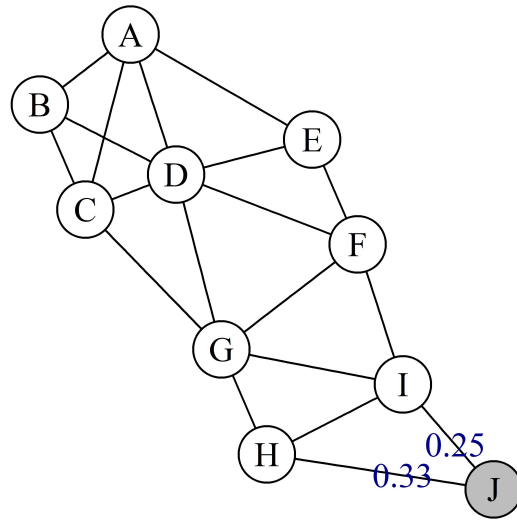
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# PageRank

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Post-doctoral researcher



# The PageRank Algorithm



$$\text{PageRank}_J = \alpha \cdot \left( \frac{1}{3} \cdot \text{PageRank}_H + \frac{1}{4} \cdot \text{PageRank}_I \right) + (1 - \alpha) \cdot e_J$$



# The PageRank Algorithm

$$PR = \alpha \cdot A \cdot PR + (1 - \alpha) \cdot e$$

```
page.rank(g)
```

```
$vector
```

	A	B	C	D	E	F	G
	0.10238312	0.07917232	0.10164910	0.14693274	0.07953551	0.10335821	0.12732387
	H	I	J				
	0.08675903	0.10994175	0.06294435				

```
$value
```

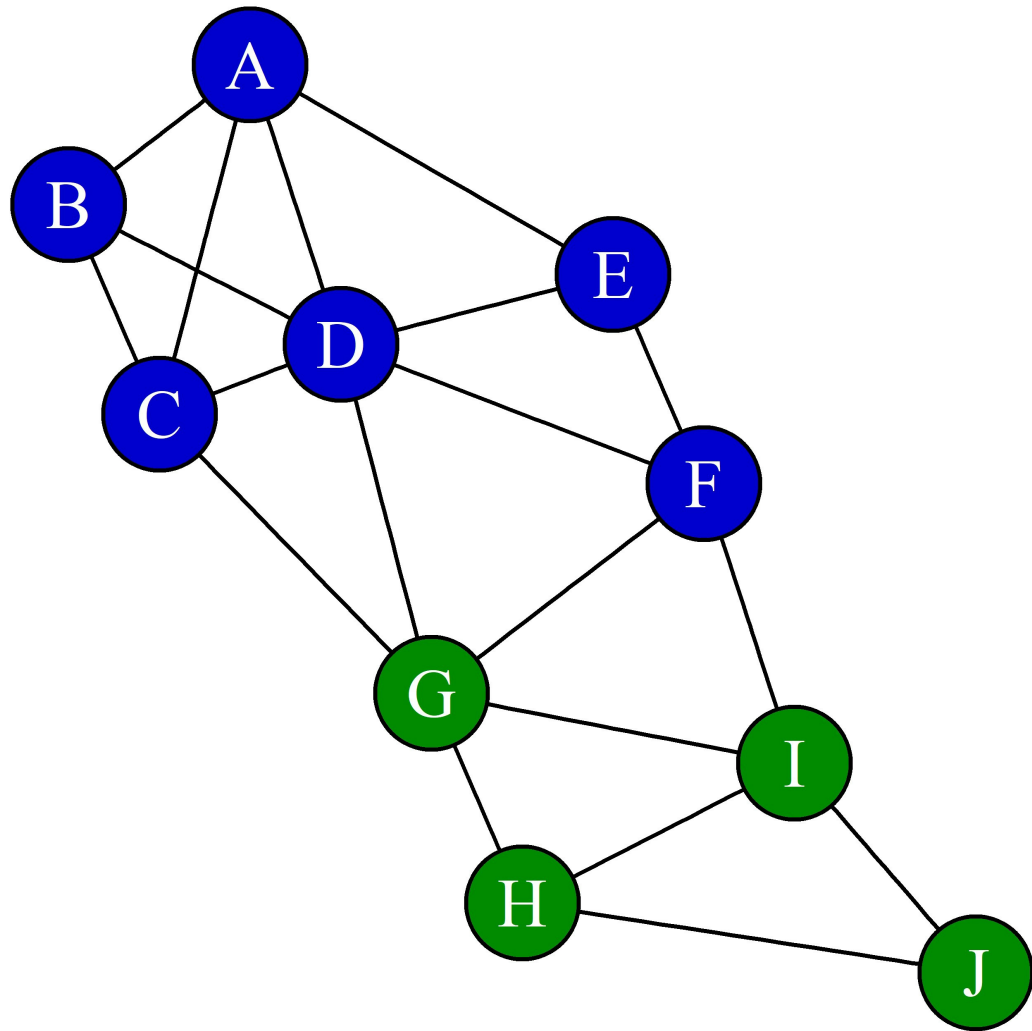
```
[1] 1
```

```
$options
```

```
NULL
```



# Personalized PageRank



```
> page.rank(g,  
  personalized = c(1,0,0,0,0,0,0,0,0,0))  
$vector  
      A      B      C  
0.25528911 0.10363533 0.12156935  
      D      E      F  
0.16625582 0.09366836 0.07466596  
      G      H      I  
0.08473039 0.03285162 0.04785657  
      J  
0.01947748  
  
$value  
[1] 1  
  
$options  
NULL
```



## PREDICTIVE ANALYTICS USING NETWORKED DATA IN R

**Let's practice!**