



# Feature allocations, probability functions, and paintboxes

Tamara Broderick

UC Berkeley

(MIT starting 2015)

# Clustering/Partition

$$\pi_7 = \{\{1, 2, 7\}, \{3, 5\}, \{4\}, \{6\}\}$$

# Clustering/Partition

$$\pi_7 = \{\{1, 2, 7\}, \{3, 5\}, \{4\}, \{6\}\}$$

Document 1

Document 2

Document 3

Document 4

Document 5

Document 6

Document 7

# Clustering/Partition

$$\pi_7 = \{\{1, 2, 7\}, \{3, 5\}, \{4\}, \{6\}\}$$

Document 1

Document 2

Document 3

Document 4

Document 5

Document 6

Document 7

Document 1	■				
Document 2	■				
Document 3		■			
Document 4			■		
Document 5		■			
Document 6				■	
Document 7	■				

# Clustering/Partition

$$\pi_7 = \{\{1, 2, 7\}, \{3, 5\}, \{4\}, \{6\}\}$$

Document 1	■				
Document 2	■				
Document 3		■			
Document 4			■		
Document 5		■			
Document 6				■	
Document 7	■				

“clusters”,  
“blocks (of a  
partition)”

# Clustering/Partition

$$\pi_7 = \{\{1, 2, 7\}, \{3, 5\}, \{4\}, \{6\}\}$$

	Arts	Econ	Sports	Science	Tech
Document 1	■				
Document 2	■				
Document 3		■			
Document 4			■		
Document 5		■			
Document 6				■	
Document 7	■				

“clusters”,  
“blocks (of a  
partition)”

# Latent feature allocation

	Arts	Econ	Sports	Science	Tech
Document 1	■				■
Document 2	■			■	■
Document 3	■	■		■	■
Document 4			■	■	■
Document 5		■			■
Document 6				■	■
Document 7					

# Latent feature allocation

	Arts	Econ	Sports	Science	Tech
Document 1	■				■
Document 2	■			■	■
Document 3	■	■		■	■
Document 4			■	■	■
Document 5		■			■
Document 6				■	■
Document 7					

“features”,  
“topics”



# Latent feature allocation

$$f_7 = \{\{1, 2, 3\}, \{3, 5\}, \{4\}, \{2, 3, 4, 6\}, \{1, 2, 3, 4, 5, 6\}\}$$

	Arts	Econ	Sports	Science	Tech
Document 1	■				■
Document 2	■			■	■
Document 3	■	■		■	■
Document 4			■	■	■
Document 5		■			■
Document 6				■	■
Document 7					

“features”,  
“topics”

# Latent feature allocation

$$f_7 = \{\{1, 2, 3\}, \{3, 5\}, \{4\}, \{2, 3, 4, 6\}, \{1, 2, 3, 4, 5, 6\}\}$$

	Arts	Econ	Sports	Science	Tech
Document 1	■				■
Document 2	■			■	■
Document 3	■	■		■	■
Document 4			■	■	■
Document 5		■			■
Document 6				■	■
Document 7					

Question:  
Can we  
characterize  
exchangeable  
feature  
distributions?

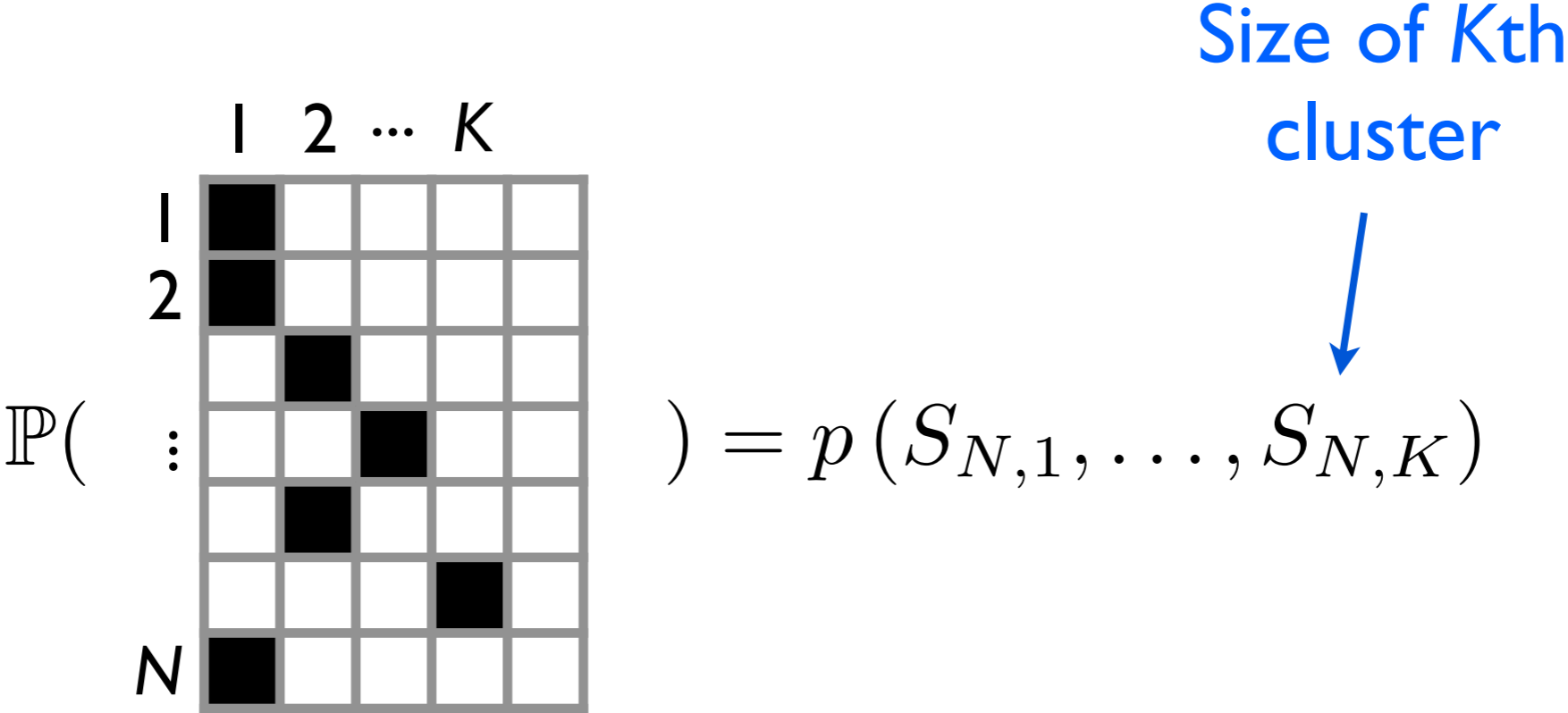
# Exchangeable probability functions

$$\mathbb{P} \left( \begin{array}{c} 1 \\ 2 \\ \vdots \\ N \end{array} \begin{array}{c} 1 \ 2 \ \dots \ K \end{array} \right)$$

# Exchangeable probability functions

$$\mathbb{P}(\begin{array}{c} 1 \\ 2 \\ \vdots \\ N \end{array} \begin{array}{c} 1 \ 2 \ \dots \ K \end{array} \begin{array}{|c|c|c|c|c|} \hline \blacksquare & \square & \square & \square & \square \\ \hline \blacksquare & \square & \square & \square & \square \\ \hline \square & \blacksquare & \square & \square & \square \\ \hline \square & \square & \blacksquare & \square & \square \\ \hline \square & \blacksquare & \square & \square & \square \\ \hline \square & \square & \square & \blacksquare & \square \\ \hline \blacksquare & \square & \square & \square & \square \\ \hline \end{array}) = p(S_{N,1}, \dots, S_{N,K})$$

# Exchangeable probability functions



# Exchangeable probability functions

Exchangeable partition probability function (EPPF)

$$\mathbb{P} \left( \begin{array}{c} 1 \\ 2 \\ \vdots \\ N \end{array} \begin{array}{c} 1 \ 2 \ \dots \ K \\ \begin{array}{|c|c|c|c|c|} \hline \blacksquare & & & & \\ \hline \blacksquare & & & & \\ \hline & \blacksquare & & & \\ \hline & & \blacksquare & & \\ \hline & \blacksquare & & & \\ \hline & & & \blacksquare & \\ \hline \blacksquare & & & & \\ \hline \end{array} \end{array} \right) = p(S_{N,1}, \dots, S_{N,K})$$

# Exchangeable probability functions

Exchangeable partition probability function (EPPF)

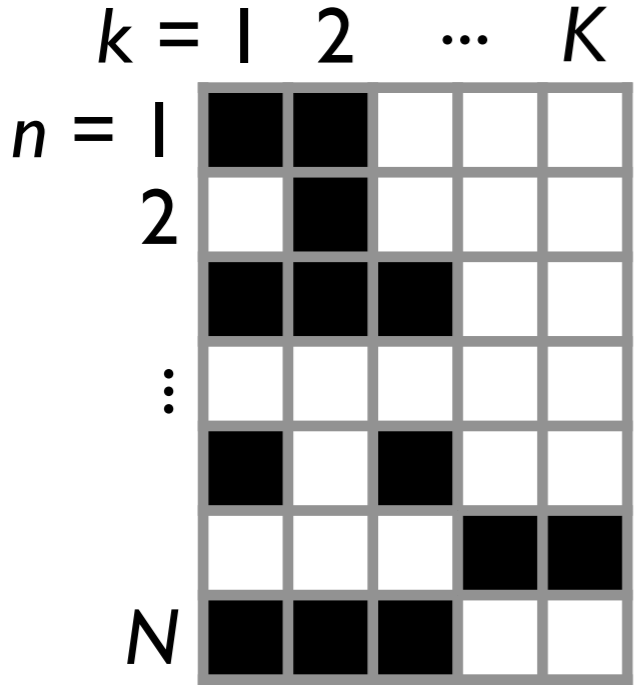
$$\mathbb{P} \left( \begin{array}{c} 1 \\ 2 \\ \vdots \\ N \end{array} \begin{array}{c} 1 \ 2 \ \dots \ K \\ \begin{array}{|c|c|c|c|c|} \hline \blacksquare & & & & \\ \hline \blacksquare & & & & \\ \hline & \blacksquare & & & \\ \hline & & \blacksquare & & \\ \hline & \blacksquare & & & \\ \hline & & & \blacksquare & \\ \hline \blacksquare & & & & \\ \hline \end{array} \end{array} \right) = p(S_{N,1}, \dots, S_{N,K})$$

“Exchangeable feature probability function” (EFPP)?

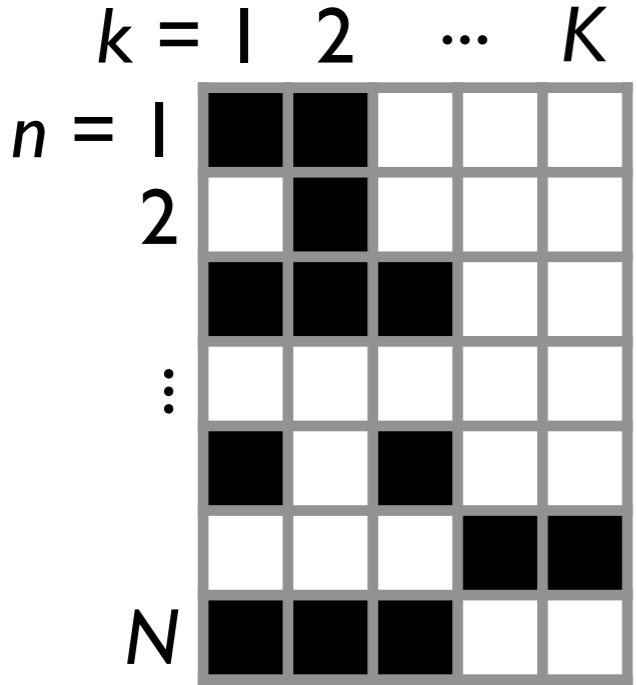
# Example: Indian buffet process



# Example: Indian buffet process



# Example: Indian buffet process



For  $n = 1, 2, \dots, N$

# Example: Indian buffet process

	$k = 1$	2	...	$K$
$n = 1$	■	■		
2		■		
⋮	■	■	■	
	■		■	
				■
				■
$N$	■	■	■	

For  $n = 1, 2, \dots, N$

1. Data point  $n$  has an existing feature  $k$  that has already occurred  $S_{n-1,k}$  times with probability  $\frac{S_{n-1,k}}{\theta + n - 1}$

# Example: Indian buffet process

	$k = 1$	2	...	$K$
$n = 1$	■	■		
2		■		
⋮	■	■	■	
	■		■	
				■
$N$	■	■	■	

For  $n = 1, 2, \dots, N$

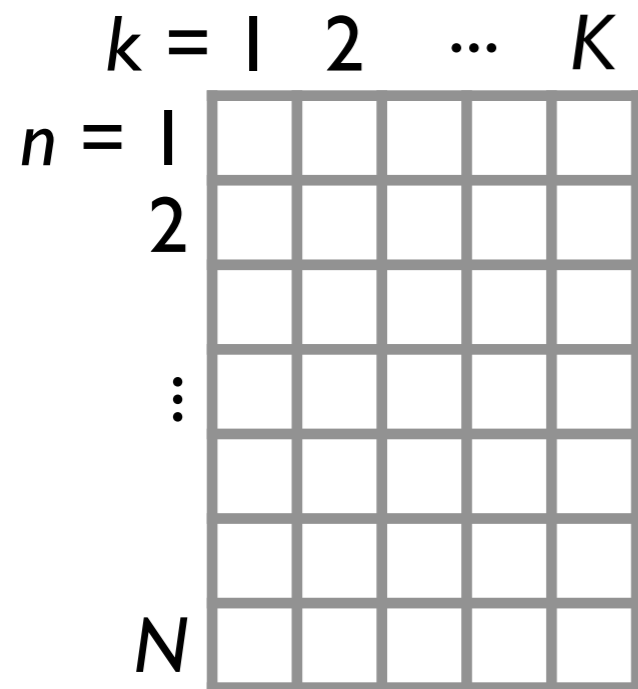
1. Data point  $n$  has an existing feature  $k$  that has already occurred  $S_{n-1,k}$  times with probability

$$\frac{S_{n-1,k}}{\theta + n - 1}$$

2. Number of new features for data

point  $n$ :  $K_n^+ = \text{Poisson} \left( \gamma \frac{\theta}{\theta + n - 1} \right)$

# Example: Indian buffet process



For  $n = 1, 2, \dots, N$

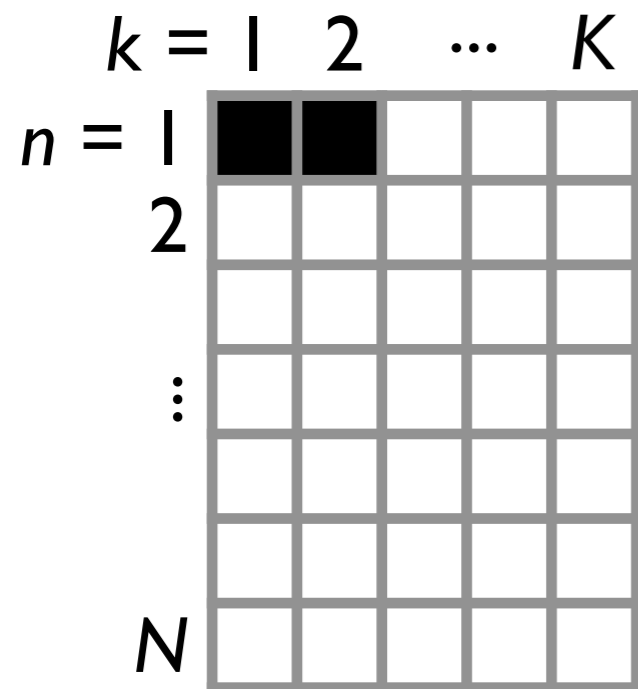
1. Data point  $n$  has an existing feature  $k$  that has already occurred  $S_{n-1,k}$  times with probability

$$\frac{S_{n-1,k}}{\theta + n - 1}$$

2. Number of new features for data

point  $n$ :  $K_n^+ = \text{Poisson} \left( \gamma \frac{\theta}{\theta + n - 1} \right)$

# Example: Indian buffet process



For  $n = 1, 2, \dots, N$

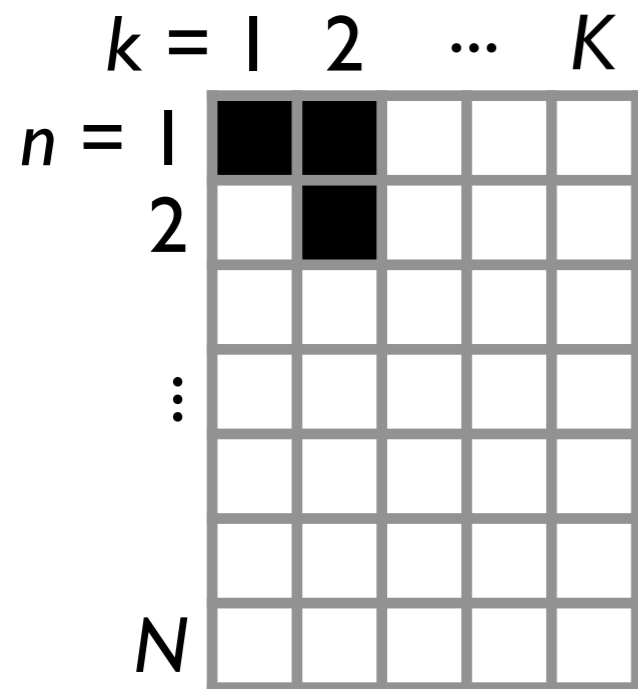
1. Data point  $n$  has an existing feature  $k$  that has already occurred  $S_{n-1,k}$  times with probability

$$\frac{S_{n-1,k}}{\theta + n - 1}$$

2. Number of new features for data

point  $n$ :  $K_n^+ = \text{Poisson} \left( \gamma \frac{\theta}{\theta + n - 1} \right)$

# Example: Indian buffet process



For  $n = 1, 2, \dots, N$

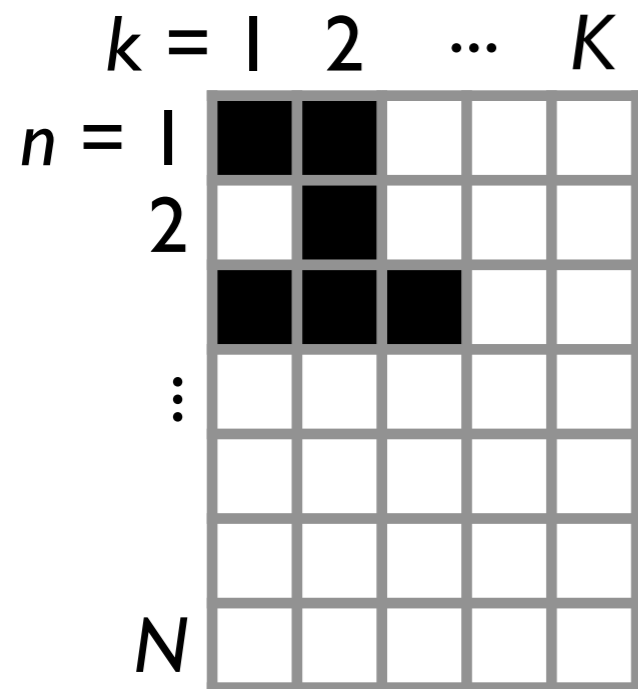
1. Data point  $n$  has an existing feature  $k$  that has already occurred  $S_{n-1,k}$  times with probability

$$\frac{S_{n-1,k}}{\theta + n - 1}$$

2. Number of new features for data

point  $n$ :  $K_n^+ = \text{Poisson} \left( \gamma \frac{\theta}{\theta + n - 1} \right)$

# Example: Indian buffet process



For  $n = 1, 2, \dots, N$

1. Data point  $n$  has an existing feature

$k$  that has already occurred  $S_{n-1,k}$

times with probability  $\frac{S_{n-1,k}}{\theta + n - 1}$

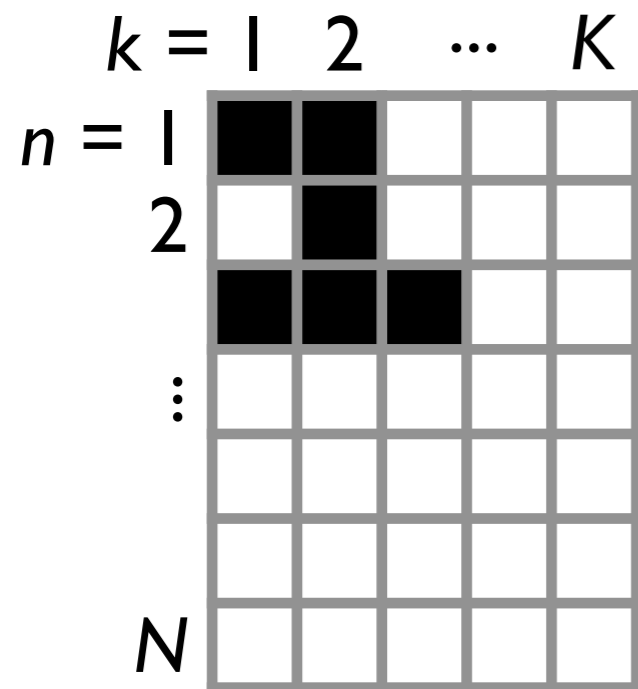
$$\frac{S_{n-1,k}}{\theta + n - 1}$$

2. Number of new features for data

point  $n$ :  $K_n^+ = \text{Poisson} \left( \gamma \frac{\theta}{\theta + n - 1} \right)$



# Example: Indian buffet process



For  $n = 1, 2, \dots, N$

1. Data point  $n$  has an existing feature

$k$  that has already occurred  $S_{n-1,k}$

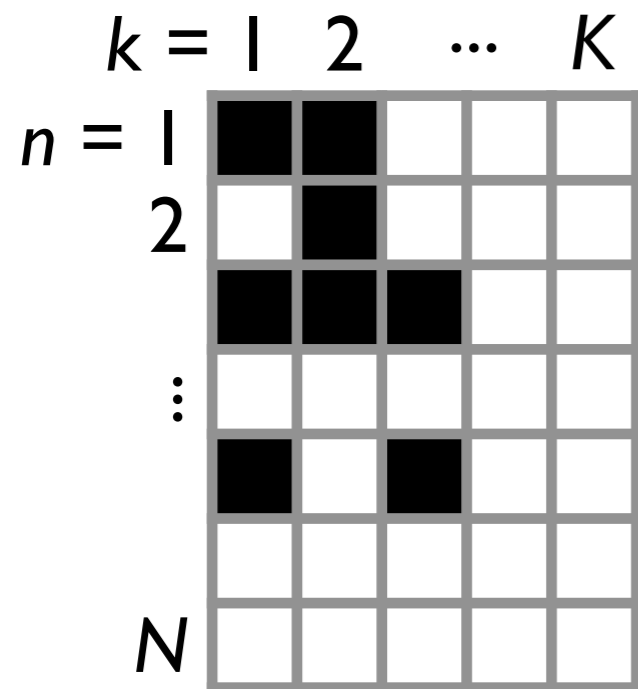
times with probability  $\frac{S_{n-1,k}}{\theta + n - 1}$

$$\frac{S_{n-1,k}}{\theta + n - 1}$$

2. Number of new features for data

point  $n$ :  $K_n^+ = \text{Poisson} \left( \gamma \frac{\theta}{\theta + n - 1} \right)$

# Example: Indian buffet process



For  $n = 1, 2, \dots, N$

1. Data point  $n$  has an existing feature

$k$  that has already occurred  $S_{n-1,k}$

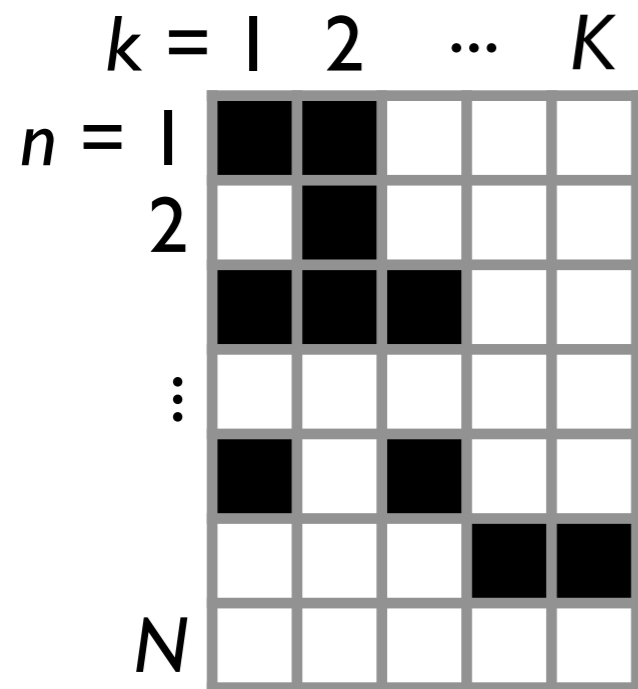
times with probability  $\frac{S_{n-1,k}}{\theta + n - 1}$

$$\frac{S_{n-1,k}}{\theta + n - 1}$$

2. Number of new features for data

point  $n$ :  $K_n^+ = \text{Poisson} \left( \gamma \frac{\theta}{\theta + n - 1} \right)$

# Example: Indian buffet process



For  $n = 1, 2, \dots, N$

1. Data point  $n$  has an existing feature  $k$  that has already occurred  $S_{n-1,k}$  times with probability

$$\frac{S_{n-1,k}}{\theta + n - 1}$$

2. Number of new features for data

point  $n$ :  $K_n^+ = \text{Poisson} \left( \gamma \frac{\theta}{\theta + n - 1} \right)$

# Example: Indian buffet process

	$k = 1$	2	...	$K$
$n = 1$	■	■		
2		■		
⋮	■	■	■	
	■		■	
				■
$N$	■	■	■	

For  $n = 1, 2, \dots, N$

1. Data point  $n$  has an existing feature  $k$  that has already occurred  $S_{n-1,k}$  times with probability

$$\frac{S_{n-1,k}}{\theta + n - 1}$$

2. Number of new features for data

point  $n$ :  $K_n^+ = \text{Poisson} \left( \gamma \frac{\theta}{\theta + n - 1} \right)$

# Exchangeable probability functions

“Exchangeable feature probability function” (EFPF)?

# Exchangeable probability functions

“Exchangeable feature probability function” (EFPF)?

Example: Indian buffet process (IBP)

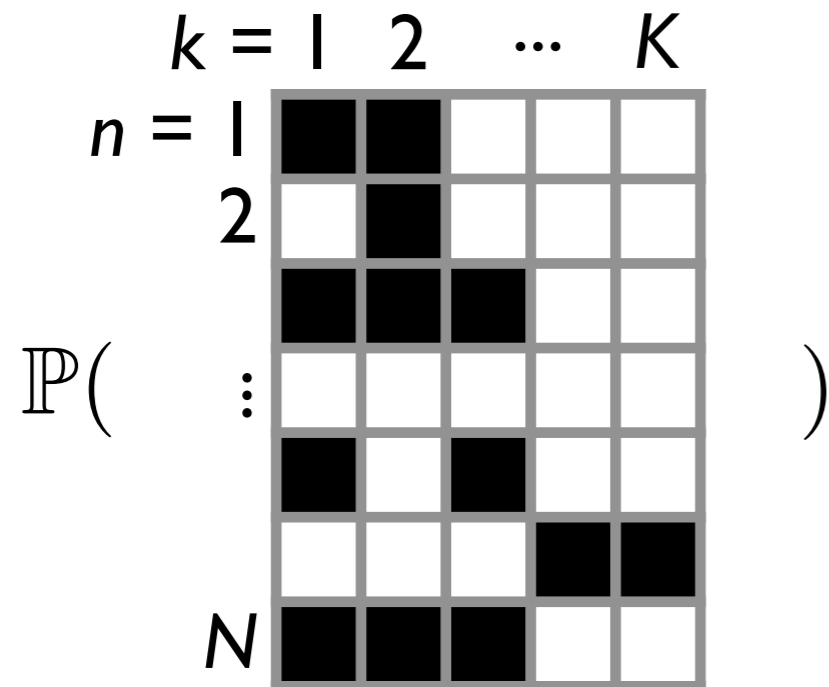
	$k = 1$	$2$	$\dots$	$K$
$n = 1$	■	■	□	□
$2$	□	■	□	□
$\vdots$	■	■	■	□
$\vdots$	□	□	□	□
$\vdots$	■	□	■	□
$\vdots$	□	□	□	■
$N$	■	■	■	□

$\mathbb{P}(\quad)$

# Exchangeable probability functions

“Exchangeable feature probability function” (EFPF)?

Example: Indian buffet process (IBP)

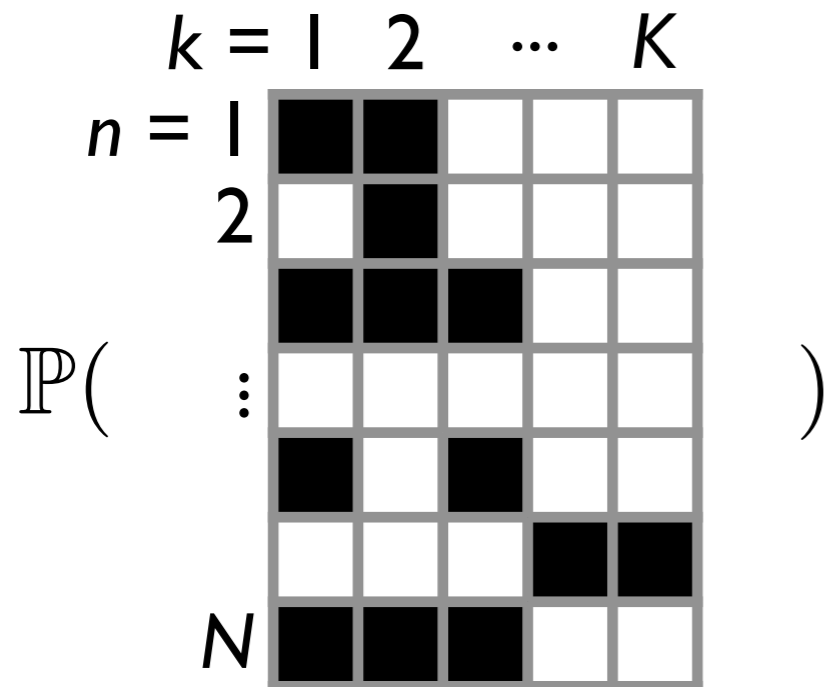


$$= \frac{1}{K_N!} (\theta \gamma)^{K_N} \exp \left( -\theta \gamma \sum_{n=1}^N (\theta + n - 1)^{-1} \right) \prod_{k=1}^{K_N} \frac{\Gamma(S_{N,k}) \Gamma(N - S_{N,k} + \theta)}{\Gamma(N + \theta)}$$

# Exchangeable probability functions

“Exchangeable feature probability function” (EFPF)?

Example: Indian buffet process (IBP)



Size of  $k$ th feature



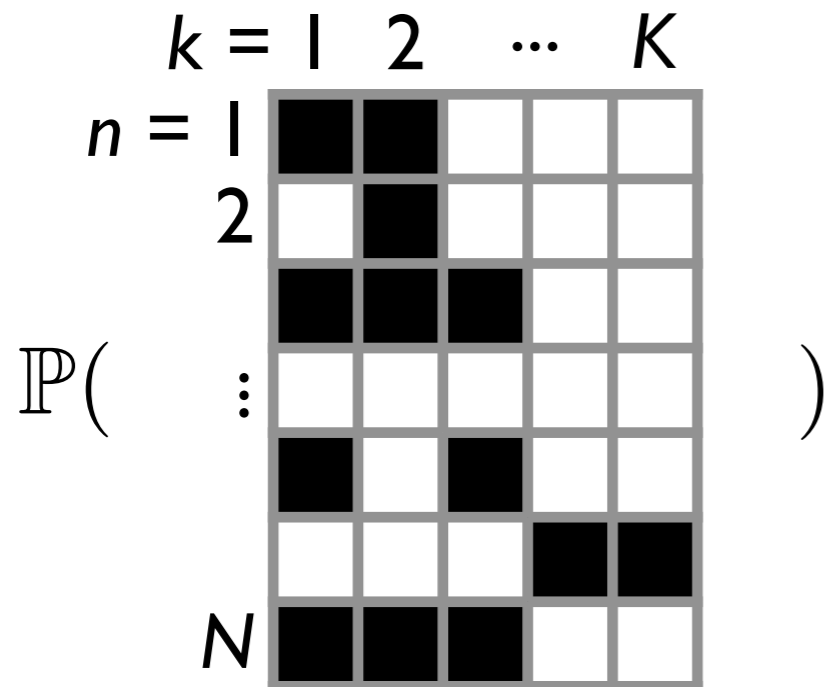
$$= \frac{1}{K_N!} (\theta \gamma)^{K_N} \exp \left( -\theta \gamma \sum_{n=1}^N (\theta + n - 1)^{-1} \right) \prod_{k=1}^{K_N} \frac{\Gamma(S_{N,k}) \Gamma(N - S_{N,k} + \theta)}{\Gamma(N + \theta)}$$



# Exchangeable probability functions

“Exchangeable feature probability function” (EFPF)?

Example: Indian buffet process (IBP)



Size of  $k$ th feature

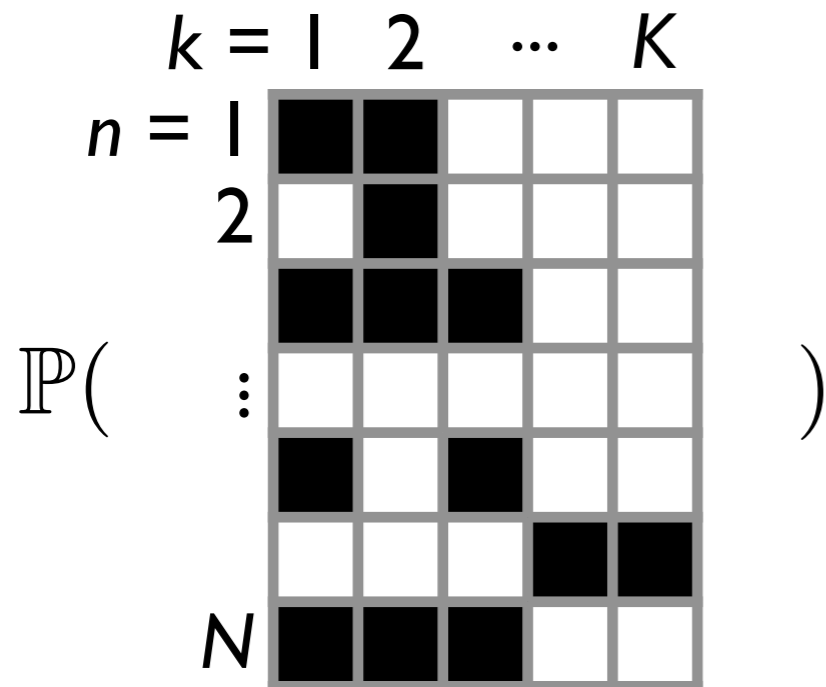
Number of features

$$= \frac{1}{K_N!} (\theta \gamma)^{K_N} \exp \left( -\theta \gamma \sum_{n=1}^N (\theta + n - 1)^{-1} \right) \prod_{k=1}^{K_N} \frac{\Gamma(S_{N,k}) \Gamma(N - S_{N,k} + \theta)}{\Gamma(N + \theta)}$$

# Exchangeable probability functions

“Exchangeable feature probability function” (EFPF)?

Example: Indian buffet process (IBP)



Number of data points

Size of  $k$ th feature

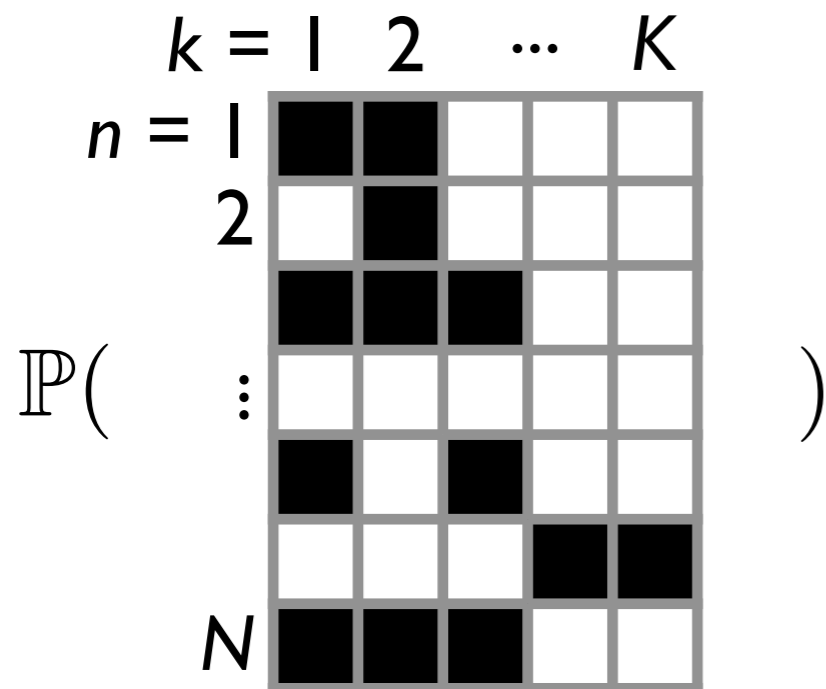
Number of features

$$= \frac{1}{K_N!} (\theta \gamma)^{K_N} \exp \left( -\theta \gamma \sum_{n=1}^N (\theta + n - 1)^{-1} \right) \prod_{k=1}^{K_N} \frac{\Gamma(S_{N,k}) \Gamma(N - S_{N,k} + \theta)}{\Gamma(N + \theta)}$$

# Exchangeable probability functions

“Exchangeable feature probability function” (EFPF)?

Example: Indian buffet process (IBP)



Number of data points

Size of  $k$ th feature

Number of features

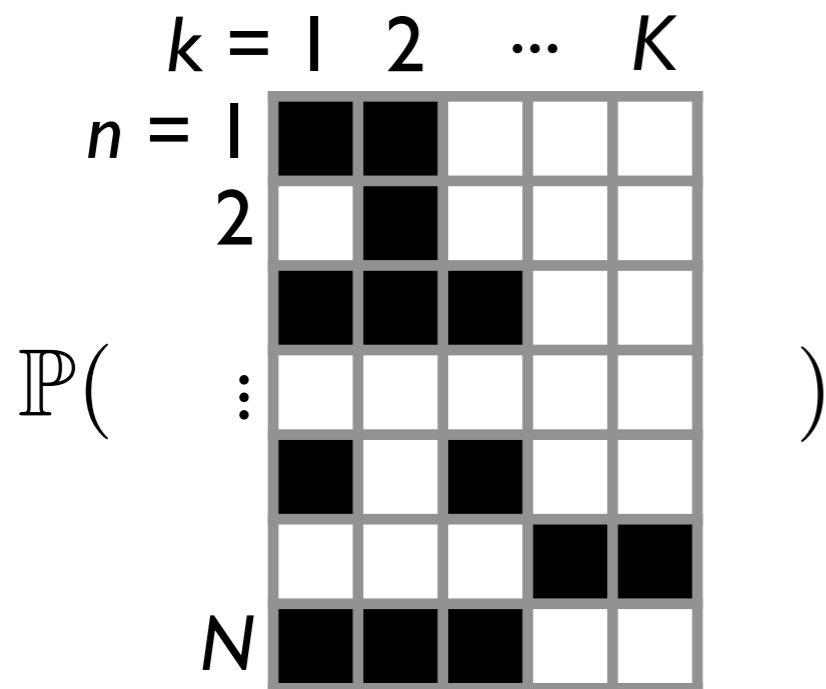
$$= \frac{1}{K_N!} (\theta \gamma)^{K_N} \exp \left( -\theta \gamma \sum_{n=1}^N (\theta + n - 1)^{-1} \right) \prod_{k=1}^{K_N} \frac{\Gamma(S_{N,k}) \Gamma(N - S_{N,k} + \theta)}{\Gamma(N + \theta)}$$

$$= p(N; S_{N,1}, S_{N,2}, \dots, S_{N,K})$$

# Exchangeable probability functions

“Exchangeable feature probability function” (EFPF)?

Example: Indian buffet process (IBP)



Number of data points

Size of  $k$ th feature

Number of features

$$= \frac{1}{K_N!} (\theta\gamma)^{K_N} \exp\left(-\theta\gamma \sum_{n=1}^N (\theta + n - 1)^{-1}\right) \prod_{k=1}^{K_N} \frac{\Gamma(S_{N,k})\Gamma(N - S_{N,k} + \theta)}{\Gamma(N + \theta)}$$

$$= p(N; S_{N,1}, S_{N,2}, \dots, S_{N,K})$$

“EFPF”

# Exchangeable probability functions

“Exchangeable feature probability function” (EFPF)?

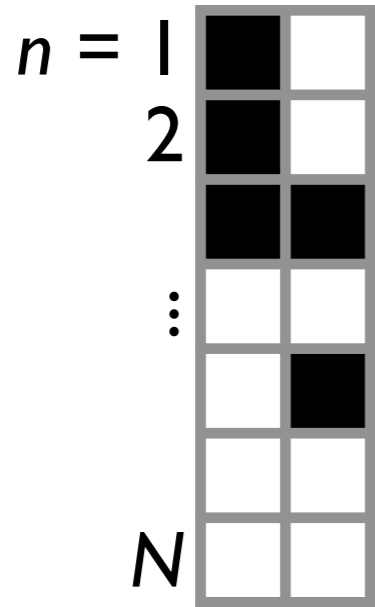
Counterexample

$n = 1$	■	□
2	■	□
	■	■
⋮	□	□
	□	■
	□	□
$N$	□	□

# Exchangeable probability functions

“Exchangeable feature probability function” (EFPF)?

Counterexample



$$\mathbb{P}(\text{row} = \begin{array}{|c|c|} \hline \blacksquare & \square \\ \hline \end{array}) = p_1$$

$$\mathbb{P}(\text{row} = \begin{array}{|c|c|} \hline \square & \blacksquare \\ \hline \end{array}) = p_2$$

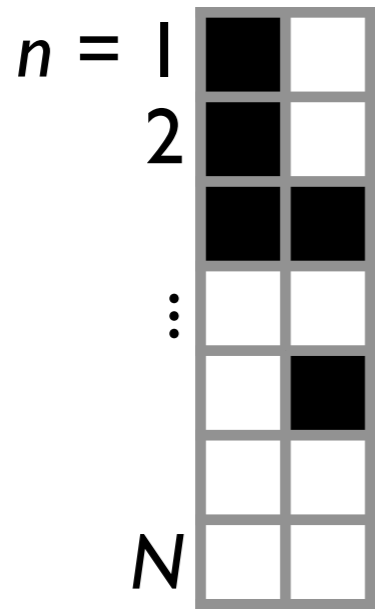
$$\mathbb{P}(\text{row} = \begin{array}{|c|c|} \hline \blacksquare & \blacksquare \\ \hline \end{array}) = p_3$$

$$\mathbb{P}(\text{row} = \begin{array}{|c|c|} \hline \square & \square \\ \hline \end{array}) = p_4$$

# Exchangeable probability functions

“Exchangeable feature probability function” (EFPF)?

Counterexample

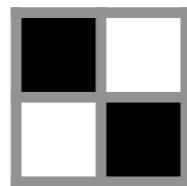


$$\mathbb{P}(\text{row} = \begin{array}{|c|c|} \hline \blacksquare & \square \\ \hline \end{array}) = p_1$$

$$\mathbb{P}(\text{row} = \begin{array}{|c|c|} \hline \square & \blacksquare \\ \hline \end{array}) = p_2$$

$$\mathbb{P}(\text{row} = \begin{array}{|c|c|} \hline \blacksquare & \blacksquare \\ \hline \end{array}) = p_3$$

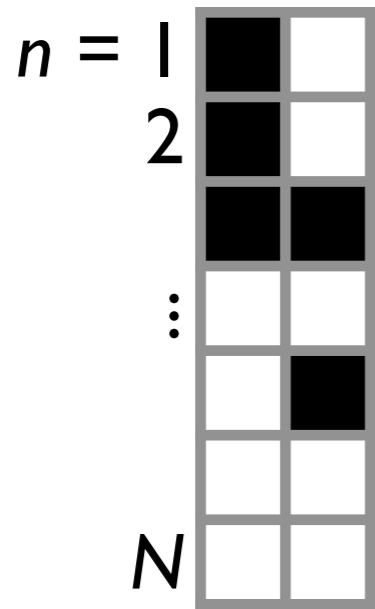
$$\mathbb{P}(\text{row} = \begin{array}{|c|c|} \hline \square & \square \\ \hline \end{array}) = p_4$$



# Exchangeable probability functions

“Exchangeable feature probability function” (EFPF)?

Counterexample



$$\mathbb{P}(\text{row} = \begin{array}{|c|c|} \hline \blacksquare & \square \\ \hline \end{array}) = p_1$$

$$\mathbb{P}(\text{row} = \begin{array}{|c|c|} \hline \square & \blacksquare \\ \hline \end{array}) = p_2$$

$$\mathbb{P}(\text{row} = \begin{array}{|c|c|} \hline \blacksquare & \blacksquare \\ \hline \end{array}) = p_3$$

$$\mathbb{P}(\text{row} = \begin{array}{|c|c|} \hline \square & \square \\ \hline \end{array}) = p_4$$

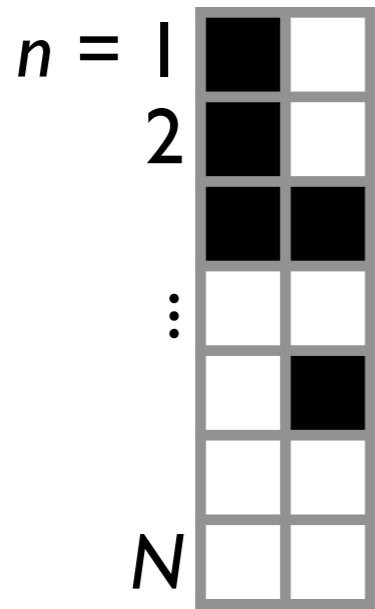
$$\mathbb{P}\left(\begin{array}{|c|c|} \hline \blacksquare & \square \\ \hline \square & \blacksquare \\ \hline \end{array}\right) \quad \mathbb{P}\left(\begin{array}{|c|c|} \hline \blacksquare & \blacksquare \\ \hline \square & \square \\ \hline \end{array}\right)$$



# Exchangeable probability functions

“Exchangeable feature probability function” (EFPF)?

Counterexample



$$\mathbb{P}(\text{row} = \begin{array}{|c|c|} \hline \blacksquare & \square \\ \hline \end{array}) = p_1$$

$$\mathbb{P}(\text{row} = \begin{array}{|c|c|} \hline \square & \blacksquare \\ \hline \end{array}) = p_2$$

$$\mathbb{P}(\text{row} = \begin{array}{|c|c|} \hline \blacksquare & \blacksquare \\ \hline \end{array}) = p_3$$

$$\mathbb{P}(\text{row} = \begin{array}{|c|c|} \hline \square & \square \\ \hline \end{array}) = p_4$$

$$\mathbb{P}(\begin{array}{|c|c|} \hline \blacksquare & \square \\ \hline \square & \blacksquare \\ \hline \end{array})$$

$$p_1 p_2$$

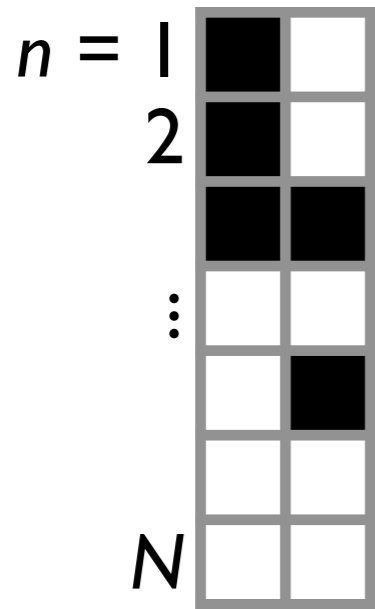
$$\mathbb{P}(\begin{array}{|c|c|} \hline \blacksquare & \blacksquare \\ \hline \square & \square \\ \hline \end{array})$$

$$p_3 p_4$$

# Exchangeable probability functions

“Exchangeable feature probability function” (EFPF)?

Counterexample



$$\mathbb{P}(\text{row} = \begin{array}{|c|c|} \hline \blacksquare & \square \\ \hline \end{array}) = p_1$$

$$\mathbb{P}(\text{row} = \begin{array}{|c|c|} \hline \square & \blacksquare \\ \hline \end{array}) = p_2$$

$$\mathbb{P}(\text{row} = \begin{array}{|c|c|} \hline \blacksquare & \blacksquare \\ \hline \end{array}) = p_3$$

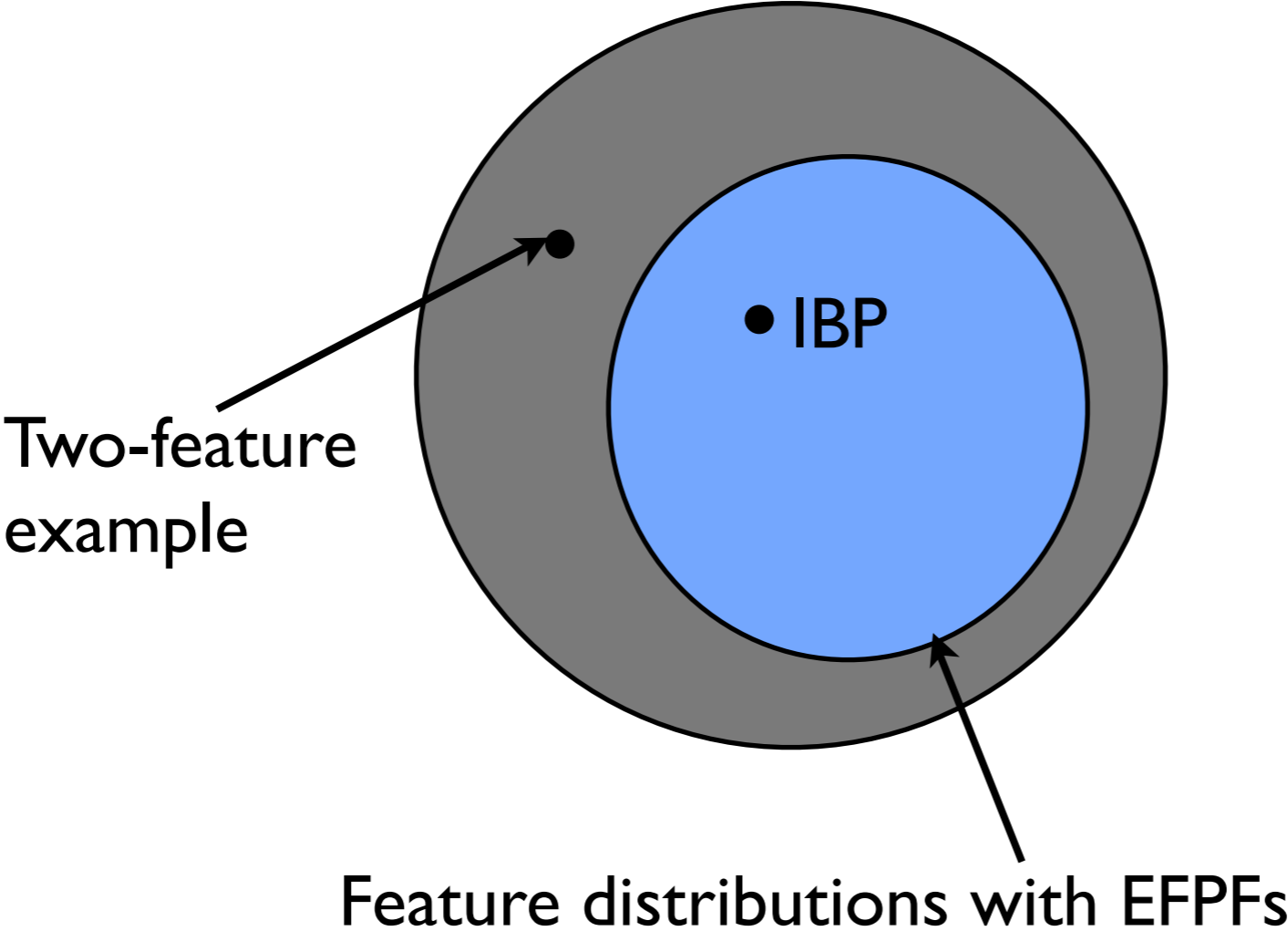
$$\mathbb{P}(\text{row} = \begin{array}{|c|c|} \hline \square & \square \\ \hline \end{array}) = p_4$$

$$\mathbb{P}\left(\begin{array}{|c|c|} \hline \blacksquare & \square \\ \hline \square & \blacksquare \\ \hline \end{array}\right) \neq \mathbb{P}\left(\begin{array}{|c|c|} \hline \blacksquare & \blacksquare \\ \hline \square & \square \\ \hline \end{array}\right)$$

$$p_1 p_2 \neq p_3 p_4$$

# Exchangeable probability functions

Exchangeable feature distributions



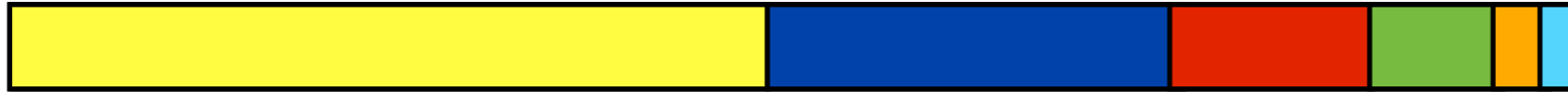
# Paintboxes

Exchangeable partition: Kingman paintbox



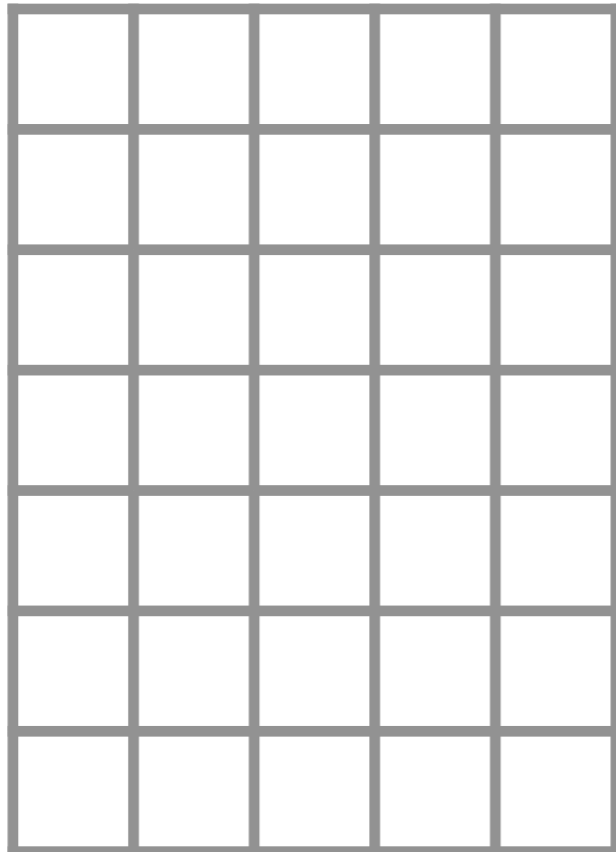
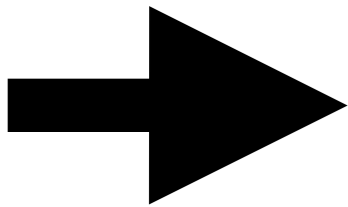
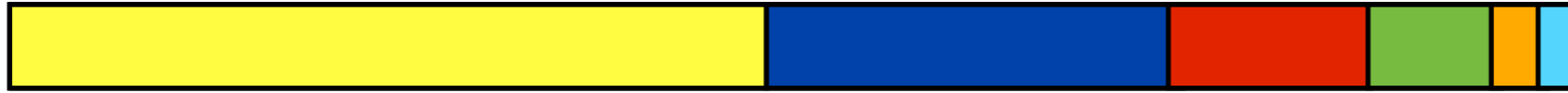
# Paintboxes

Exchangeable partition: Kingman paintbox



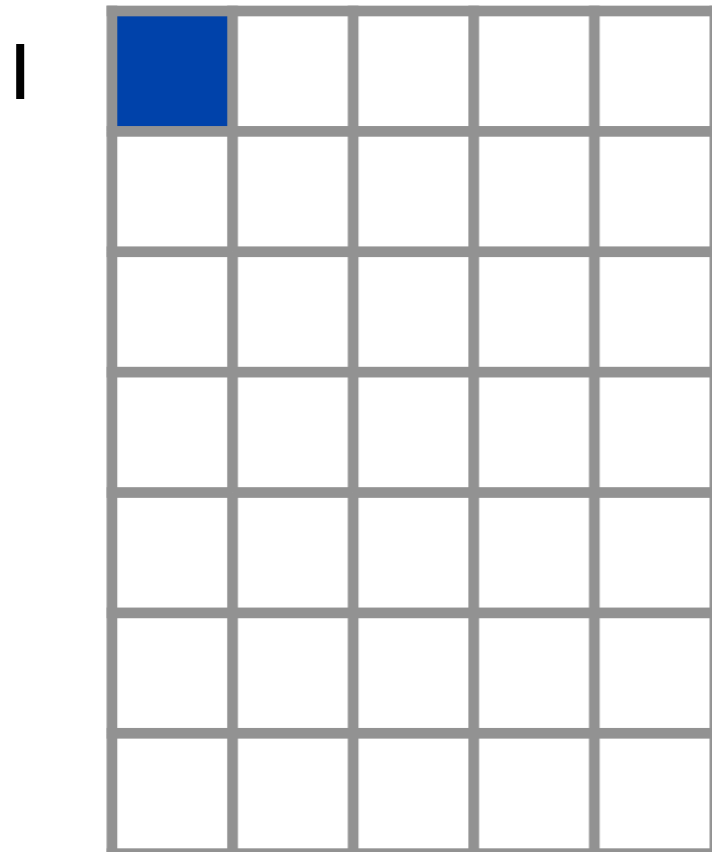
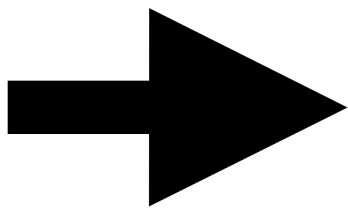
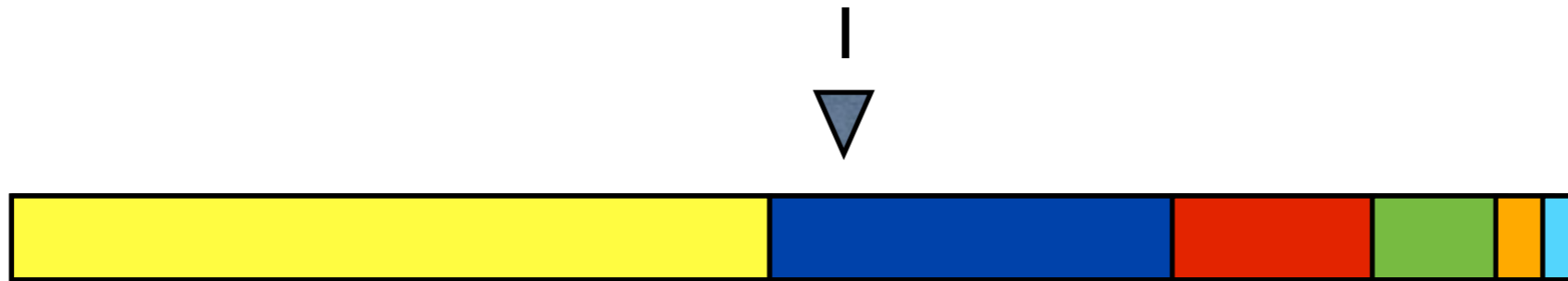
# Paintboxes

Exchangeable partition: Kingman paintbox



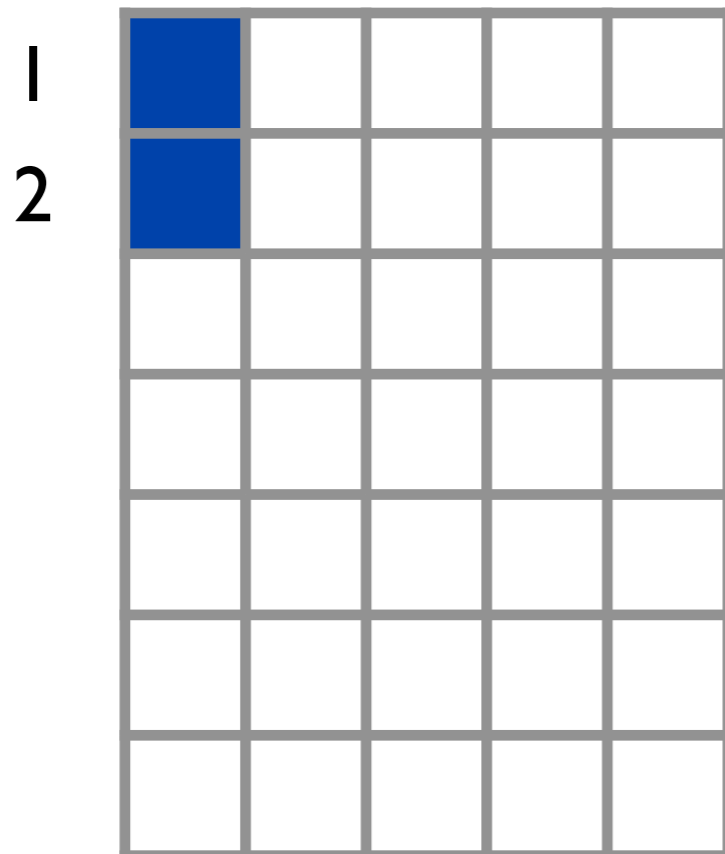
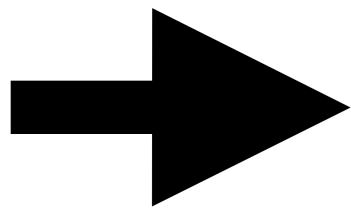
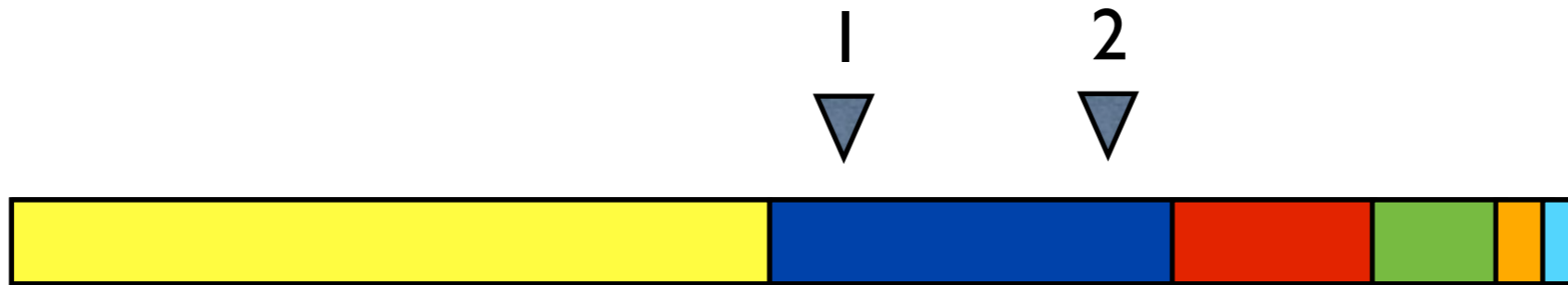
# Paintboxes

Exchangeable partition: Kingman paintbox



# Paintboxes

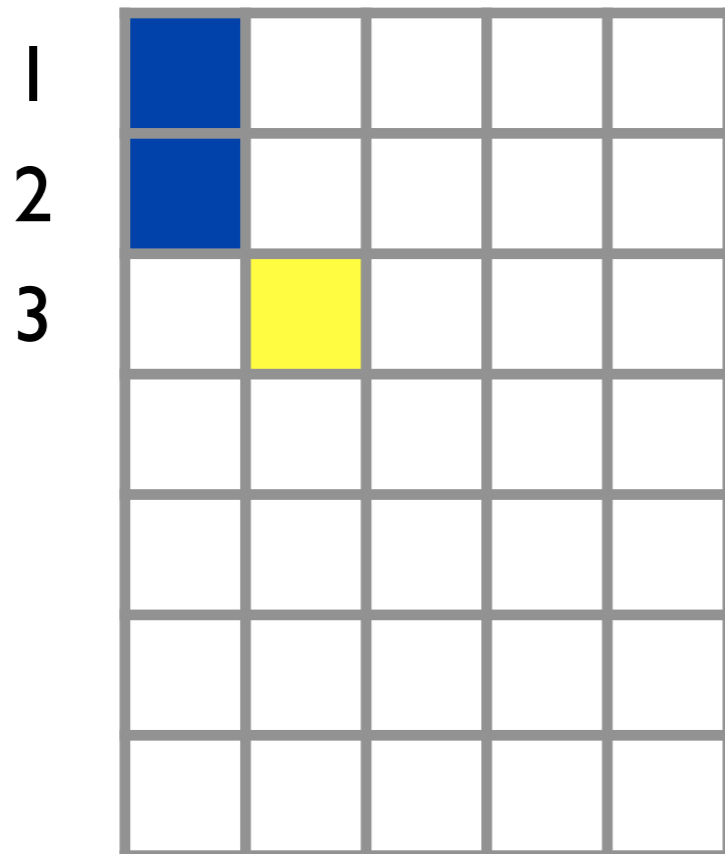
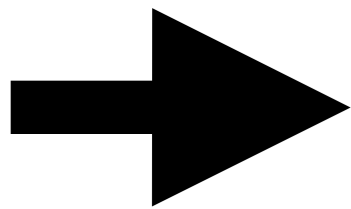
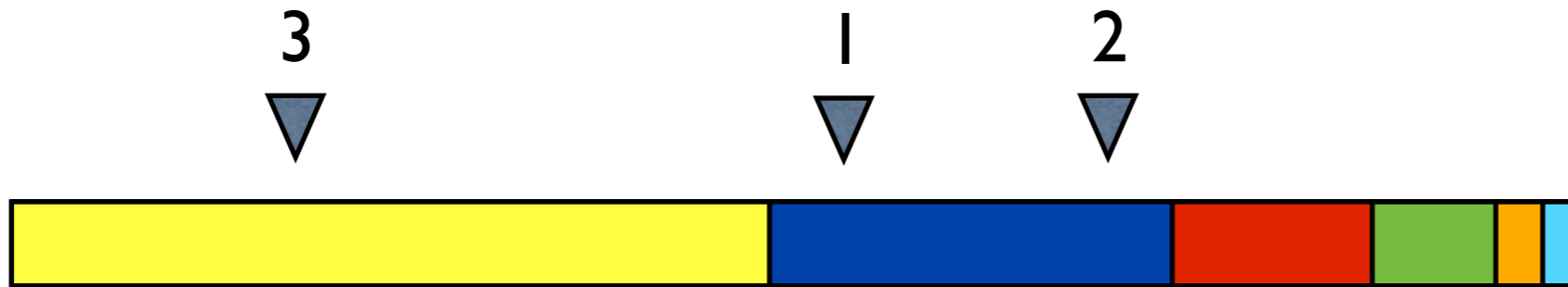
Exchangeable partition: Kingman paintbox





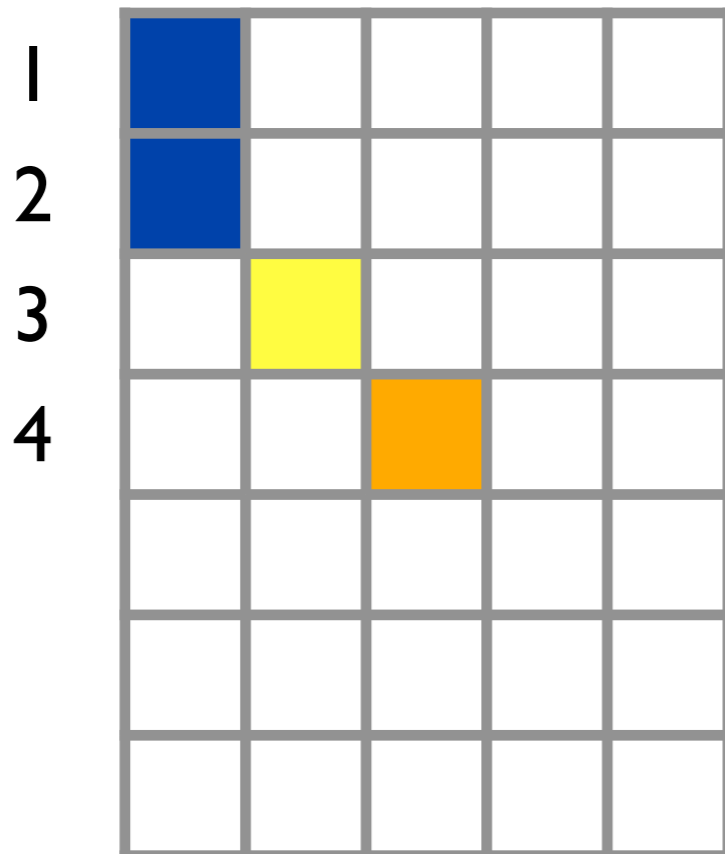
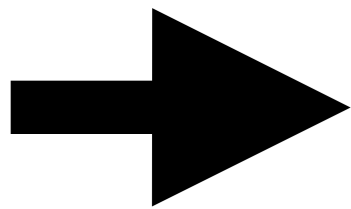
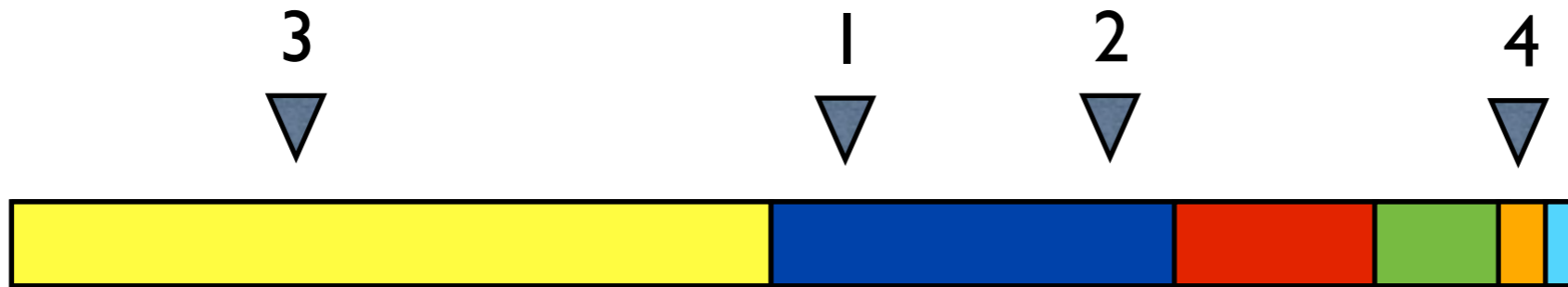
# Paintboxes

Exchangeable partition: Kingman paintbox



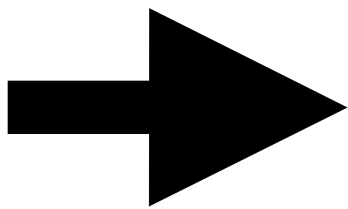
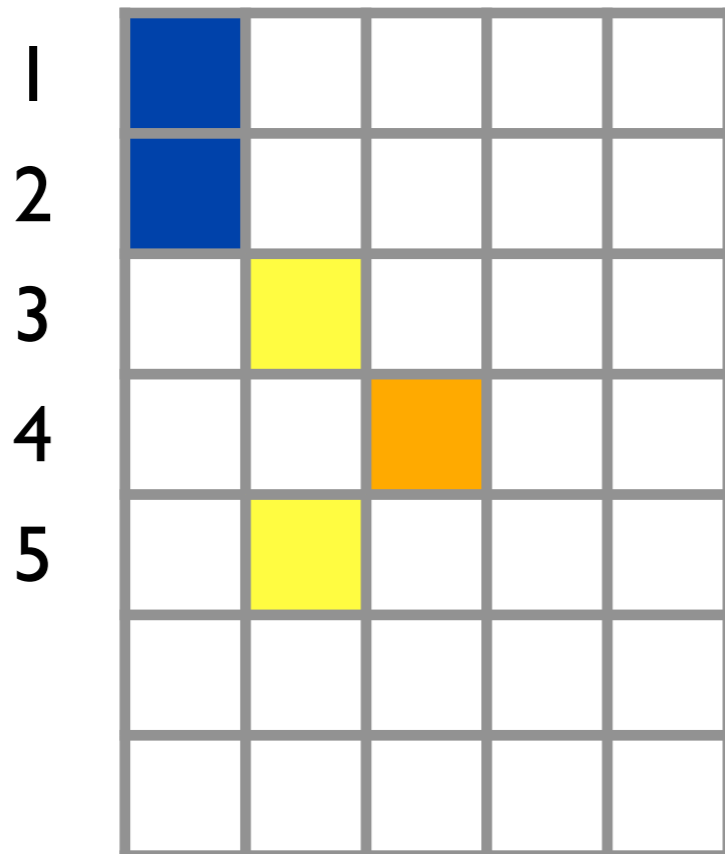
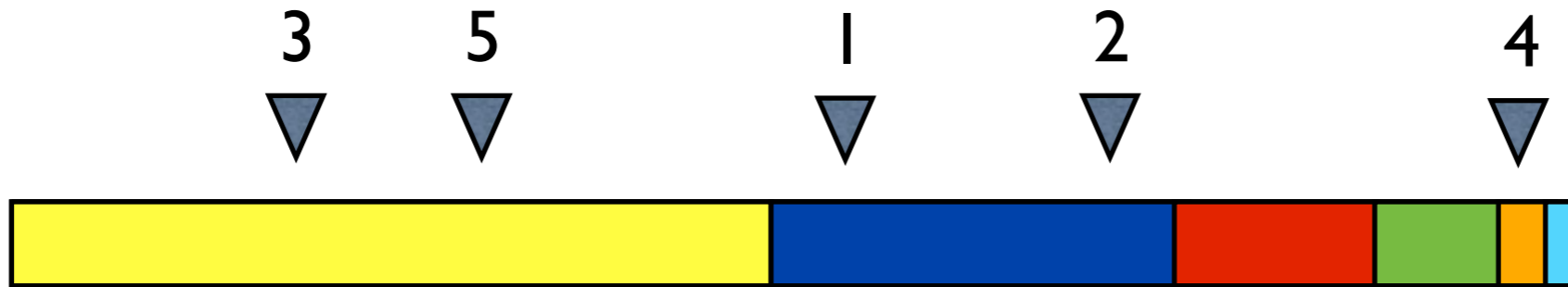
# Paintboxes

Exchangeable partition: Kingman paintbox



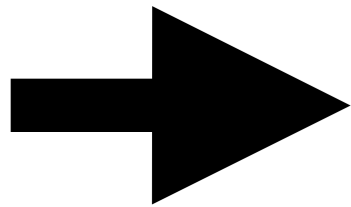
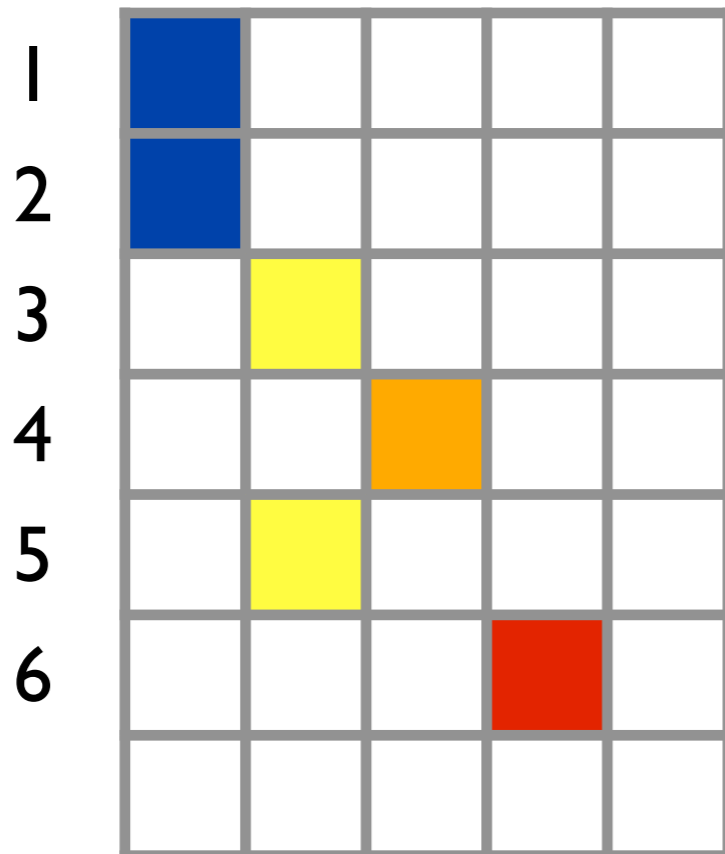
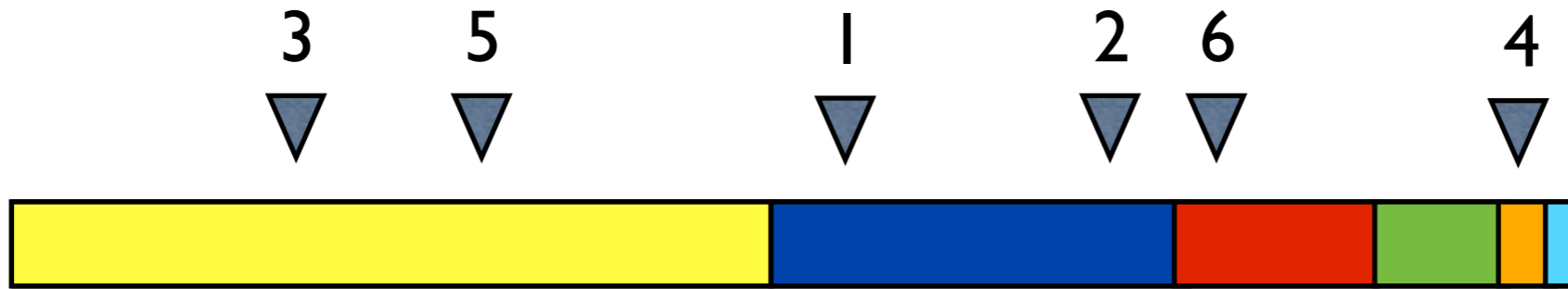
# Paintboxes

Exchangeable partition: Kingman paintbox



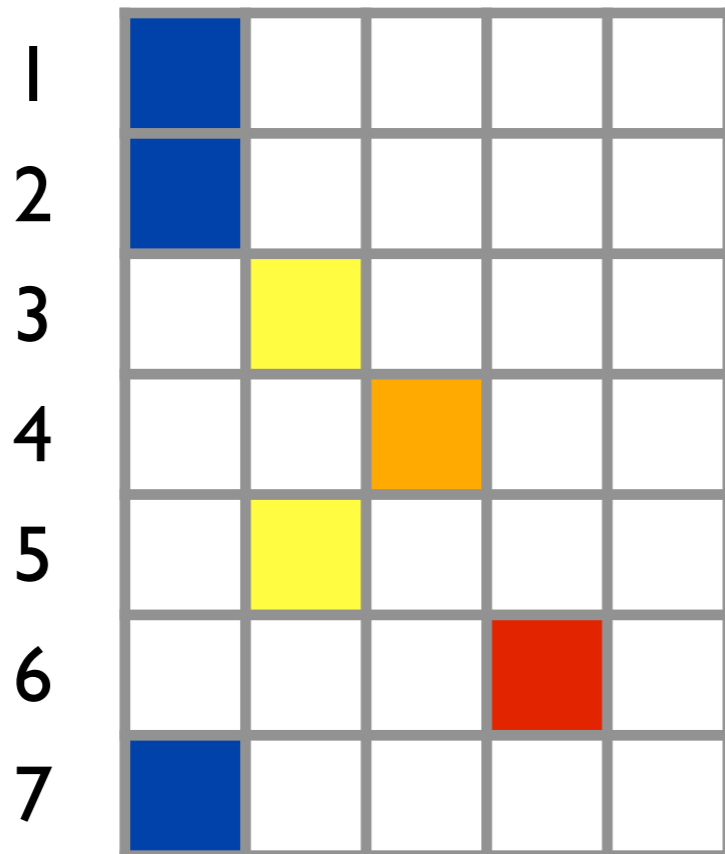
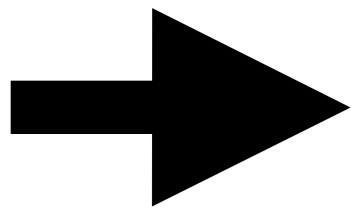
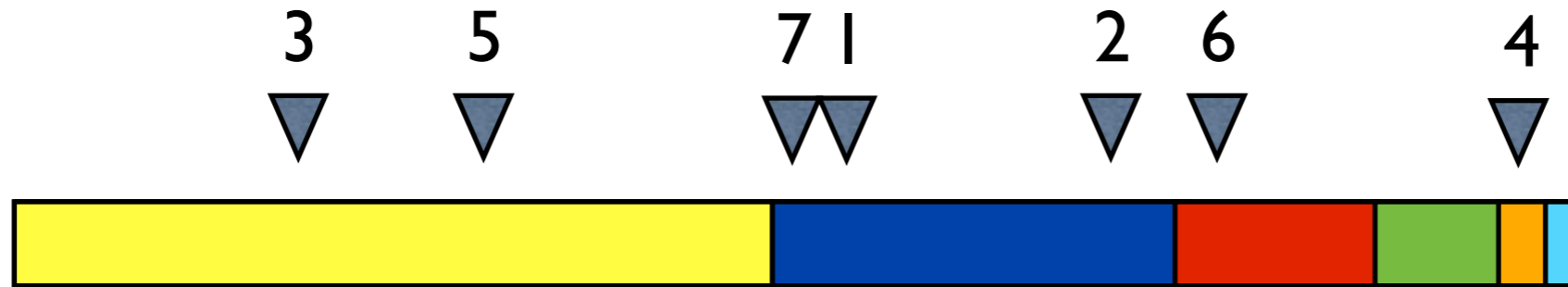
# Paintboxes

Exchangeable partition: Kingman paintbox



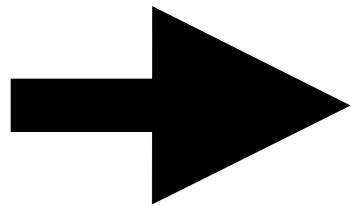
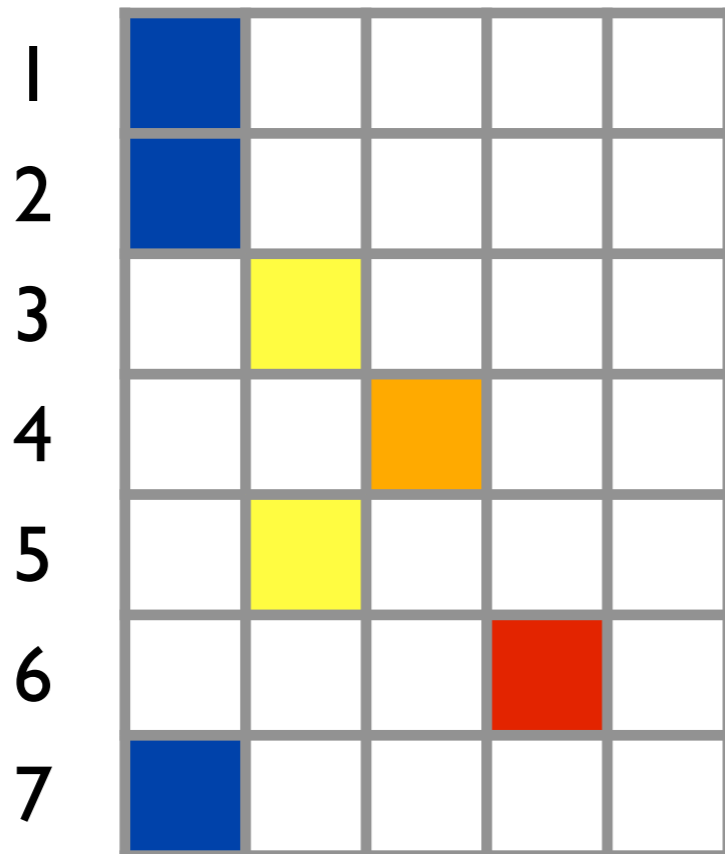
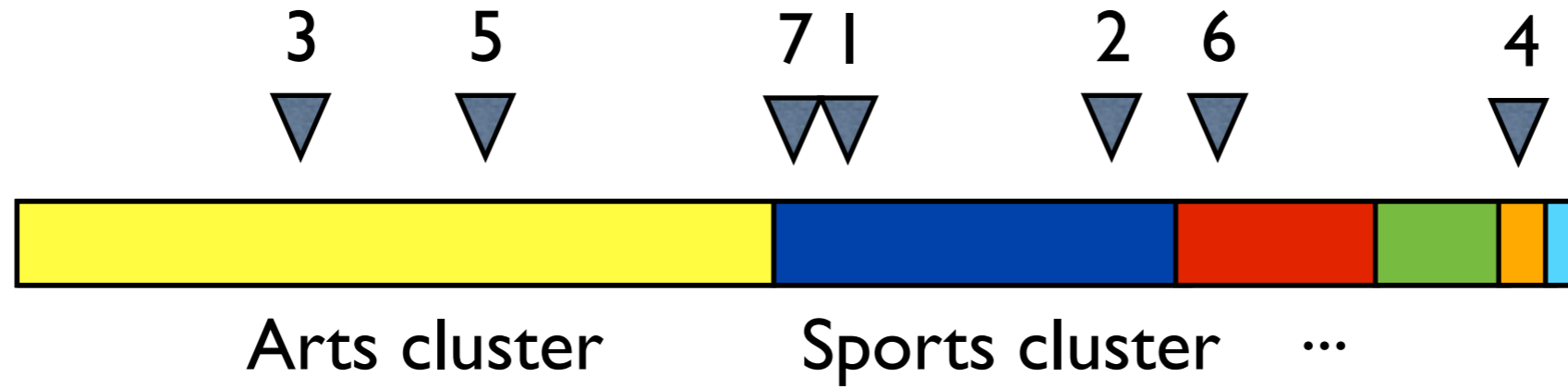
# Paintboxes

Exchangeable partition: Kingman paintbox



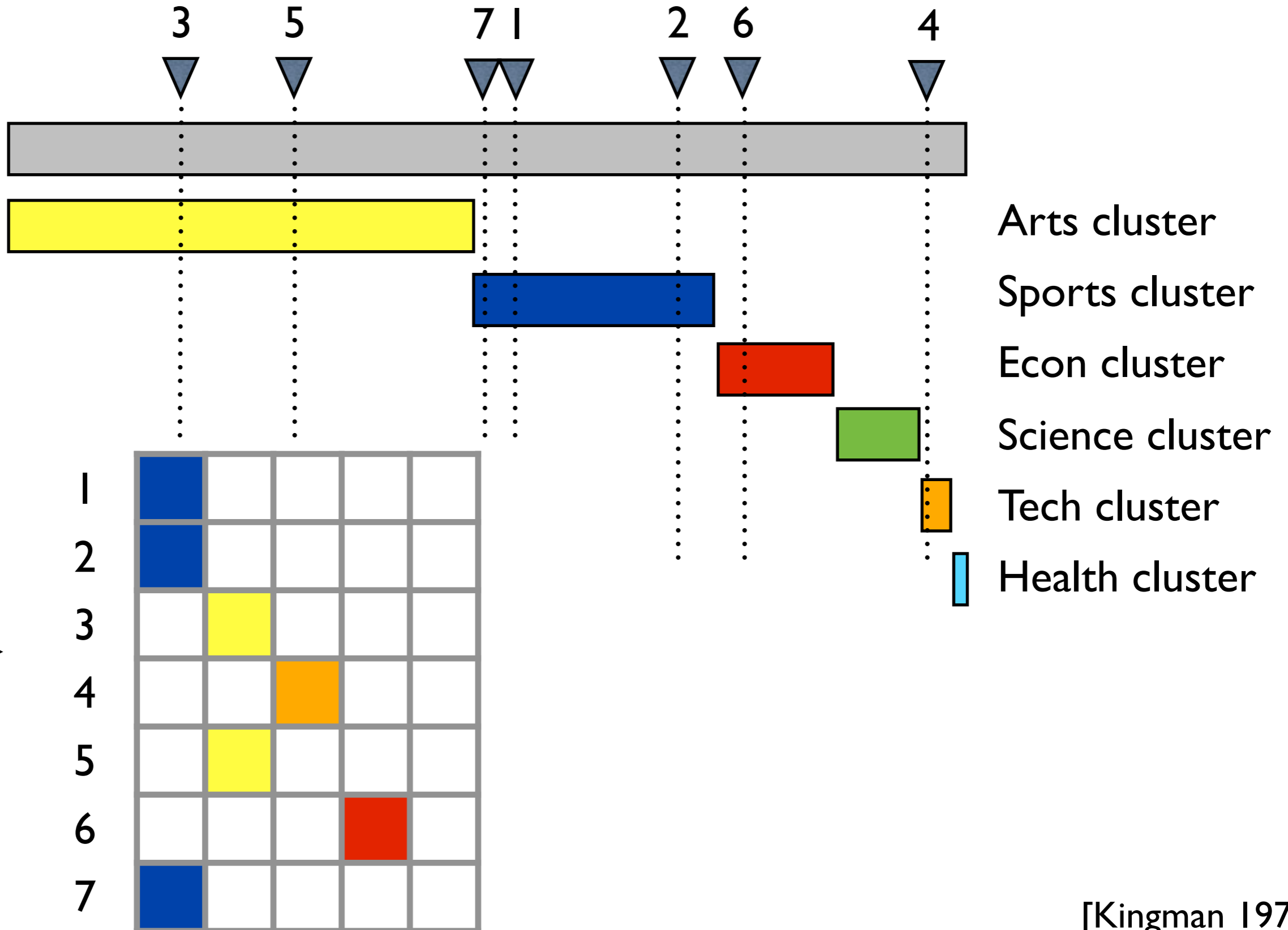
# Paintboxes

Exchangeable partition: Kingman paintbox

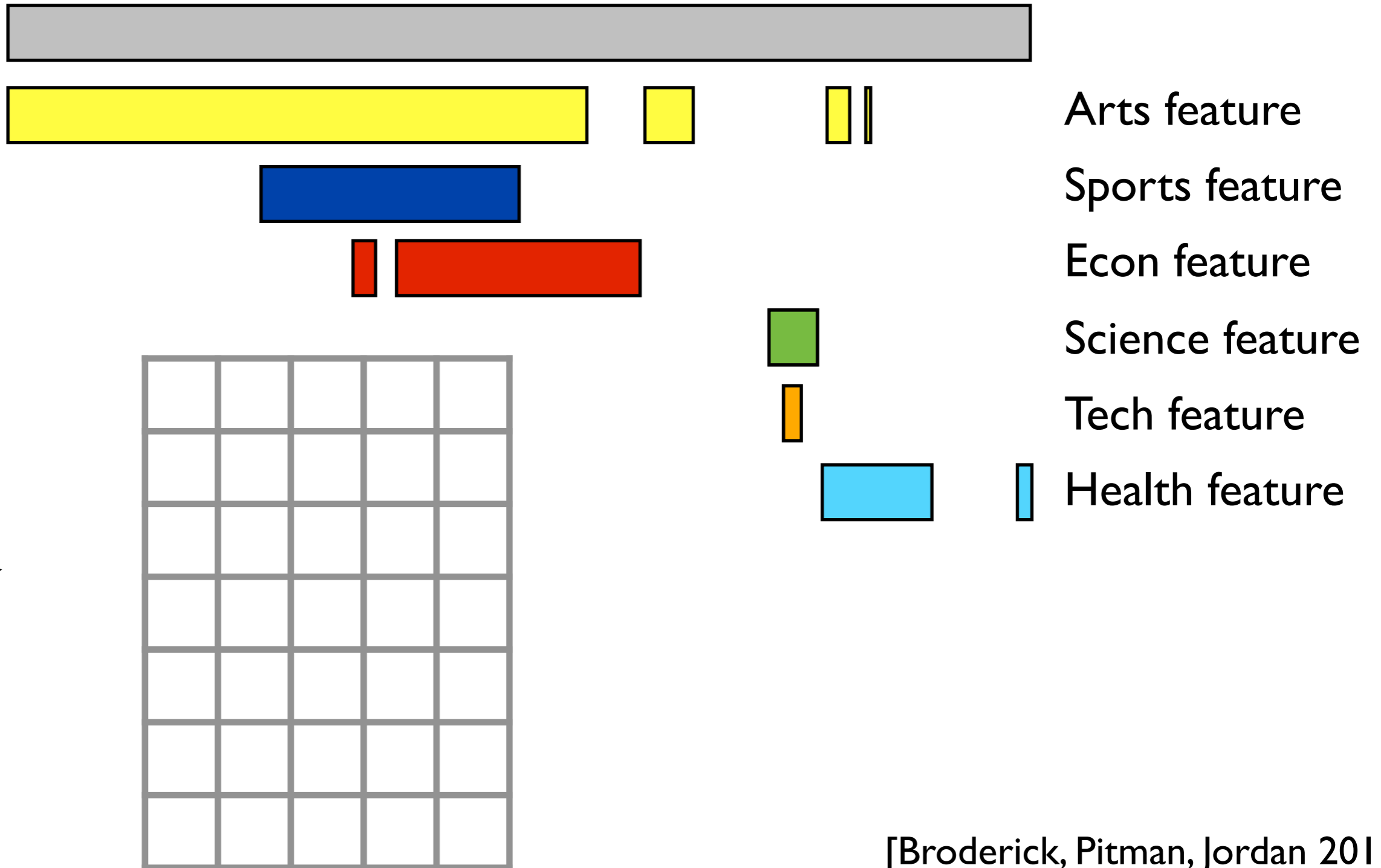


# Paintboxes

Exchangeable partition: Kingman paintbox



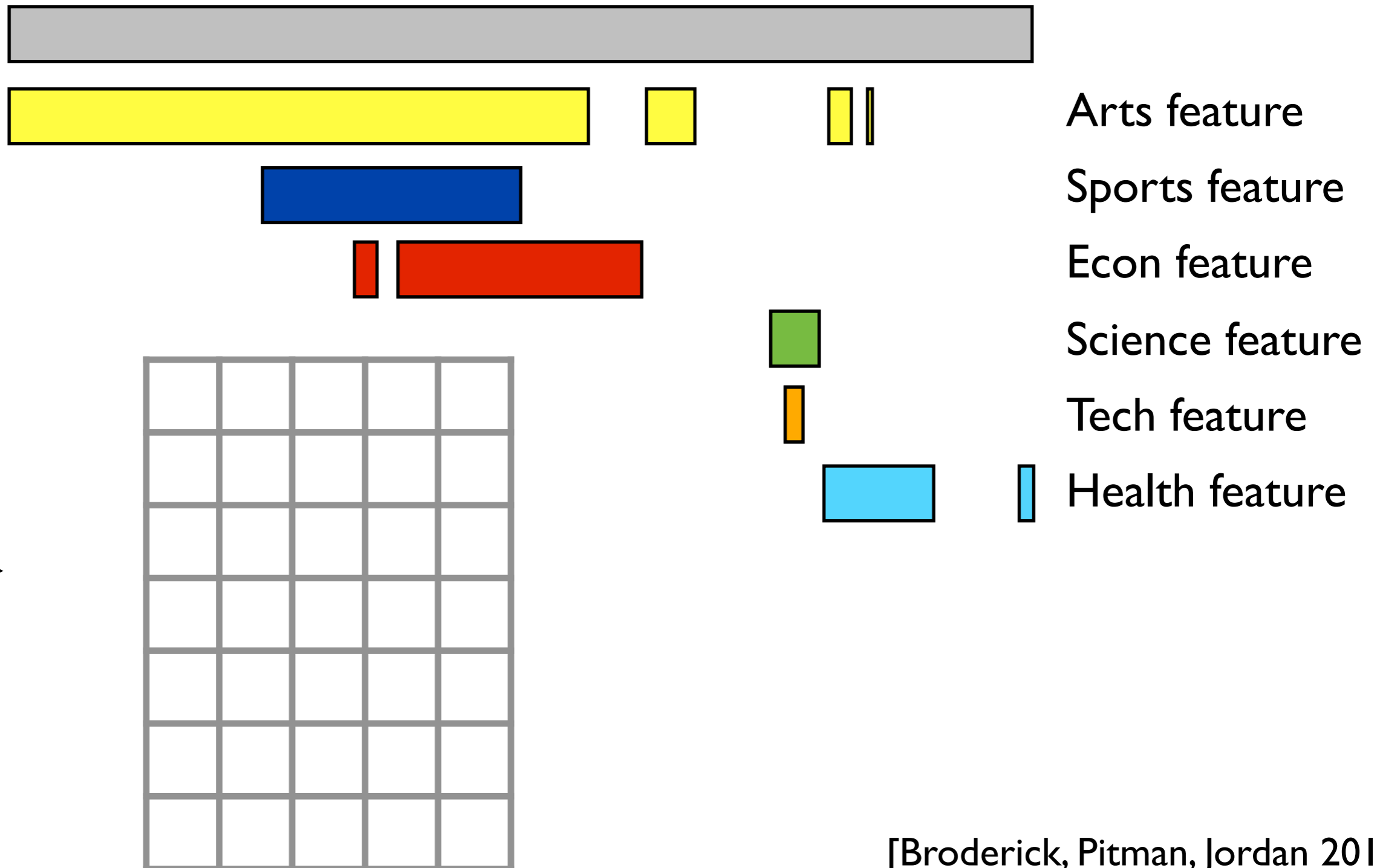
# Paintboxes





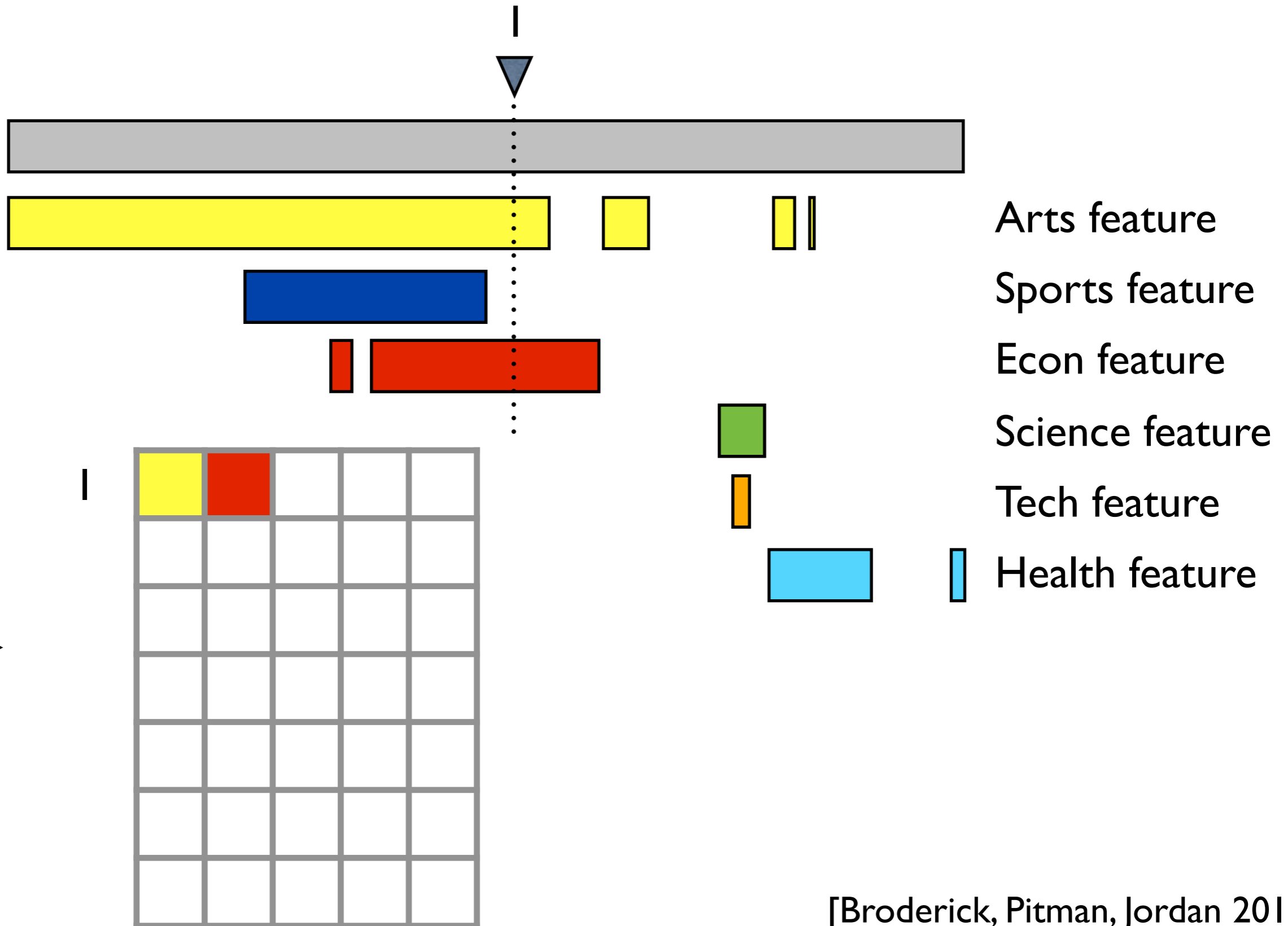
# Paintboxes

Exchangeable feature allocation: feature paintbox



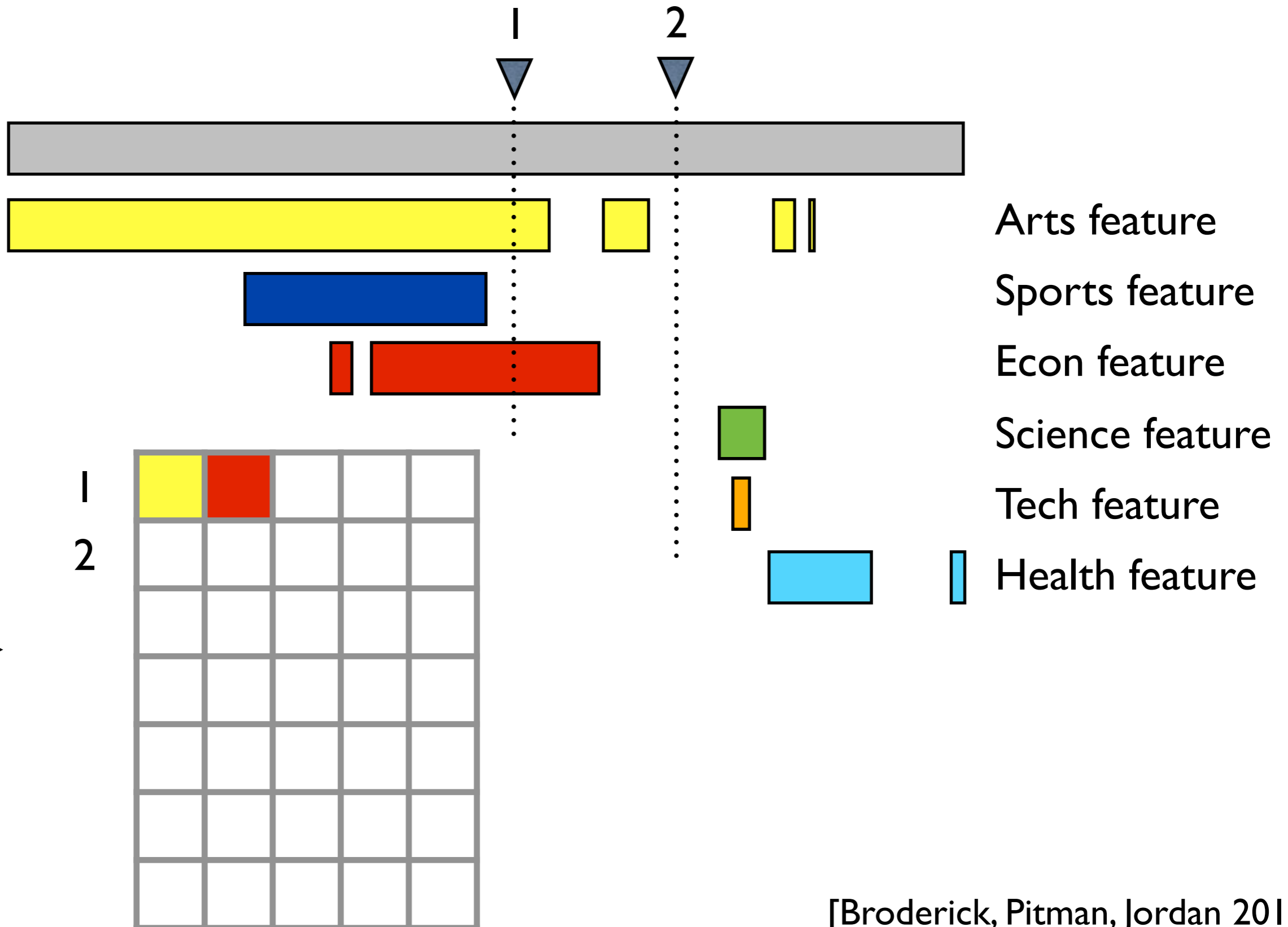
# Paintboxes

Exchangeable feature allocation: feature paintbox



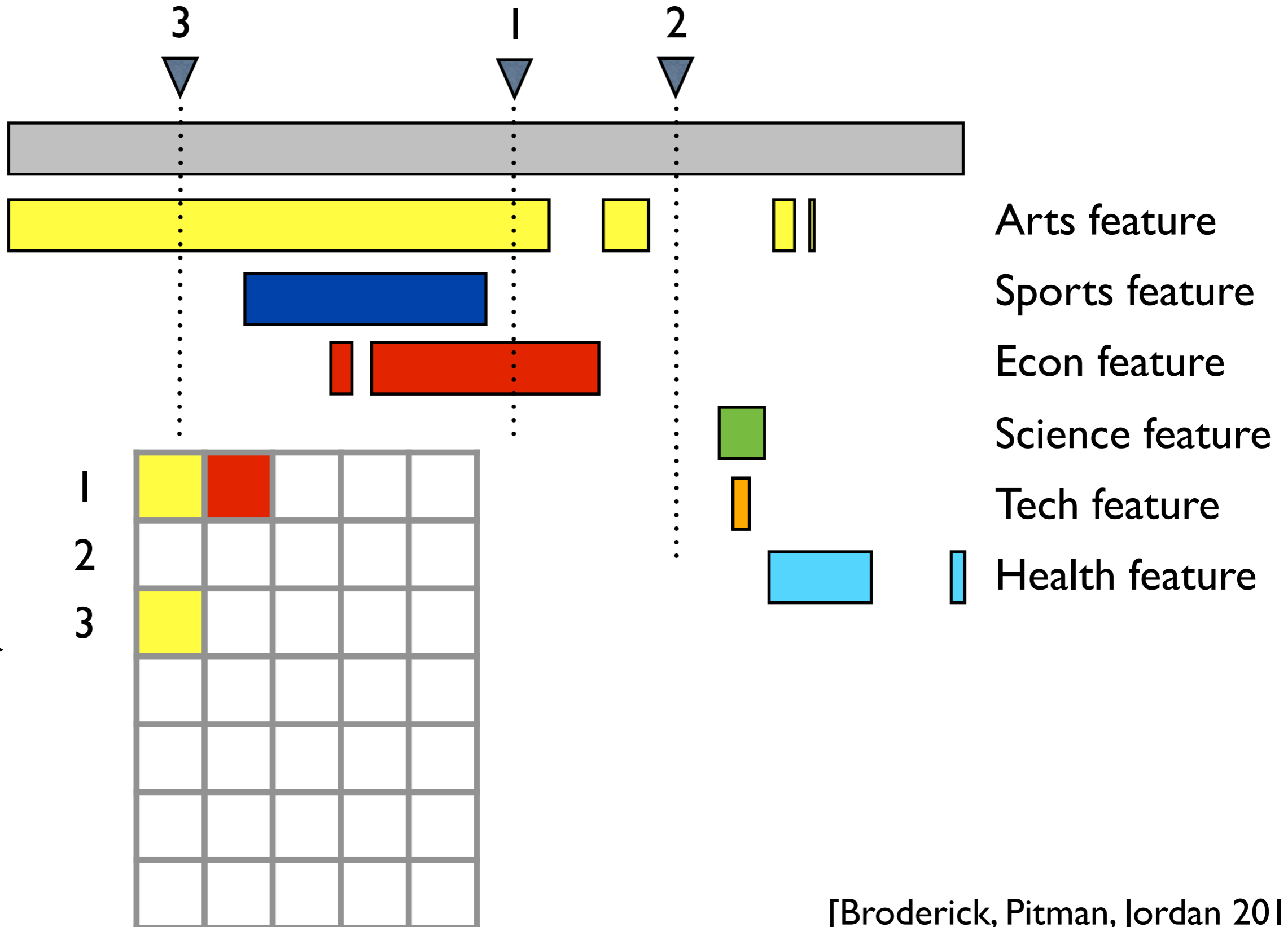
# Paintboxes

Exchangeable feature allocation: feature paintbox



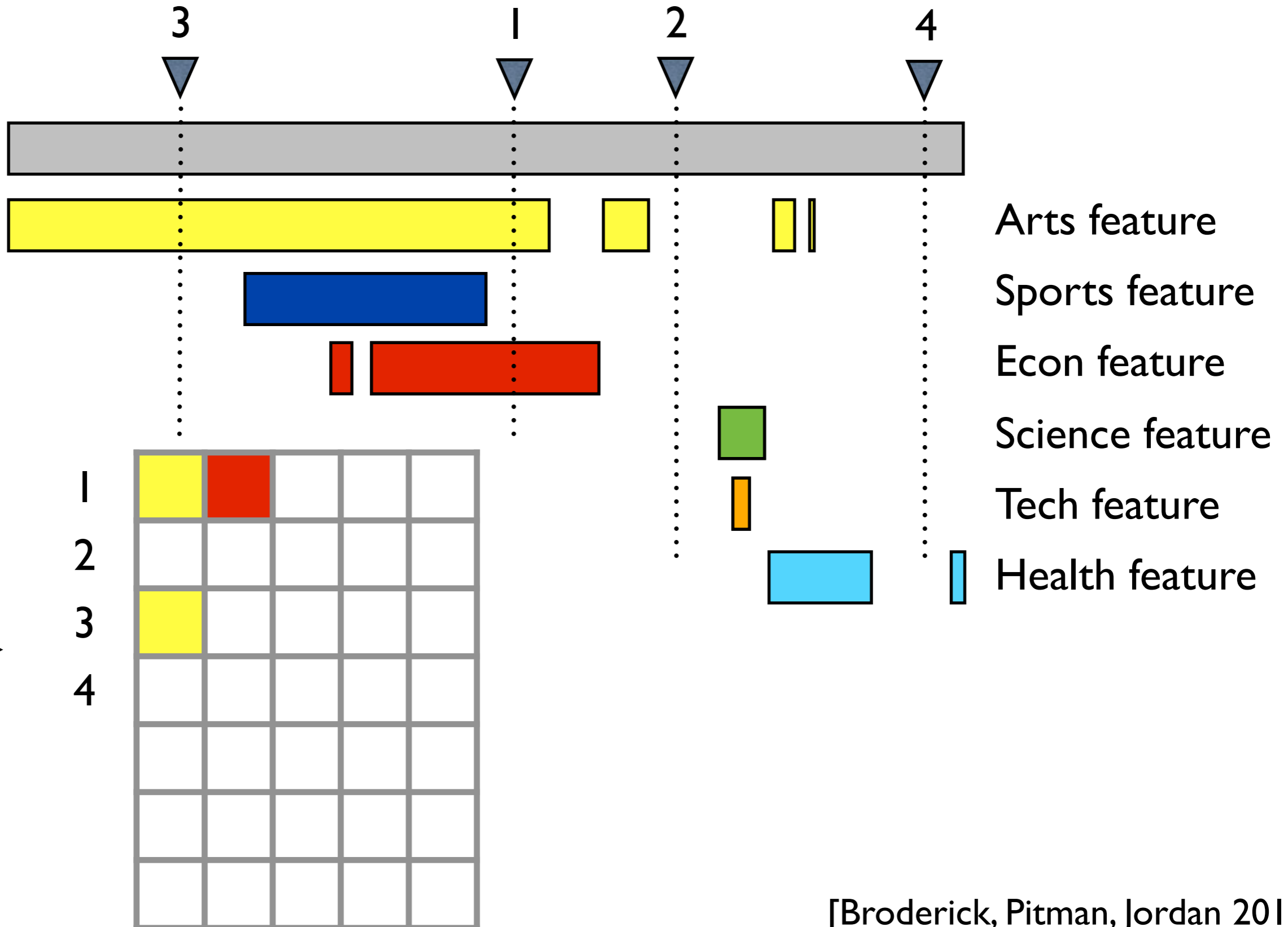
# Paintboxes

Exchangeable feature allocation: feature paintbox



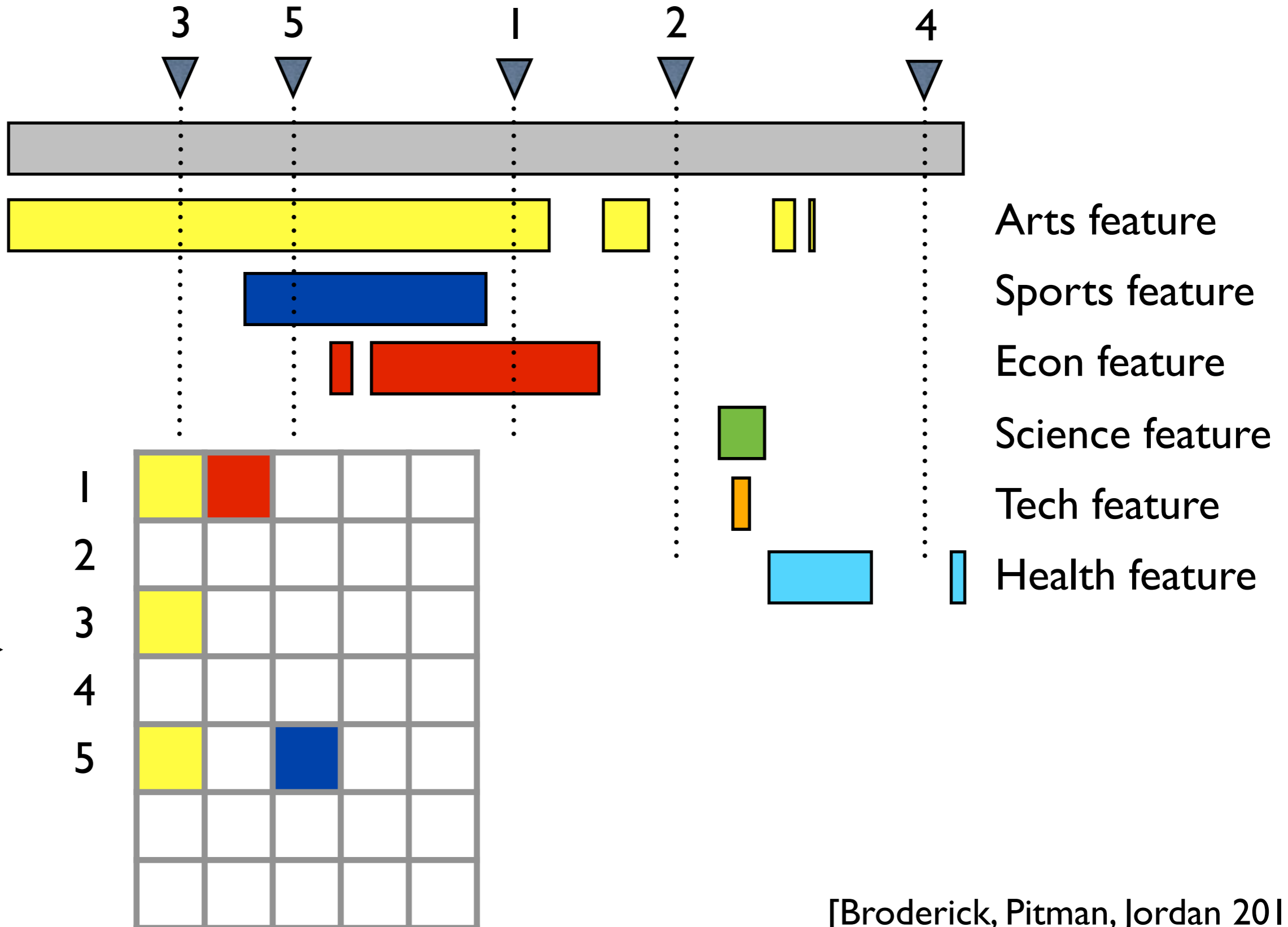
# Paintboxes

Exchangeable feature allocation: feature paintbox



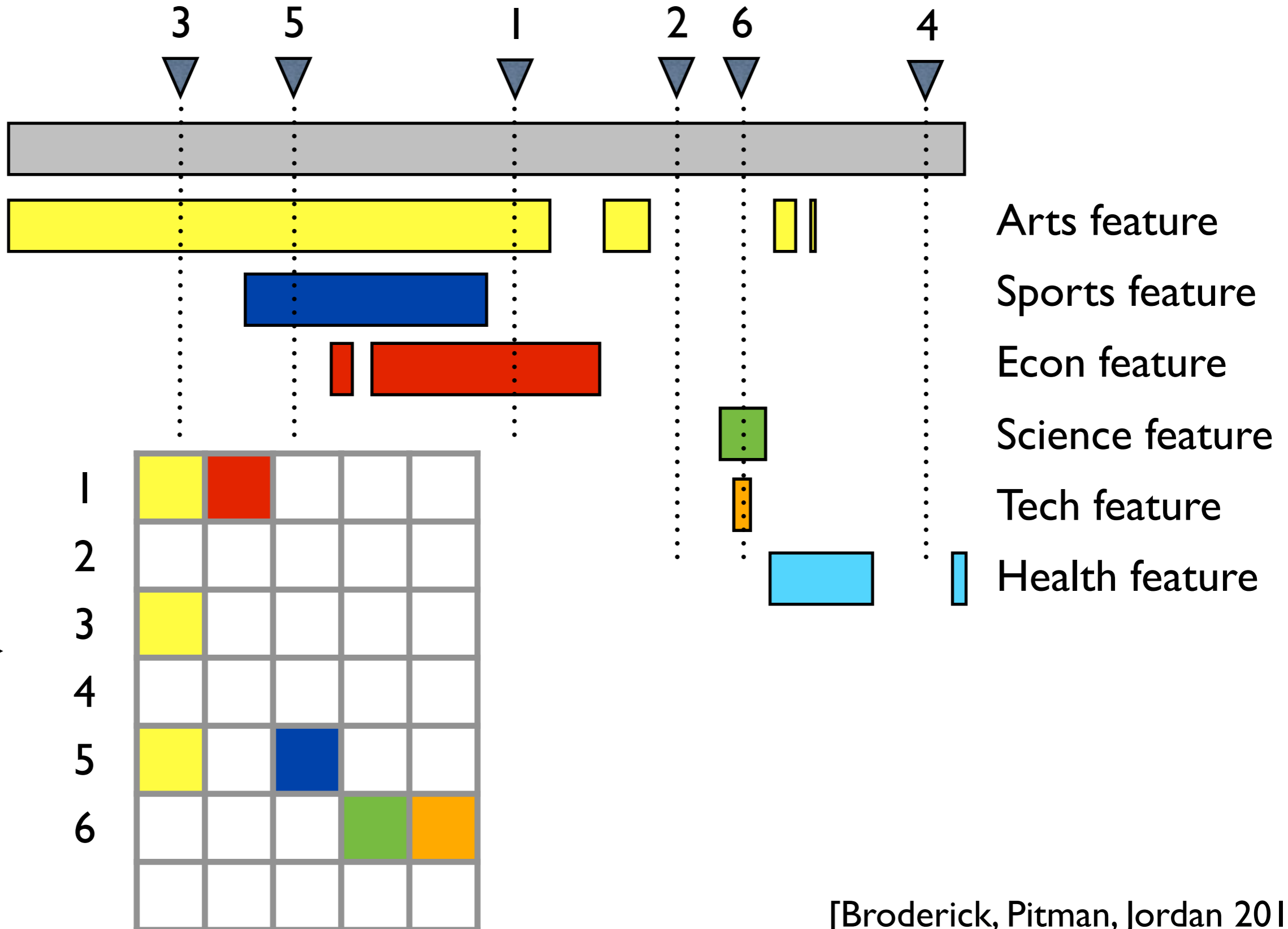
# Paintboxes

Exchangeable feature allocation: feature paintbox



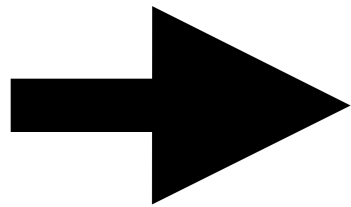
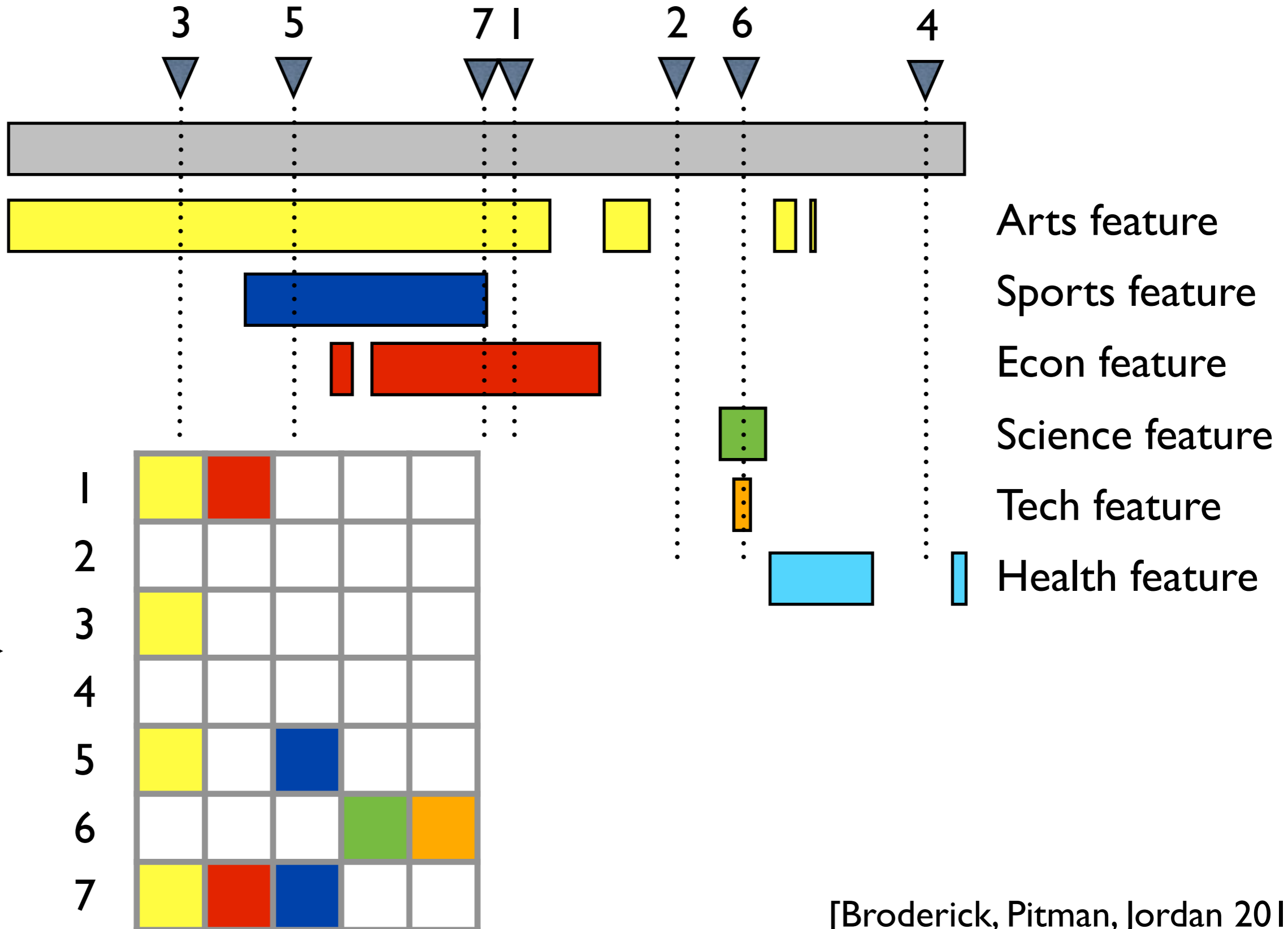
# Paintboxes

Exchangeable feature allocation: feature paintbox



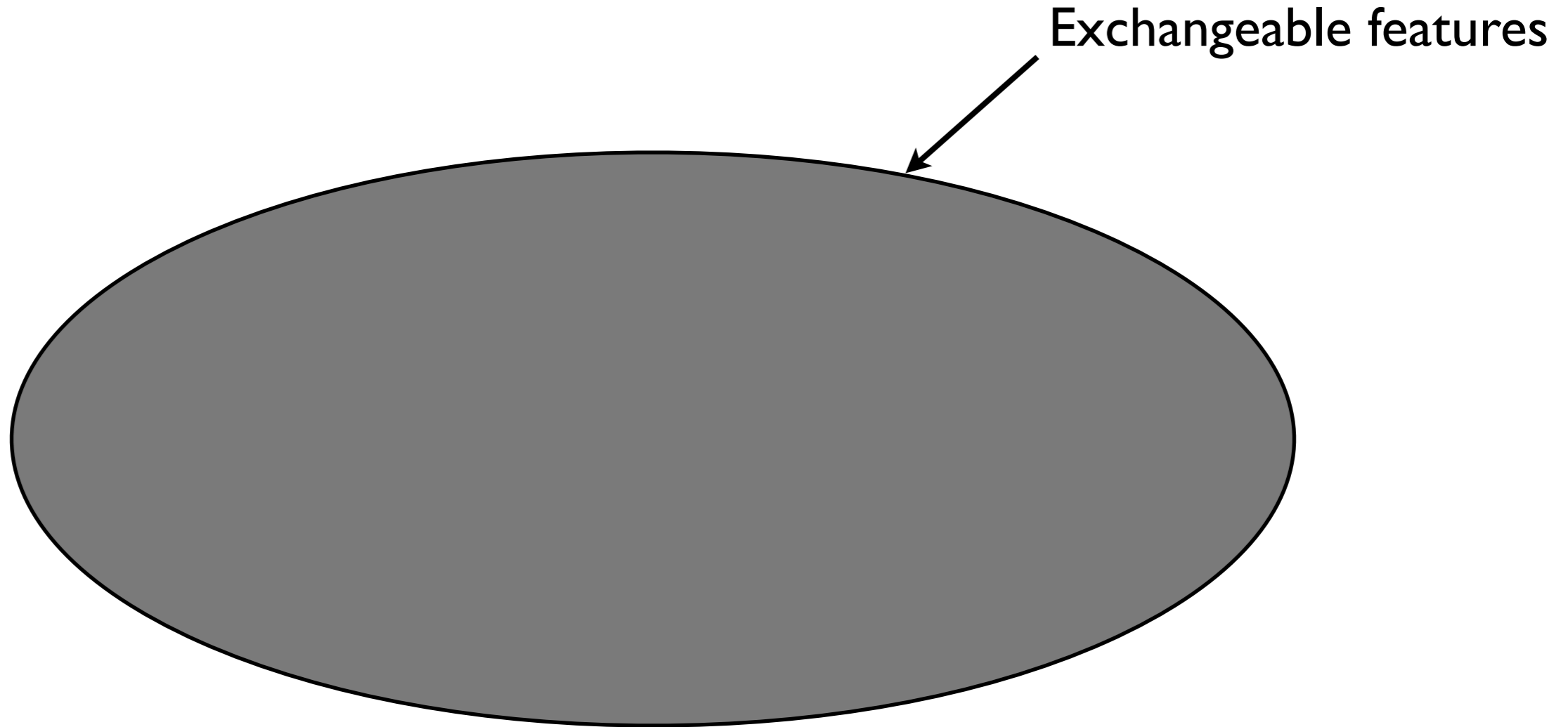
# Paintboxes

Exchangeable feature allocation: feature paintbox



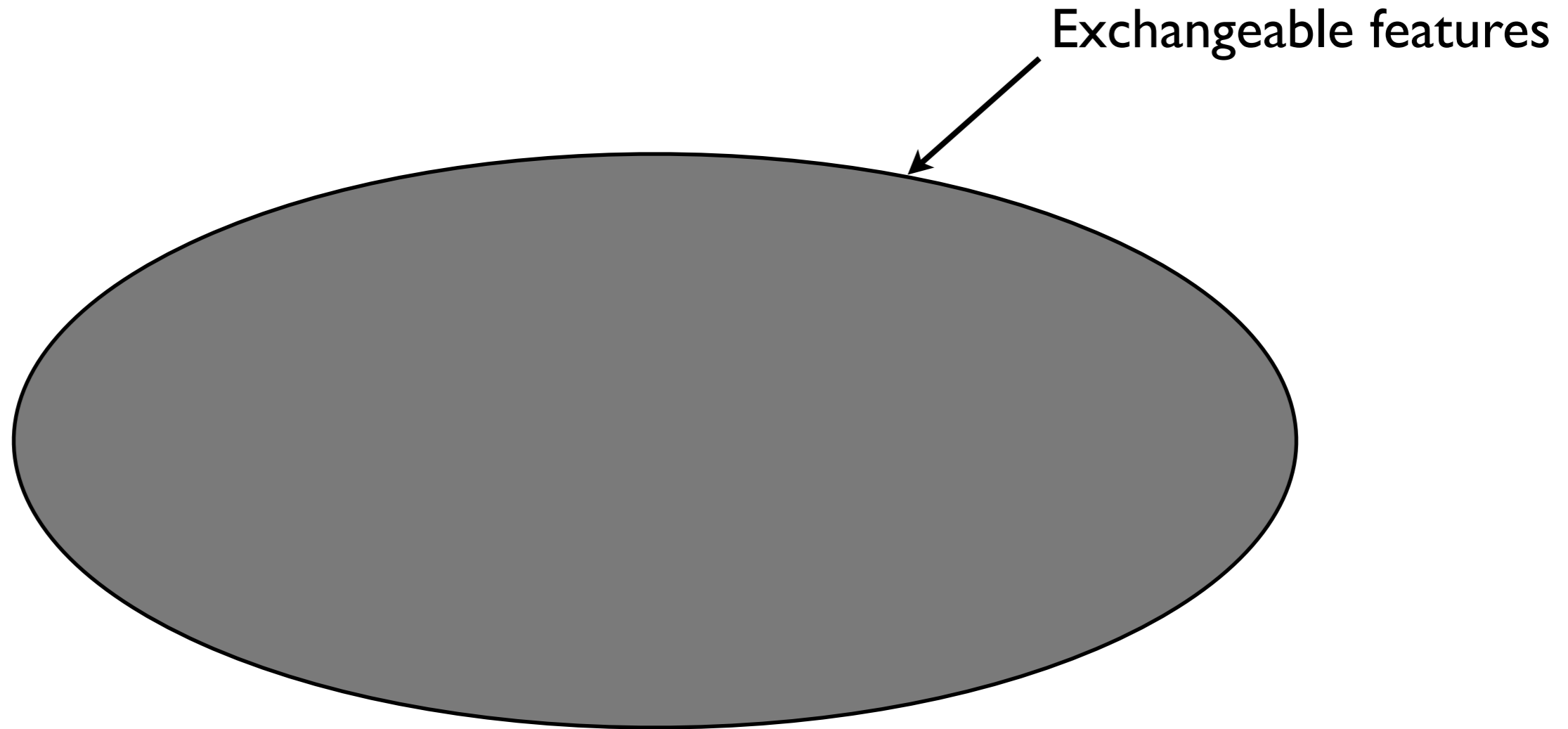


# Conclusions



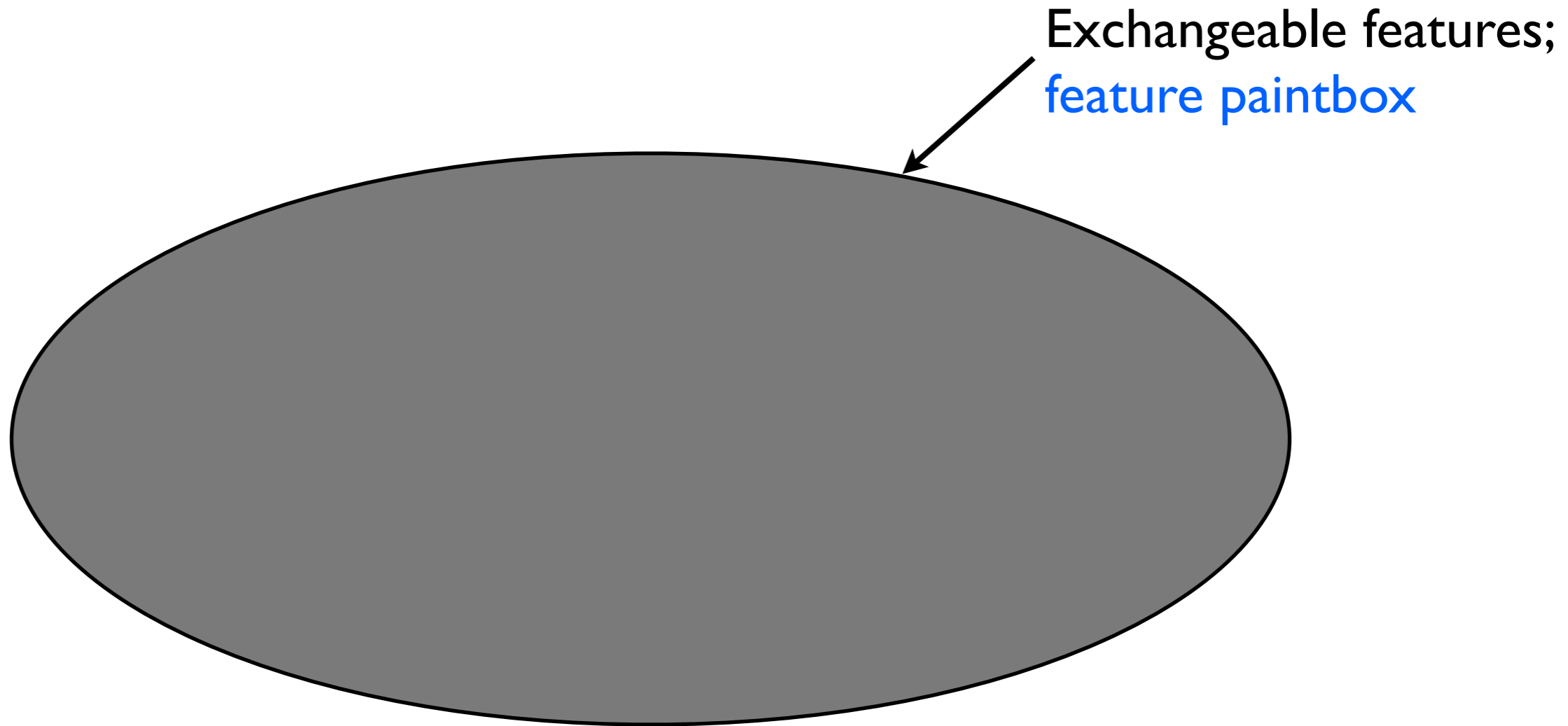
# Conclusions

- Feature paintbox: characterization of exchangeable feature models



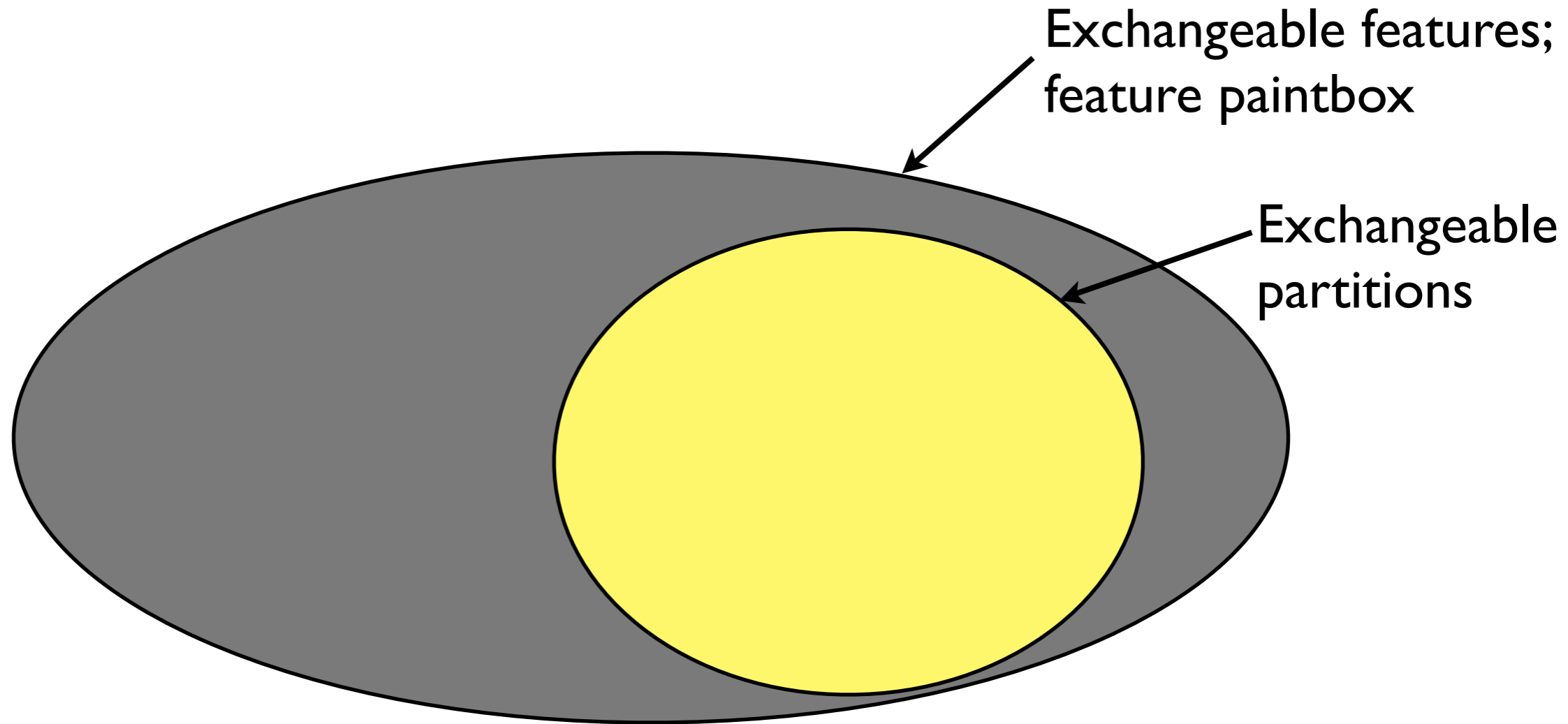
# Conclusions

- Feature paintbox: characterization of exchangeable feature models



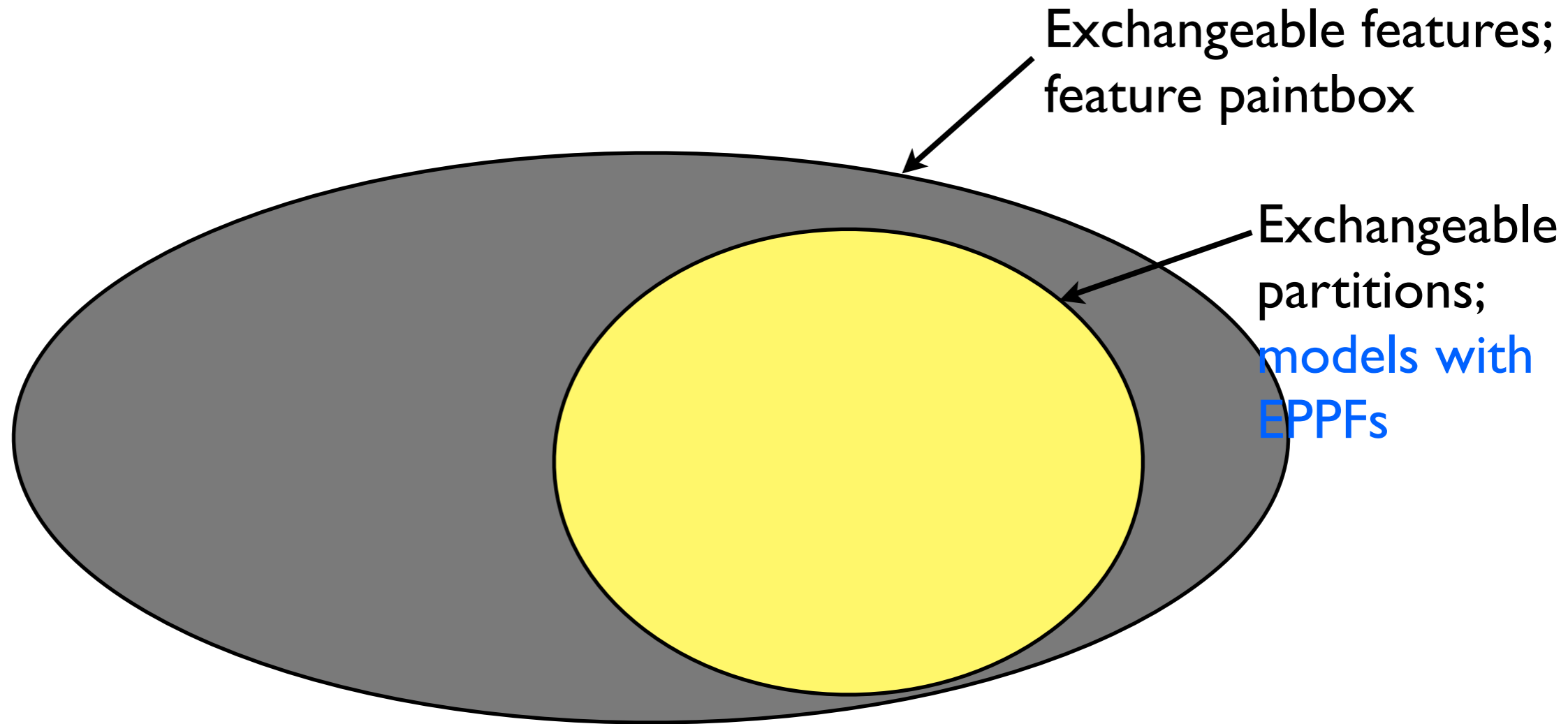
# Conclusions

- Feature paintbox: characterization of exchangeable feature models



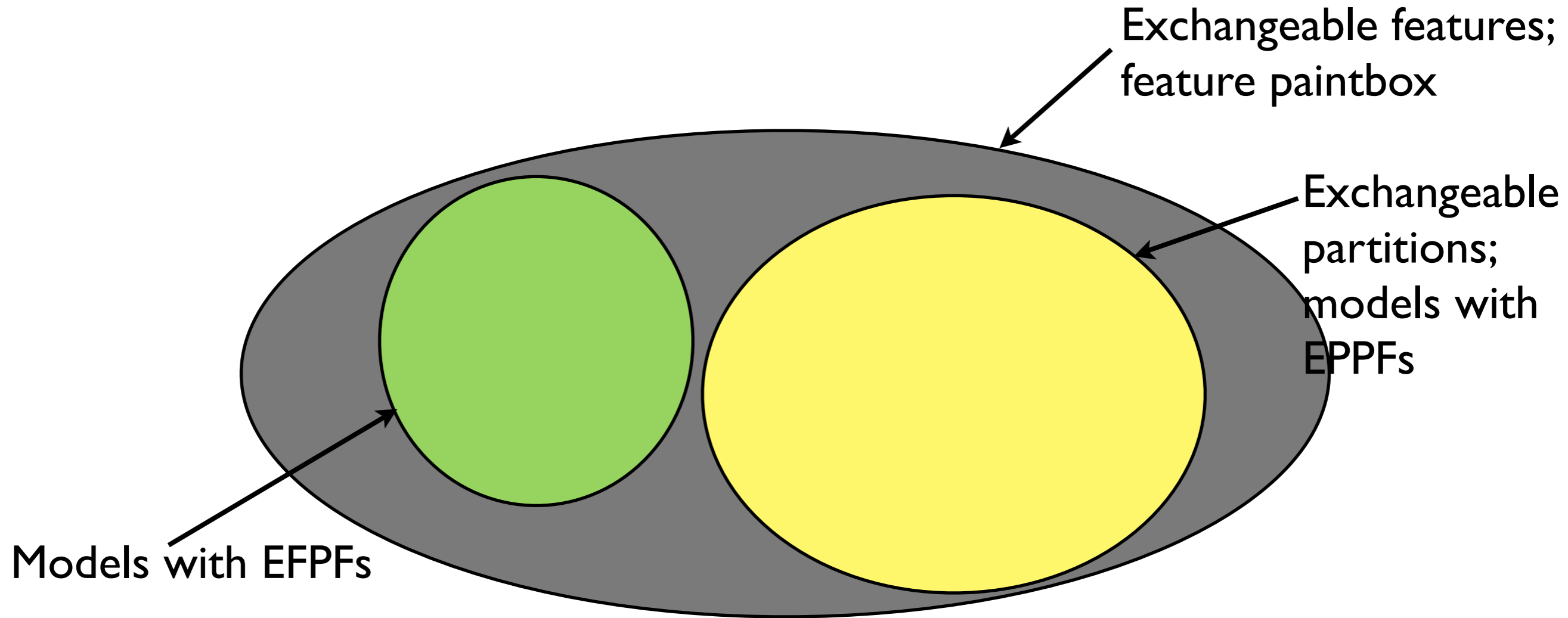
# Conclusions

- Feature paintbox: characterization of exchangeable feature models



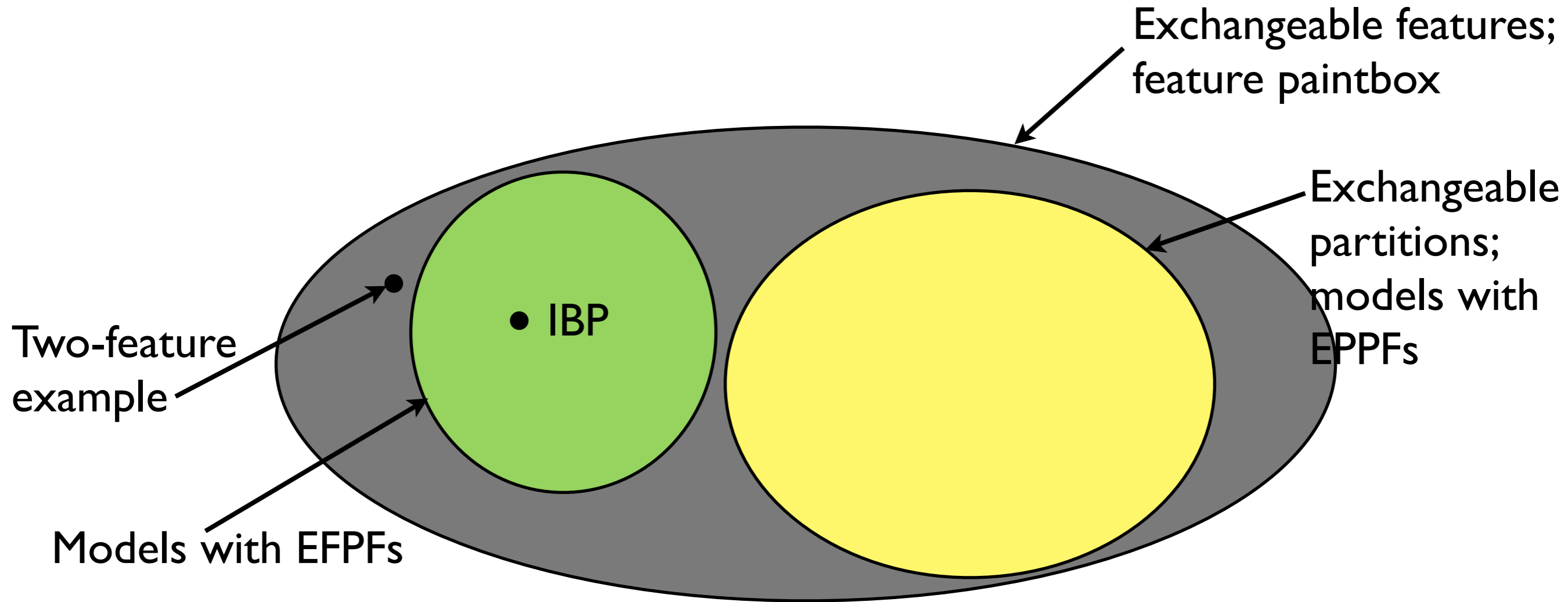
# Conclusions

- Feature paintbox: characterization of exchangeable feature models



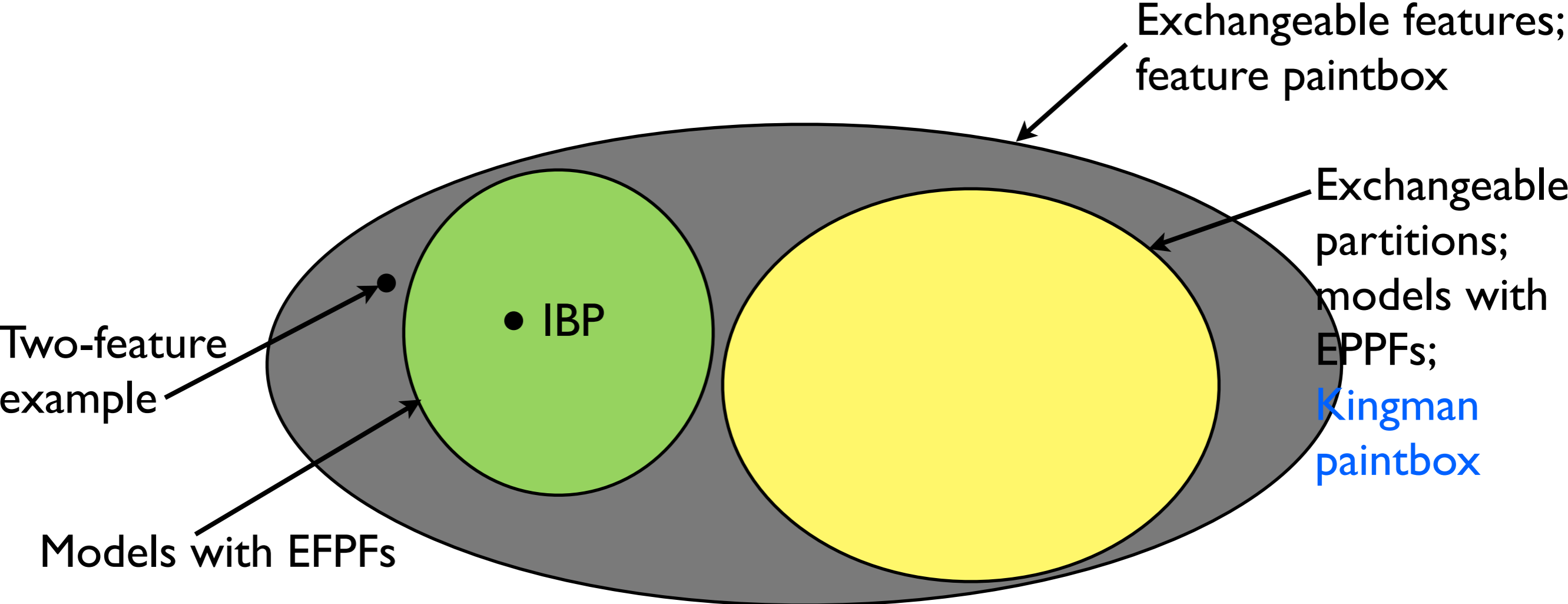
# Conclusions

- Feature paintbox: characterization of exchangeable feature models



# Conclusions

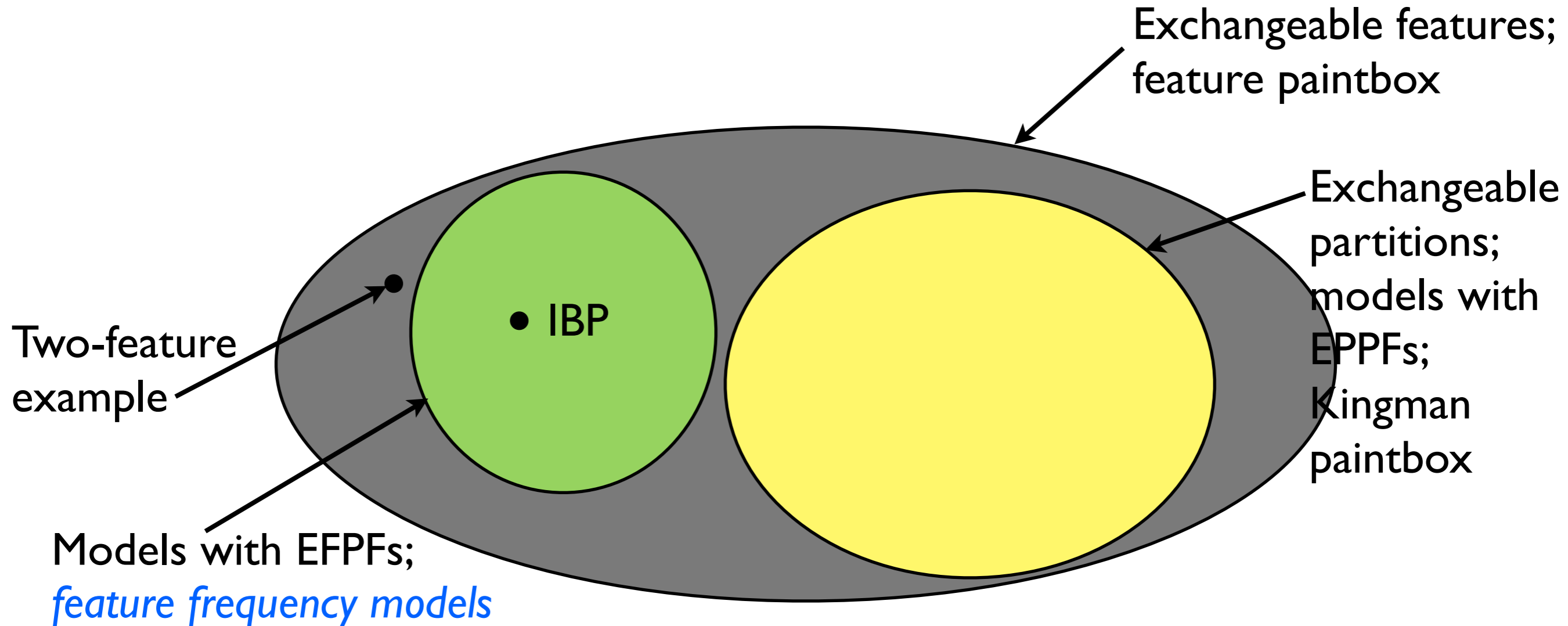
- Feature paintbox: characterization of exchangeable feature models





# Conclusions

- Feature paintbox: characterization of exchangeable feature models



# References

T. Broderick, J. Pitman, and M. I. Jordan. Feature allocations, probability functions, and paintboxes. *Bayesian Analysis*, 8(4):801-836, 2013.

T. Broderick, M. I. Jordan, and J. Pitman. Cluster and feature modeling from combinatorial stochastic processes. *Statistical Science*, 28(3):289-312, 2013.

# Further References

T. Griffiths and Z. Ghahramani. Infinite latent feature models and the Indian buffet process. In *Neural Information Processing Systems*, 2006.

N. L. Hjort. Nonparametric Bayes estimators based on beta processes in models for life history data. *Annals of Statistics*, 18(3):1259–1294, 1990.

J. F. C. Kingman. The representation of partition structures. *Journal of the London Mathematical Society*, 2(2):374, 1978.

J. Pitman. Exchangeable and partially exchangeable random partitions. *Probability Theory and Related Fields*, 102(2):145–158, 1995.

R. Thibaux and M. I. Jordan. Hierarchical beta processes and the Indian buffet process. In *International Conference on Artificial Intelligence and Statistics*, 2007.