



Uncertainty reasoning and machine learning

Uncertainty, Decision and Evaluation in Machine Learning

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AOS4 master courses

Outline

- Imprecise Dirichlet Model (IDM)
 - Frequentist and Bayesian Approaches
 - Imprecise Dirichlet Model
- Applications in classification tasks
- Evaluate Classifiers

Outline

- **Imprecise Dirichlet Model (IDM)**
 - Frequentist and Bayesian Approaches
 - Imprecise Dirichlet Model
- Applications in classification tasks
- Evaluate Classifiers

The Importance of Sample Size (Exercise 1)

Coin	Small	Large
Flips	2	$2 \cdot 10^6$
Heads	50%	50%
Tails	50%	50%

- For both coins, a frequentist says

$$p_{\text{Heads}} = p_{\text{Tails}} = 1/2$$

- Do Bayesians say the same thing?

The Importance of Sample Size (Exercise 1)

Coin	Small	Large
Flips	2	$2 \cdot 10^6$
Heads	50%	50%
Tails	50%	50%

- For both coins, a frequentist says

$$p_{\text{Heads}} = p_{\text{Tails}} = 1/2$$

- Do Bayesians say the same thing?

Coin	Small	Large
Flips	4	$4 \cdot 10^6$
Heads	25%	25%
Tails	75%	75%

- For both coins, a frequentist says

$$p_{\text{Heads}} = 0.25, p_{\text{Tails}} = 0.75$$

- Do Bayesians say the same thing?

The Importance of Sample Size (Exercise 1)

Coin	Small	Large
Flips	2	$2 \cdot 10^6$
Heads	50%	50%
Tails	50%	50%

- For both coins, a frequentist says

$$p_{\text{Heads}} = p_{\text{Tails}} = 1/2$$

- Do Bayesians say the same thing?

Coin	Small	Large
Flips	4	$4 \cdot 10^6$
Heads	25%	25%
Tails	75%	75%

- For both coins, a frequentist says

$$p_{\text{Heads}} = 0.25, p_{\text{Tails}} = 0.75$$

- Do Bayesians say the same thing?

Advocators	α_x	s	p_H^S	p_T^S	p_H^L	p_T^L
Haldane (1948)	0	0	???	???	???	???
Perks (1947)	$1/ \mathcal{V} $	1	???	???	???	???
Jeffreys (1946, 1961)	$1/2$	$ \mathcal{V} /2$???	???	???	???
Bayes-Laplace	1	$ \mathcal{V} $???	???	???	???

The Importance of Sample Size (Solution 1)

Coin	Small	Large
Flips	2	$2 \cdot 10^6$
Heads	50%	50%
Tails	50%	50%

- For both coins, a frequentist says

$$p_{\text{Heads}} = p_{\text{Tails}} = 1/2$$

- Do Bayesians say the same thing? ←
Yes!

Coin	Small	Large
Flips	4	$4 \cdot 10^6$
Heads	25%	25%
Tails	75%	75%

- For both coins, a frequentist says

$$p_{\text{Heads}} = 0.25, p_{\text{Tails}} = 0.75$$

- Do Bayesians say the same thing?

The Importance of Sample Size (Solution 1)

Coin	Small	Large
Flips	2	$2 \cdot 10^6$
Heads	50%	50%
Tails	50%	50%

- For both coins, a frequentist says

$$p_{\text{Heads}} = p_{\text{Tails}} = 1/2$$

- Do Bayesians say the same thing? ← **Yes!**

Coin	Small	Large
Flips	4	$4 \cdot 10^6$
Heads	25%	25%
Tails	75%	75%

- For both coins, a frequentist says

$$p_{\text{Heads}} = 0.25, p_{\text{Tails}} = 0.75$$

- Do Bayesians say the same thing?

Advocators	α_v	s	p_H^S	p_T^S	p_H^L	p_T^L
Haldane (1948)	0	0	0.25	0.75	0.25	0.75
Perks (1947)	$1/ V $	1	0.3	0.7	0.25	0.75
Jeffreys (1946, 1961)	$1/2$	$ V /2$	0.3	0.7	0.25	0.75
Bayes-Laplace	1	$ V $	0.33	0.67	0.25	0.75

The Importance of Sample Size (Exercise 2)

Coin	Small	Large
Flips	2	$2 \cdot 10^6$
Heads	0%	0%
Tails	100%	100%

- For both coins, a frequentist says

$$p_{\text{Heads}} = 0, p_{\text{Tails}} = 1$$

- Do Bayesians say the same thing?

The Importance of Sample Size (Exercise 2)

Coin	Small	Large
Flips	2	$2 \cdot 10^6$
Heads	0%	0%
Tails	100%	100%

- For both coins, a frequentist says

$$p_{\text{Heads}} = 0, p_{\text{Tails}} = 1$$

- Do Bayesians say the same thing?

Advocators	α_x	s	p_H^S	p_T^S	p_H^L	p_T^L
Haldane (1948)	0	0	???	???	???	???
Perks (1947)	$1/ \mathcal{V} $	1	???	???	???	???
Jeffreys (1946, 1961)	$1/2$	$ \mathcal{V} /2$???	???	???	???
Bayes-Laplace	1	$ \mathcal{V} $???	???	???	???

The Importance of Sample Size (Solution 2)

Coin	Small	Large
Flips	2	$2 \cdot 10^6$
Heads	0%	0%
Tails	100%	100%

- For both coins, a frequentist says
 $p_{\text{Heads}} = 0, p_{\text{Tails}} = 1$
- Do Bayesians say the same thing?

Advocators	α_x	s	p_H^S	p_T^S	p_H^L	p_T^L
Haldane (1948)	0	0	0	1	0	1
Perks (1947)	$1/ \mathcal{V} $	1	0.17	0.83	$3 \cdot 10^{-7}$	$1 - 3 \cdot 10^{-7}$
Jeffreys	$1/ \mathcal{V} $	1	0.17	0.83	$3 \cdot 10^{-7}$	$1 - 3 \cdot 10^{-7}$
Bayes-Laplace	1	$ \mathcal{V} $	0.25	0.75	$5 \cdot 10^{-7}$	$1 - 5 \cdot 10^{-7}$

Outline

- **Imprecise Dirichlet Model (IDM)**
 - Frequentist and Bayesian Approaches
 - Imprecise Dirichlet Model
- Applications in classification tasks
- Evaluate Classifiers

The Importance of Sample Size (Exercise 3)

Coin	Small	Large
Flips	2	$2 \cdot 10^6$
Heads	50%	50%
Tails	50%	50%

- For both coins, a frequentist says
 $\theta_{\text{Heads}} = \theta_{\text{Tails}} = 1/2$
- Bayesians would say the same thing
- Would IDM say the same thing?

The Importance of Sample Size (Exercise 3)

Coin	Small	Large
Flips	2	$2 \cdot 10^6$
Heads	50%	50%
Tails	50%	50%

- For both coins, a frequentist says
 $\theta_{\text{Heads}} = \theta_{\text{Tails}} = 1/2$
- Bayesians would say the same thing
- Would IDM say the same thing?

	\underline{P}_H^S	\overline{P}_H^S	\underline{P}_H^L	\overline{P}_H^L
$s = 1$???	???	???	???
$s = 2$???	???	???	???

The Importance of Sample Size (Solution 3)

Coin	Small	Large	<ul style="list-style-type: none"> For both coins, a frequentist says $\theta_{\text{Heads}} = \theta_{\text{Tails}} = 1/2$ Bayesians would say the same thing Would IDM say the same thing?
Flips	2	$2 \cdot 10^6$	
Heads	50%	50%	
Tails	50%	50%	

	\underline{P}_H^S	\overline{P}_H^S	\underline{P}_H^L	\overline{P}_H^L
$s = 1$	0.33	0.67	$0.5 - 3 \cdot 10^{-7}$	$0.5 + 3 \cdot 10^{-7}$
$s = 2$	0.25	0.75	$0.5 - 5 \cdot 10^{-7}$	$0.5 + 5 \cdot 10^{-7}$

The Importance of Sample Size (Exercise 4)

Coin	Small	Large
Flips	2	$2 \cdot 10^6$
Heads	0%	0%
Tails	100%	100%

- For both coins, a frequentist says
 $\theta_{\text{Heads}} = 0, \theta_{\text{Tails}} = 1$
- Bayesians would say different things
- What would IDM say?

The Importance of Sample Size (Exercise 4)

Coin	Small	Large
Flips	2	$2 \cdot 10^6$
Heads	0%	0%
Tails	100%	100%

- For both coins, a frequentist says

$$\theta_{\text{Heads}} = 0, \theta_{\text{Tails}} = 1$$

- Bayesians would say different things
- What would IDM say?

Advocators	α_x	s	p_H^S	p_T^S	p_H^L	p_T^L
Haldane (1948)	0	0	0	1	0	1
Perks (1947)	$1/ \mathcal{V} $	1	0.17	0.83	$3 \cdot 10^{-7}$	$1 - 3 \cdot 10^{-7}$
Jeffreys	$1/ \mathcal{V} $	1	0.17	0.83	$3 \cdot 10^{-7}$	$1 - 3 \cdot 10^{-7}$
Bayes-Laplace	1	$ \mathcal{V} $	0.25	0.75	$5 \cdot 10^{-7}$	$1 - 5 \cdot 10^{-7}$

The Importance of Sample Size (Exercise 4)

Coin	Small	Large
Flips	2	$2 \cdot 10^6$
Heads	0%	0%
Tails	100%	100%

- For both coins, a frequentist says

$$\theta_{\text{Heads}} = 0, \theta_{\text{Tails}} = 1$$

- Bayesians would say different things
- What would IDM say?

Advocators	α_x	s	p_H^S	p_T^S	p_H^L	p_T^L
Haldane (1948)	0	0	0	1	0	1
Perks (1947)	$1/ \mathcal{V} $	1	0.17	0.83	$3 \cdot 10^{-7}$	$1 - 3 \cdot 10^{-7}$
Jeffreys	$1/ \mathcal{V} $	1	0.17	0.83	$3 \cdot 10^{-7}$	$1 - 3 \cdot 10^{-7}$
Bayes-Laplace	1	$ \mathcal{V} $	0.25	0.75	$5 \cdot 10^{-7}$	$1 - 5 \cdot 10^{-7}$

IDM	\underline{p}_H^S	\overline{p}_H^S	\underline{p}_H^L	\overline{p}_H^L
$s = 1$???	???	???	???
$s = 2$???	???	???	???

The Importance of Sample Size (Solution 4)

Coin	Small	Large	• For both coins, a frequentist says
Flips	2	$2 \cdot 10^6$	$\theta_{\text{Heads}} = 0, \theta_{\text{Tails}} = 1$
Heads	0%	0%	• Bayesians would say different things
Tails	100%	100%	• What would IDM say?

Advocators	α_X	s	p_H^S	p_T^S	p_H^L	p_T^L
Haldane (1948)	0	0	0	1	0	1
Perks (1947)	$1/ \mathcal{V} $	1	0.17	0.83	$3 \cdot 10^{-7}$	$1 - 3 \cdot 10^{-7}$
Jeffreys	$1/ \mathcal{V} $	1	0.17	0.83	$3 \cdot 10^{-7}$	$1 - 3 \cdot 10^{-7}$
Bayes-Laplace	1	$ \mathcal{V} $	0.25	0.75	$5 \cdot 10^{-7}$	$1 - 5 \cdot 10^{-7}$

IDM	\underline{p}_H^S	\overline{p}_H^S	\underline{p}_H^L	\overline{p}_H^L
$s = 1$	0	0.33	0	$5 \cdot 10^{-7}$
$s = 2$	0	0.50	0	10^{-6}

Determine \mathcal{D} (Exercise 5)

Coin	Small	Large
Flips	2	$2 \cdot 10^6$
Heads	$[0, 1]$	$[5, 10]$
Tails	$[1, 2]$	$[5, 2 \cdot 10^6]$

- Recap: $\mathcal{D} = \{\mathbf{D} | n_v \in \mathbf{n}_v, \sum_{v \in \mathcal{V}} n_v = n\}$
- What is \mathcal{D}^S for the first coin?
- What is \mathcal{D}^L for the second coin?

Determine \mathcal{D} (Exercise 5)

Coin	Small	Large
Flips	2	$2 \cdot 10^6$
Heads	$[0, 1]$	$[5, 10]$
Tails	$[1, 2]$	$[5, 2 \cdot 10^6]$

- Recap: $\mathcal{D} = \{\mathbf{D} | n_v \in \mathbf{n}_v, \sum_{v \in \mathcal{V}} n_v = n\}$
- What is \mathcal{D}^S for the first coin?
- What is \mathcal{D}^L for the second coin?

Coin	Small	\mathbf{D}_1	\mathbf{D}_2
Flips	2	2	2
Heads	$[0, 1]$	0	1
Tails	$[1, 2]$	2	1

Determine \mathcal{D} (Exercise 5)

Coin	Small	Large	<ul style="list-style-type: none"> Recap: $\mathcal{D} = \{\mathbf{D} n_V \in \mathbf{n}_V, \sum_{V \in \mathcal{V}} n_V = n\}$ What is \mathcal{D}^S for the first coin? What is \mathcal{D}^L for the second coin?
Flips	2	$2 \cdot 10^6$	
Heads	$[0, 1]$	$[5, 10]$	
Tails	$[1, 2]$	$[5, 2 \cdot 10^6]$	

Coin	Small	\mathbf{D}_1	\mathbf{D}_2
Flips	2	2	2
Heads	$[0, 1]$	0	1
Tails	$[1, 2]$	2	1

Coin	Large	\mathbf{D}_1	\mathbf{D}_2	\mathbf{D}_3	\mathbf{D}_4	\mathbf{D}_5	\mathbf{D}_6
Flips	$n = 2 \cdot 10^6$	n	n	n	n	n	n
Heads	$[5, 10]$???	???	???	???	???	???
Tails	$[5, n]$???	???	???	???	???	???

Determine \mathcal{D} (Solution 5)

Coin	Small	Large
Flips	2	$2 \cdot 10^6$
Heads	$[0, 1]$	$[5, 10]$
Tails	$[1, 2]$	$[5, 2 \cdot 10^6]$

- Recap: $\mathcal{D} = \{\mathbf{D} | n_v \in \mathbf{n}_v, \sum_{v \in \mathcal{V}} n_v = n\}$
- What is \mathcal{D}^S for the first coin?
- What is \mathcal{D}^L for the second coin?

Coin	Small	\mathbf{D}_1	\mathbf{D}_2
Flips	2	2	2
Heads	$[0, 1]$	0	1
Tails	$[1, 2]$	2	1

Coin	Large	\mathbf{D}_1	\mathbf{D}_2	\mathbf{D}_3	\mathbf{D}_4	\mathbf{D}_5	\mathbf{D}_6
Flips	$n = 2 \cdot 10^6$	n	n	n	n	n	n
Heads	$[5, 10]$	5	6	7	8	9	10
Tails	$[5, n]$	$n - 5$	$n - 6$	$n - 7$	$n - 8$	$n - 9$	$n - 10$

Compute Lower and Upper Expectations (Exercise 6)

Interval posterior mean $\theta_v^*|\mathcal{D}$ of $\theta_v|\mathcal{D}$:

$$\underline{E}(\theta_v|\mathcal{D}) = \min_{\mathbf{D} \in \mathcal{D}} \underline{E}(\theta_v|\mathbf{D}) = \min_{\mathbf{D} \in \mathcal{D}} n_v/(n+s), \quad (1)$$

$$\overline{E}(\theta_v|\mathcal{D}) = \max_{\mathbf{D} \in \mathcal{D}} \overline{E}(\theta_v|\mathbf{D}) = \max_{\mathbf{D} \in \mathcal{D}} (n_v+s)/(n+s). \quad (2)$$

Compute Lower and Upper Expectations (Exercise 6)

Interval posterior mean $\theta_v^*|\mathcal{D}$ of $\theta_v|\mathcal{D}$:

$$\underline{E}(\theta_v|\mathcal{D}) = \min_{\mathbf{D} \in \mathcal{D}} \underline{E}(\theta_v|\mathbf{D}) = \min_{\mathbf{D} \in \mathcal{D}} n_v/(n+s), \quad (1)$$

$$\overline{E}(\theta_v|\mathcal{D}) = \max_{\mathbf{D} \in \mathcal{D}} \overline{E}(\theta_v|\mathbf{D}) = \max_{\mathbf{D} \in \mathcal{D}} (n_v+s)/(n+s). \quad (2)$$

Coin	Small	\mathbf{D}_1	\mathbf{D}_2	$\underline{E}(\theta_v \mathcal{D})$	$\overline{E}(\theta_v \mathcal{D})$
Flips	2	2	2		
Heads	[0, 1]	0	1	???	???
Tails	[1, 2]	2	1	???	???

Compute Lower and Upper Expectations (Exercise 6)

Interval posterior mean $\theta_v^*|\mathcal{D}$ of $\theta_v|\mathcal{D}$:

$$\underline{E}(\theta_v|\mathcal{D}) = \min_{\mathbf{D} \in \mathcal{D}} \underline{E}(\theta_v|\mathbf{D}) = \min_{\mathbf{D} \in \mathcal{D}} n_v/(n+s), \quad (1)$$

$$\overline{E}(\theta_v|\mathcal{D}) = \max_{\mathbf{D} \in \mathcal{D}} \overline{E}(\theta_v|\mathbf{D}) = \max_{\mathbf{D} \in \mathcal{D}} (n_v+s)/(n+s). \quad (2)$$

Coin	Small	\mathbf{D}_1	\mathbf{D}_2	$\underline{E}(\theta_v \mathcal{D})$	$\overline{E}(\theta_v \mathcal{D})$
Flips	2	2	2		
Heads	[0, 1]	0	1	???	???
Tails	[1, 2]	2	1	???	???

Coin	Large	\mathbf{D}_1	...	\mathbf{D}_6	$\underline{E}(\theta_v \mathcal{D})$	$\overline{E}(\theta_v \mathcal{D})$
Flips	$n = 2 \cdot 10^6$	n	...	n		
Heads	[5, 10]	5	...	10	???	???
Tails	[5, n]	$n-5$...	$n-10$???	???

Compute Lower and Upper Expectations (Solution 6)

Interval posterior mean $\theta_v^*|\mathcal{D}$ of $\theta_v|\mathcal{D}$:

$$\underline{E}(\theta_v|\mathcal{D}) = \min_{\mathbf{D} \in \mathcal{D}} \underline{E}(\theta_v|\mathbf{D}) = \min_{\mathbf{D} \in \mathcal{D}} n_v/(n+s), \quad (3)$$

$$\overline{E}(\theta_v|\mathcal{D}) = \max_{\mathbf{D} \in \mathcal{D}} \overline{E}(\theta_v|\mathbf{D}) = \max_{\mathbf{D} \in \mathcal{D}} (n_v+s)/(n+s). \quad (4)$$

Coin	Small	\mathbf{D}_1	\mathbf{D}_2	$\underline{E}(\theta_v \mathcal{D})$	$\overline{E}(\theta_v \mathcal{D})$
Flips	2	2	2		
Heads	[0, 1]	0	1	0/(2+s)	(1+s)/(2+s)
Tails	[1, 2]	2	1	1/(2+s)	(2+s)/(2+s)

Compute Lower and Upper Expectations (Solution 6)

Interval posterior mean $\theta_v^*|\mathcal{D}$ of $\theta_v|\mathcal{D}$:

$$\underline{E}(\theta_v|\mathcal{D}) = \min_{\mathbf{D} \in \mathcal{D}} \underline{E}(\theta_v|\mathbf{D}) = \min_{\mathbf{D} \in \mathcal{D}} n_v/(n+s), \quad (3)$$

$$\overline{E}(\theta_v|\mathcal{D}) = \max_{\mathbf{D} \in \mathcal{D}} \overline{E}(\theta_v|\mathbf{D}) = \max_{\mathbf{D} \in \mathcal{D}} (n_v+s)/(n+s). \quad (4)$$

Coin	Small	\mathbf{D}_1	\mathbf{D}_2	$\underline{E}(\theta_v \mathcal{D})$	$\overline{E}(\theta_v \mathbf{D})$
Flips	2	2	2		
Heads	[0, 1]	0	1	$0/(2+s)$	$(1+s)/(2+s)$
Tails	[1, 2]	2	1	$1/(2+s)$	$(2+s)/(2+s)$

Coin	Large	\mathbf{D}_1	...	\mathbf{D}_6	$\underline{E}(\theta_v \mathcal{D})$	$\overline{E}(\theta_v \mathbf{D})$
Flips	$n = 2 \cdot 10^6$	n	...	n		
Heads	[5, 10]	5	...	10	$5/(n+s)$	$(10+s)/(n+s)$
Tails	[5, n]	$n-5$...	$n-10$	$(n-10)/(n+s)$	$(n-5+s)/(n+s)$

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Determine Possible Precise Data Set (Exercise 7)

$\mathbf{x}' \in \mathbf{D}_\epsilon(\mathbf{x})$	$Y \subset \mathcal{Y} = \{\text{Apple, Banana, Tomato}\}$
\mathbf{x}'_1	Apple or Banana, but not Tomato
\mathbf{x}'_2	Banana or Tomato, but not Apple
\mathbf{x}'_3	Apple or Tomato, but not Banana
\mathbf{x}'_4	Tomato
\mathbf{x}'_5	Tomato
\mathbf{x}'_6	Banana
\mathbf{x}'_7	Banana

7

$$n = 7, \mathbf{n}_A = ???, \mathbf{n}_B = ???, \mathbf{n}_T = ??? \quad (5)$$

Determine Possible Precise Data Set (Exercise 7)

$\mathbf{x}' \in \mathbf{D}_\epsilon(\mathbf{x})$	$Y \subset \mathcal{Y} = \{\text{Apple, Banana, Tomato}\}$
\mathbf{x}'_1	Apple or Banana, but not Tomato
\mathbf{x}'_2	Banana or Tomato, but not Apple
\mathbf{x}'_3	Apple or Tomato, but not Banana
\mathbf{x}'_4	Tomato
\mathbf{x}'_5	Tomato
\mathbf{x}'_6	Banana
\mathbf{x}'_7	Banana

7

$n = 7, n_A = ???, n_B = ???, n_T = ???$ (5)

	\mathbf{D}_1	\mathbf{D}_2	\mathbf{D}_3	\mathbf{D}_4	\mathbf{D}_5	\mathbf{D}_6	\mathbf{D}_7	\mathbf{D}_8
n_A	???	???	???	???	???	???	???	???
n_B	???	???	???	???	???	???	???	???
n_T	???	???	???	???	???	???	???	???

Determine Possible Precise Data Set (Solution 7)

$\mathbf{x}' \in \mathbf{D}_\epsilon(\mathbf{x})$	$Y \subset \mathcal{Y} = \{\text{Apple, Banana, Tomato}\}$
\mathbf{x}'_1	Apple or Banana, but not Tomato
\mathbf{x}'_2	Banana or Tomato, but not Apple
\mathbf{x}'_3	Apple or Tomato, but not Banana
\mathbf{x}'_4	Tomato
\mathbf{x}'_5	Tomato
\mathbf{x}'_6	Banana
\mathbf{x}'_7	Banana

$$n = 7, \mathbf{n}_A = \{0, 1, 2\}, \mathbf{n}_B = \{2, 3, 4\}, \mathbf{n}_T = \{2, 3, 4\} \quad (6)$$

Determine Possible Precise Data Set (Solution 7)

$\mathbf{x}' \in \mathbf{D}_\epsilon(\mathbf{x})$	$Y \subset \mathcal{Y} = \{\text{Apple, Banana, Tomato}\}$
\mathbf{x}'_1	Apple or Banana, but not Tomato
\mathbf{x}'_2	Banana or Tomato, but not Apple
\mathbf{x}'_3	Apple or Tomato, but not Banana
\mathbf{x}'_4	Tomato
\mathbf{x}'_5	Tomato
\mathbf{x}'_6	Banana
\mathbf{x}'_7	Banana

$$n = 7, \mathbf{n}_A = \{0, 1, 2\}, \mathbf{n}_B = \{2, 3, 4\}, \mathbf{n}_T = \{2, 3, 4\} \quad (6)$$

	\mathbf{D}_1	\mathbf{D}_2	\mathbf{D}_3	\mathbf{D}_4	\mathbf{D}_5	\mathbf{D}_6	\mathbf{D}_7	\mathbf{D}_8
n_A	0	0	1	1	1	2	2	2
n_B	3	4	2	3	4	2	3	4
n_T	4	3	4	3	2	4	3	3

Compute Lower and Upper Expectations (Exercise 8)

	D_1	D_2	D_3	D_4	D_5	D_6	D_7	D_8
n_A	0	0	1	1	1	2	2	2
n_B	3	4	2	3	4	2	3	4
n_T	4	3	4	3	2	4	3	3

Using IDM to estimate **interval posterior mean** $\theta_y^*|\mathcal{D}$ of $\theta_y|\mathcal{D}$:

$$\underline{E}(\theta_y|\mathbf{x}) = \min_{\mathbf{D} \in \mathcal{D}} \underline{E}(\theta_y|\mathbf{D}) = \min_{\mathbf{D} \in \mathcal{D}} n_y/(n+s), \quad (7)$$

$$\overline{E}(\theta_y|\mathbf{x}) = \max_{\mathbf{D} \in \mathcal{D}} \overline{E}(\theta_y|\mathbf{D}) = \max_{\mathbf{D} \in \mathcal{D}} (n_y+s)/(n+s). \quad (8)$$

Compute Lower and Upper Expectations (Exercise 8)

	D_1	D_2	D_3	D_4	D_5	D_6	D_7	D_8
n_A	0	0	1	1	1	2	2	2
n_B	3	4	2	3	4	2	3	4
n_T	4	3	4	3	2	4	3	3

Using IDM to estimate **interval posterior mean** $\theta_y^*|\mathcal{D}$ of $\theta_y|\mathcal{D}$:

$$\underline{E}(\theta_y|\mathbf{x}) = \min_{\mathbf{D} \in \mathcal{D}} \underline{E}(\theta_y|\mathbf{D}) = \min_{\mathbf{D} \in \mathcal{D}} n_y/(n+s), \quad (7)$$

$$\overline{E}(\theta_y|\mathbf{x}) = \max_{\mathbf{D} \in \mathcal{D}} \overline{E}(\theta_y|\mathbf{D}) = \max_{\mathbf{D} \in \mathcal{D}} (n_y+s)/(n+s). \quad (8)$$

	$\underline{E}(\theta_y \mathbf{x})$	$\overline{E}(\theta_y \mathbf{x})$
A	???	???
B	???	???
T	???	???

Compute Lower and Upper Expectations (Solution 8)

	D_1	D_2	D_3	D_4	D_5	D_6	D_7	D_8
n_A	0	0	1	1	1	2	2	2
n_B	3	4	2	3	4	2	3	4
n_T	4	3	4	3	2	4	3	3

Using IDM to estimate **interval posterior mean** $\theta_y^*|\mathcal{D}$ of $\theta_y|\mathcal{D}$:

$$\underline{E}(\theta_y|\mathbf{x}) = \min_{\mathbf{D} \in \mathcal{D}} \underline{E}(\theta_y|\mathbf{D}) = \min_{\mathbf{D} \in \mathcal{D}} n_y/(n+s), \quad (9)$$

$$\bar{E}(\theta_y|\mathbf{x}) = \max_{\mathbf{D} \in \mathcal{D}} \bar{E}(\theta_y|\mathbf{D}) = \max_{\mathbf{D} \in \mathcal{D}} (n_y+s)/(n+s). \quad (10)$$

	$\underline{E}(\theta_y \mathbf{x})$	$\bar{E}(\theta_y \mathbf{x})$
A	$0/(7+s)$	$(2+s)/(7+s)$
B	$2/(7+s)$	$(4+s)/(7+s)$
T	$2/(7+s)$	$(4+s)/(7+s)$

Outline

- Imprecise Dirichlet Model (IDM)
- Applications in classification tasks
- Evaluate Classifiers

Set-Based Utility Functions (Exercise 9)

Recap: Few commonly used **utility functions**:

$$g_{\alpha}(|Y|) = \frac{\alpha}{|Y|} - \frac{\alpha - 1}{|Y|^2}.$$

Exercise: The maximum value of α such that $g_{\alpha}(|Y|) \leq 1, \forall Y \subset \mathcal{Y} \setminus \emptyset$.

References I