

MATH5905 - Assignment 1

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I declare that all work on this assignment is my own, unless acknowledged. I have read and understand the University Rules in respect to Student Academic Misconduct.

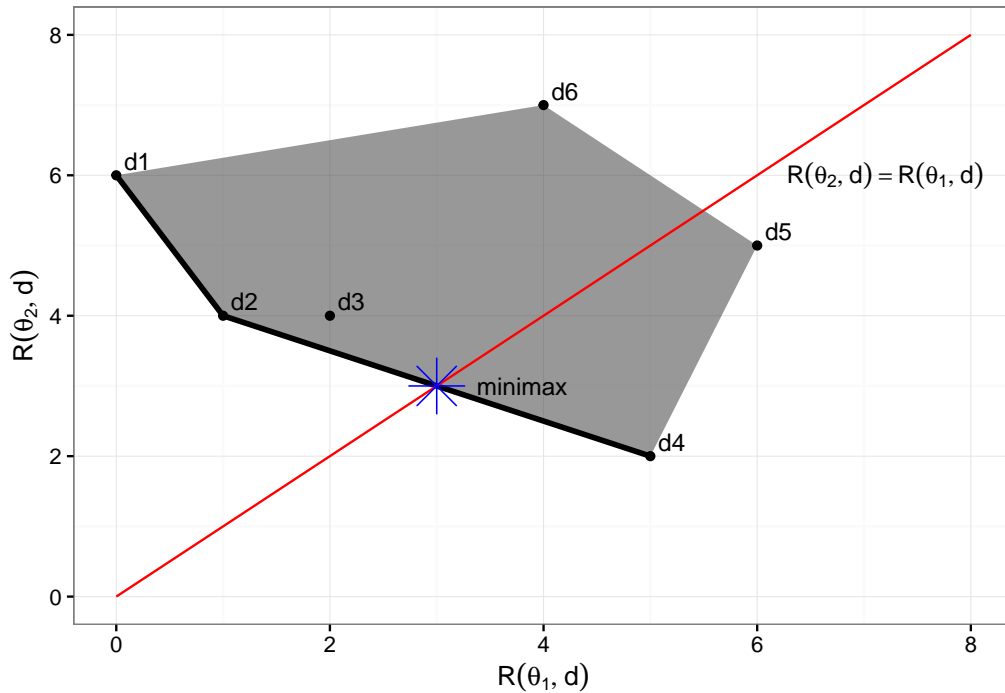
1. Consider a decision problem with parameters space $\Theta = \{\theta_1, \theta_2\}$ and a set of non randomised decisions $D = \{d_i, 1 \leq i \leq 6\}$ with risk points $\{R(\theta_1, d_i), R(\theta_2, d_i)\}$ as follows:

i	1	2	3	4	5	6
$R(\theta_1, d_i)$	0	1	2	5	6	4
$R(\theta_2, d_i)$	6	4	4	2	5	7

- a) Find the minimax rule(s) amongst the **nonrandomized** rules in D ;

The minimax rules are the set $\{d_2, d_3\}$ since the minimal between the 6 maximum values is $\{6, 4, 4, 5, 6, 7\}$.

- b) Plot the risk set of all **randomised** rules \mathcal{D} generated by the set of rules in D .



The risk set of all randomised rules generated by the rules in D is given by the shaded area generated via the convex, S . The points of the decision rules show the non-randomised decision rules, while the admissible decision rules are given along the shaded lower left boundary of points $\{d_1, d_2, d_4\}$.

- c) Find the risk point of the minimax rule in \mathcal{D} and determine its minimax risk.

Using points $(1, 4)$, $(5, 2)$, the slope of the curve is $m = \frac{4-2}{1-5} = -\frac{1}{2}$. We can then use $(1, 4)$ to find the intercept, $R_2 = -0.5R_1 + c \rightarrow c = 4.5$. Therefore we have to solve for the equations $R_1 = R_2$

and $R_2 = -0.5R_1 + 4.5$ to find the intersect.

$$R_1 = -0.5R_1 + 4.5 \rightarrow R_1 = 3$$

Therefore the minimax risk is equal to 3.

- d) Define the minimax rule in the set \mathcal{D} in terms of rules in D.

The minimax rule is a randomised rule of the form $d^* = \alpha d_2 + (1 - \alpha)d_4$, such that $R(\theta_1, d^*) = R(\theta_2, d^*)$. This point is marked in S , and represents the randomised decision rule.

We can now find alpha, to express the minimax rule in terms of the rules in D:

$$\alpha d_2 + (1 - \alpha)d_4 = 3$$

$$\text{Solving for } (1, 4)$$

$$\alpha \cdot 1 + (1 - \alpha) \cdot 4 = 3$$

$$\alpha = \frac{1}{3}$$

So we choose d_2 with probability $\frac{1}{3}$ And we choose d_4 with probability $\frac{2}{3}$

- e) For which prior on $\{\theta_1, \theta_2\}$ is the minimax rule a Bayes rule?

The minimax rule occurs along the line $R_2 = -\frac{1}{2}R_1 + c$. Given the Bayes rule can be found along $\alpha R_1 + (1 - \alpha)R_2 = c$, we need to set α such that:

$$\frac{\alpha}{1 - \alpha} = \frac{1}{2} \rightarrow \alpha = \frac{1}{3}$$

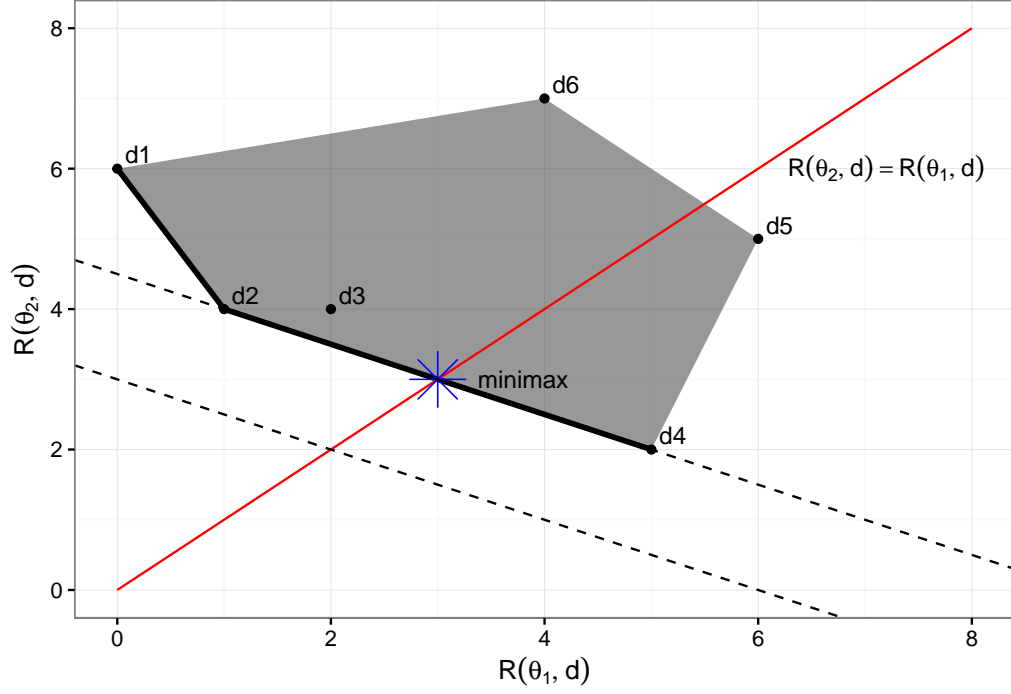
.

Then the prior is $\{\theta_1, \theta_2\} = \{\frac{1}{3}, \frac{2}{3}\}$.

$$\frac{1}{3}R_1 + \frac{2}{3}R_2 = c$$

$$R_2 = \frac{1}{2}R_1 + \frac{3}{2}c$$

With these priors and $c = 4.5$ the Bayes rule will include the minimax rule.



f) Determine the Bayes rule and the Bayes risk for the prior $(\frac{4}{5}, \frac{1}{5})$ on $\{\theta_1, \theta_2\}$.

The Bayes rule is determined by the straight line $\pi_1 R_1 + \pi_2 R_2 = c$ representing a class of decision rules with the same Bayes Risk. By varying c , we get the family of straight lines, choosing the one that intersects S . In this case, $c = 6$

$$\frac{4}{5}R_1 + \frac{1}{5}R_2 = c$$

Solving for the point (0,6)

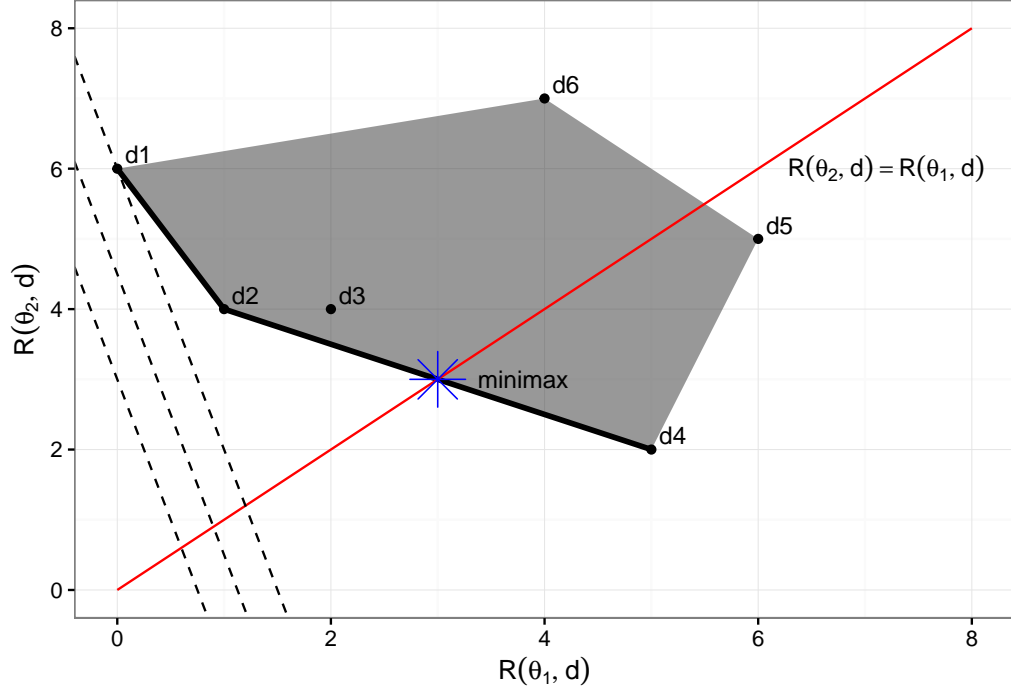
$$\frac{1}{5} \cdot 6 = c \rightarrow c = \frac{6}{5}$$

So the optimal line is:

$$\begin{aligned} \frac{4}{5}R_1 + \frac{1}{5}R_2 &= \frac{6}{5} \\ R_2 &= -4R_1 + 6 \end{aligned}$$

This intersects the admissible rules with d_1 , so the bayes risk given the prior is:

$$\frac{4}{5}R(\theta_1, d_1) + \frac{1}{5}R(\theta_2, d_1) = 1.2$$



g) For $\epsilon = \frac{4}{5}$, illustrate on the risk set the risk points of all rules which are ϵ -minimax.

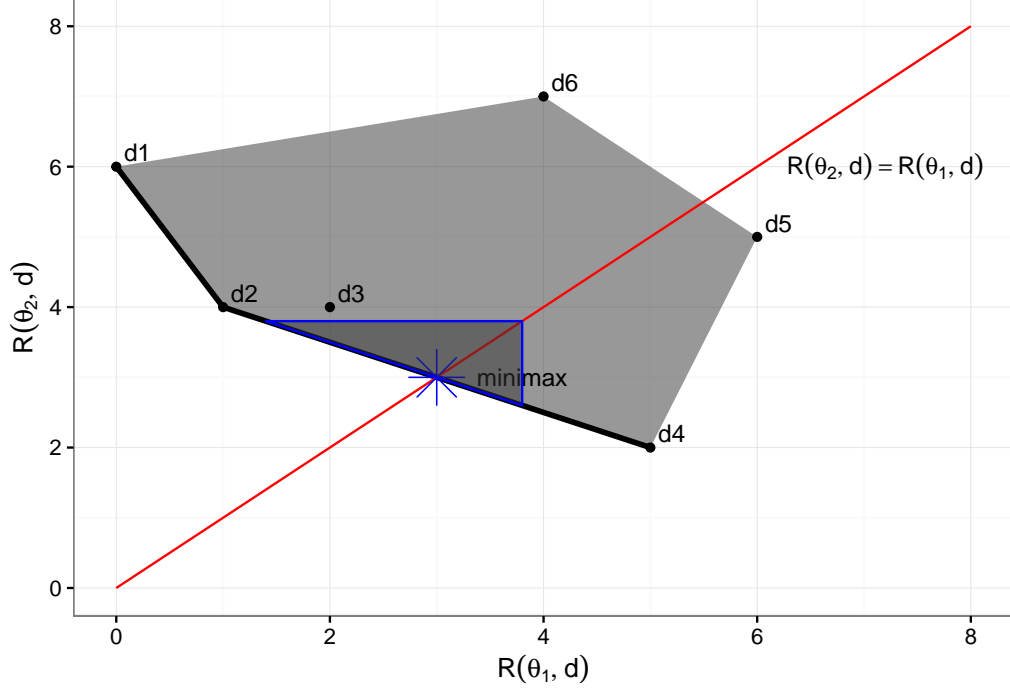
The ϵ -minimax decision rule is equal to

$$\sup_{\theta \in \Theta} R(\theta, \delta_\epsilon) \leq \inf_{\delta \in \mathcal{D}} \sup_{\theta \in \Theta} R(\theta, \delta) + \epsilon$$

The δ_ϵ rules are those those that are:

$$\sup_{\theta \in \Theta} R(\theta, \delta_\epsilon) \leq 3 + \frac{4}{5}$$

These include $\{d_2, d_3\}$



2. In a sequence of consecutive years $1, 2, \dots, T$ an annual number of high-risk events is recorded by a bank. The random number N_t of high-risk events in a given year is modelled via $\text{Poisson}(\lambda)$ distribution. This gives a sequence of independent counts n_1, n_2, \dots, n_T . The prior on λ is $\text{Gamma}(a, b)$ with known $a > 0, b > 0$:

$$\tau(\lambda) = \frac{\lambda^{a-1} e^{-\lambda/b}}{\Gamma(a) b^a}, \lambda > 0.$$

- a) Determine the Bayesian estimator of the intensity λ with respect to quadratic loss.

The distribution of the T events is given by:

$$\begin{aligned} f(x_1, \dots, x_n | \lambda) &= \frac{\lambda^{\sum_{i=1}^n x_i} e^{-\lambda n}}{\prod_{i=1}^n x_i!} \\ f(x_1, \dots, x_n, \lambda) &= f(X | \lambda) \tau(\lambda) \\ &= \frac{\lambda^{\sum_{i=1}^n x_i + a - 1} e^{-\lambda(n+1/b)}}{\prod_{i=1}^n x_i! \Gamma(a) b^a} \\ h(x_1, \dots, x_n) &= \int_0^\infty f(x_1, \dots, x_n, \lambda) \partial \lambda \\ &= \int_0^\infty \frac{\lambda^{\sum_{i=1}^n x_i + a - 1} e^{-\lambda(n+1/b)}}{\prod_{i=1}^n x_i! \Gamma(a) b^a} \partial \lambda \\ &= \frac{1}{\prod_{i=1}^n x_i! \Gamma(a) b^a} \int_0^\infty \lambda^{\sum_{i=1}^n x_i + a - 1} e^{-\lambda(n+1/b)} \partial \lambda \end{aligned}$$

Noticing that the integral is a gamma distribution, we can multiply by a constant and, and integrating over the entire distribution is equal to 1.

$$h(x_1, \dots, x_n) = \frac{\Gamma(\sum_{i=1}^n x_i + a)}{\prod_{i=1}^n x_i! \Gamma(a) b^a (n + 1/b)^{\sum_{i=1}^n x_i + a}}$$

The conditional prior distribtuion is therefore:

$$\begin{aligned}
\tau(\lambda|x_1, \dots, x_n) &= \frac{f(x_1, \dots, x_n, \lambda)}{h(x_1, \dots, x_n)} \\
&= \frac{\lambda^{\sum_{i=1}^n x_i + a - 1} e^{-\lambda(n+1/b)}}{\prod_{i=1}^n x_i! \Gamma(a) b^a} / \frac{\Gamma(\sum_{i=1}^n x_i + a)}{\prod_{i=1}^n x_i! \Gamma(a) b^a (n+1/b)^{\sum_{i=1}^n x_i + a}} \\
&= \frac{(n+1/b)^{\sum_{i=1}^n x_i + a}}{\Gamma(\sum_{i=1}^n x_i + a)} \lambda^{\sum_{i=1}^n x_i + a - 1} e^{-\lambda(n+1/b)} \sim \text{Gamma}(a + \sum x_i, n+1/b)
\end{aligned}$$

The Bayes estimator with respect to quadratic loss is equal to the expected value of the conditional prior distribution $E(\lambda) = \int_{\Lambda} \lambda \tau(\lambda|X) d\lambda$. For a Gamma distribution, the mean is given by $\frac{\alpha}{\beta}$. Therefore;

$$\hat{\lambda}_{\text{Bayes}} = \frac{a + \sum x_i}{n + 1/b}$$

- b) Assume $a = 3, b = 2$. Within the last seven years counts were 2, 4, 7, 3, 4, 4, 5. Find the Bayes's estimate of the intensity λ using quadratic loss and the above data.

$$\hat{\lambda}_{\text{Bayes}} = \frac{3 + 29}{7 + 1/2} = \frac{64}{15} \approx 4.267$$

- c) The bank claims that the yearly intensity λ is less than 4. Test the bank's claim via Bayesian testing with a zero-one loss.

The null hypothesis is:

$$H_0 : \lambda < 4$$

$$H_1 : \lambda > 4$$

Testing the hypothesis:

$$\begin{aligned}
\tau(\lambda \in \Lambda_0 | X) &= \int_0^4 \frac{(n+1/b)^{\sum_{i=1}^n x_i + a}}{\Gamma(\sum_{i=1}^n x_i + a)} \lambda^{\sum_{i=1}^n x_i + a - 1} e^{-\lambda(n+1/b)} d\lambda \\
&= \int_0^4 \frac{(7+1/2)^{29+3}}{\Gamma(29+3)} \lambda^{29+3-1} e^{-\lambda(7+1/2)} d\lambda
\end{aligned}$$

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```r
fn <- function(x) ((7.5)^32*x^(31)*exp(-x*(7.5)))/(gamma(32));
integrate(fn,0,4)
```

### 0.381357 with absolute error < 2.6e-06

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The test is rejected if $H_0 < \frac{1}{2}$, so we reject the null hypothesis that the intensity λ is less than 4 and accept the alternative that $\lambda > 4$.

3. Let X_1, X_2, \dots, X_n be i.i.d. uniform in $(0, \theta)$ and let the prior on θ be the Pareto prior given by $\tau(\theta) = \beta \alpha^\beta \theta^{-(\beta+1)}, \theta > \alpha$. (Here $\alpha > 0$ and $\beta > 0$ are assumed to be known). Show that the Bayes estimator with respect to quadratic loss is given by $\hat{\theta} = \max(\alpha, x_{(n)}) \frac{n+\beta}{n+\beta-1}$. Justify all steps in derivation.

$$f(x_1, \dots, x_n | \theta) = \theta^{-n}, \theta > x_{(n)}$$

$$\tau(\theta) = \beta \alpha^\beta \theta^{-(\beta+1)}, \theta > \alpha$$

We can calculate the conditional prior distribution:

$$\begin{aligned}
f(x_1, \dots, x_n, \theta) &= \beta \alpha^\beta \theta^{-(\beta+n+1)}, \theta > x_{(n)}, \alpha \\
h(x_1, \dots, x_n) &= \int_{x_{(n)}}^{\infty} \beta \alpha^\beta \theta^{-(\beta+n+1)} \partial \theta \\
&= \begin{cases} \frac{\beta \alpha^\beta}{(n+\beta) \alpha^{n+\beta}} = \frac{\beta}{(n+\beta) \alpha^n}, & \text{if } \alpha > x_{(n)} \\ \frac{\beta \alpha^\beta}{(n+\beta) x_{(n)}^{n+\beta}}, & \text{if } \alpha < x_{(n)} \end{cases} \\
\tau(\theta | x_1, \dots, x_n) &= \begin{cases} \frac{\beta \alpha^\beta \theta^{-(\beta+n+1)}}{\frac{\beta}{(n+\beta) \alpha^n}} = \alpha^{n+\beta} (n+\beta) \theta^{-(\beta+n+1)}, & \text{if } \alpha > x_{(n)} \\ \frac{\beta \alpha^\beta \theta^{-(\beta+n+1)}}{\frac{\beta \alpha^\beta}{(n+\beta) x_{(n)}^{n+\beta}}} = x_{(n)}^{n+\beta} (n+\beta) \theta^{-(\beta+n+1)}, & \text{if } \alpha < x_{(n)} \end{cases} \\
&= \max(\alpha, x_{(n)})^{n+\beta} (n+\beta) \theta^{-(\beta+n+1)} \sim \text{Pareto}(\max(\alpha, x_{(n)}), n+\beta)
\end{aligned}$$

The Bayes estimator with respect to quadratic loss is equal to the expected value of the conditional prior distribution $E(\theta) = \int_{\Theta} \theta \tau(\theta | X) \partial \theta$. For a Pareto distribution, the mean is given by $\frac{\alpha \beta}{\alpha - 1}$. Therefore;

$$\hat{\theta} = \max(\alpha, x_{(n)}) \frac{n + \beta}{n + \beta - 1}$$

4. At a critical stage in the development of a new aeroplane in the UK (jointly contracted with the French), a decision must be taken to continue or to abandon the project. The financial viability of the project can be measured by a parameter $\theta \in (0, 1)$, the project being profitable if $\theta > \frac{1}{2}$, the cost to the taxpayer of continuing the project is $(\frac{1}{2} - \theta)$ (in units of \$ billion) whereas if $\theta > \frac{1}{2}$, it is zero (since the project will be privatised if profitable). Data x provides information about θ : the prototype aeroplane is subjected to trials, each independently having probability θ of success, and the data x consists of the total number of trials required for the first successful result to be obtained (in other words, we observe a single realisation of a geometrically distributed random variable). If $\theta > \frac{1}{2}$ the cost of abandoning the project is $(\theta - \frac{1}{2})$ (due to contractual arrangements for purchasing the aeroplane from the French), whereas if $\theta < \frac{1}{2}$ it is zero. Two actions are on the table: a_0 : continue the project and a_1 : abandon it.

- i) Derive the Bayesian decision rule, for a given prior on θ . In particular, show that the rule is equivalent to comparing the posterior mean of θ given x with a threshold constant.

There are two actions: a_0 : continue the project a_1 : abandon the project.

$$\text{The cost of continuing: } L(\theta, a_0) = \begin{cases} 0, & \text{if } \theta > \frac{1}{2} \\ \frac{1}{2} - \theta, & \text{if } \theta < \frac{1}{2} \end{cases}$$

$$\text{The cost of abandoning: } L(\theta, a_1) = \begin{cases} \theta - \frac{1}{2}, & \text{if } \theta > \frac{1}{2} \\ 0, & \text{if } \theta < \frac{1}{2} \end{cases}$$

Project will be continued if the cost is less than abandoning.

$$\begin{aligned}
E_{\theta}(L(\theta, a_0)) &> E_{\theta}(L(\theta, a_1)) \\
\int_0^{\frac{1}{2}} \left(\frac{1}{2} - \theta \right) \tau(\theta | x) \partial \theta &> \int_{\frac{1}{2}}^1 \left(\theta - \frac{1}{2} \right) \tau(\theta | x) \partial \theta
\end{aligned}$$

$$\begin{aligned}
0 &> \int_0^1 \left(\theta - \frac{1}{2} - \frac{1}{2} + \theta \right) \tau(\theta|x) \partial\theta \\
0 &> \int_0^1 \left(\theta - \frac{1}{2} \right) \tau(\theta|x) \partial\theta \\
0 &> \int_0^1 (\theta) \tau(\theta|x) \partial\theta - \frac{1}{2} \\
\frac{1}{2} &> \int_0^1 (\theta) \tau(\theta|x) \partial\theta = \mu(\theta|x)
\end{aligned}$$

- ii) The Minister of Aviation has prior density $6\theta(1-\theta)$ for θ . The prime minister has prior density $4\theta^3$. For which values of x will there be most serious ministerial disagreement?

There will be disagreement when one Minister wants to abandon the project, while the other wants to continue it.

The conditional distribution on x is

$$f(x; \theta) = \theta(1-\theta)^{x-1}, x = 1, 2, \dots$$

Both the Minister for Aviation and the Prime Minister have beta prior distributions.

For the Minister for Aviation (MA):

$$\tau_{MA}(\theta) = 6\theta(1-\theta) \sim \text{Beta}(2, 2)$$

So calculating the conditional prior distribution:

$$\begin{aligned}
f(x, \theta) &= 6\theta^2(1-\theta)^x \\
h(x) &= \int_0^1 6\theta^2(1-\theta)^x \partial\theta \\
&= 6 \cdot \text{beta}(3, x+1) \\
\tau(\theta|x) &= \frac{6\theta^2(1-\theta)^x}{6 \cdot \text{beta}(3, x+1)} \\
&= \frac{1}{\text{beta}(3, x+1)} \theta^2(1-\theta)^x \sim \text{Beta}(3, x+1)
\end{aligned}$$

The prior distribution for the Prime Minister (PM):

$$\tau_{PM}(\theta) = 4\theta^3 \sim \text{Beta}(5, 1)$$

So calculating the conditional prior distribution:

$$\begin{aligned}
f(x, \theta) &= 4\theta^4(1-\theta)^{x-1} \\
h(x) &= \int_0^1 4\theta^4(1-\theta)^{x-1} \partial\theta \\
&= 4 \cdot \text{beta}(5, x) \\
\tau(\theta|x) &= \frac{4\theta^4(1-\theta)^{x-1}}{4 \cdot \text{beta}(5, x)} \\
&= \frac{1}{\text{beta}(5, x)} \theta^4(1-\theta)^{x-1} \sim \text{Beta}(5, x)
\end{aligned}$$

If $x \sim \text{Beta}(a, b)$ then $E(x) = \frac{a}{a+b}$.

The Minister of Aviation will want to abandon if $\frac{3}{4+x} < \frac{1}{2}$, ie $x > 2$ with $x = 2$ being on the edge. The Prime Minister will want to abandon if $\frac{5}{5+x} < \frac{1}{2}$, ie $x > 5$ with $x = 5$ being on the edge. Therefore there will be the greatest amount of disagreement when $x \in \{3, 4\}$.