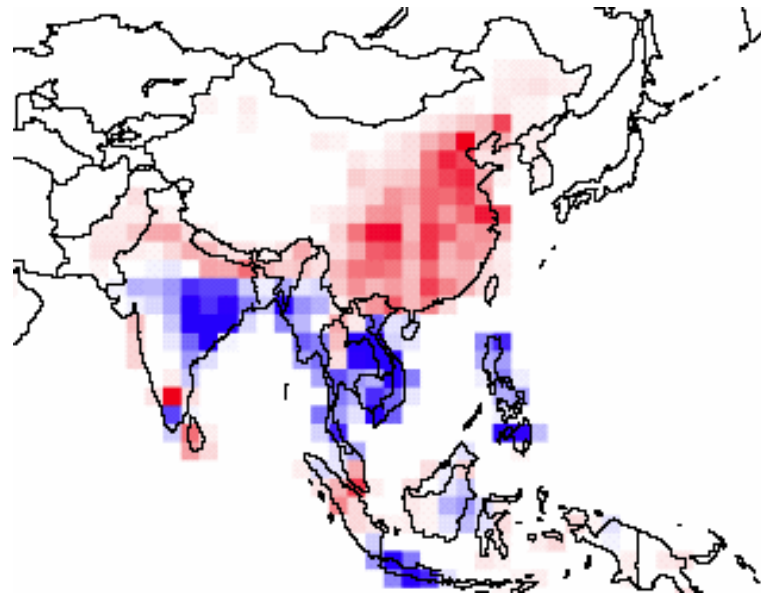


More Information! Conditional Probability and Inverse Methods

Peter Zoogman

Harvard Atmospheres Journal Club

April 12, 2012



Bayesian Analysis – not very intuitive!

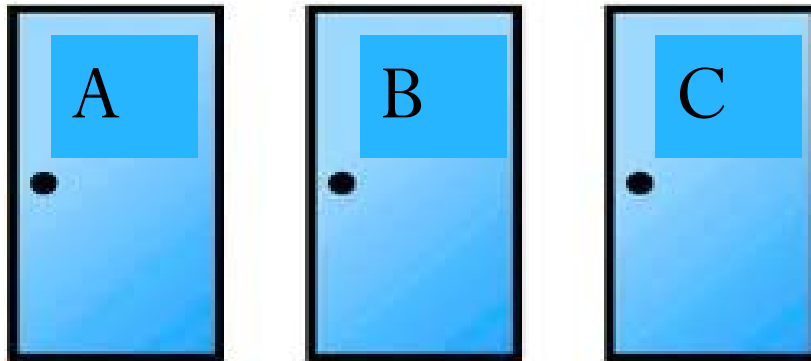
- Humans are bad at combining a priori information with new observations
- Conditional Probability examples
- 2-box inversion to constrain emissions
- Arellano et al paper on application of these methods



The Monty Hall Problem

You're given the choice of three doors:

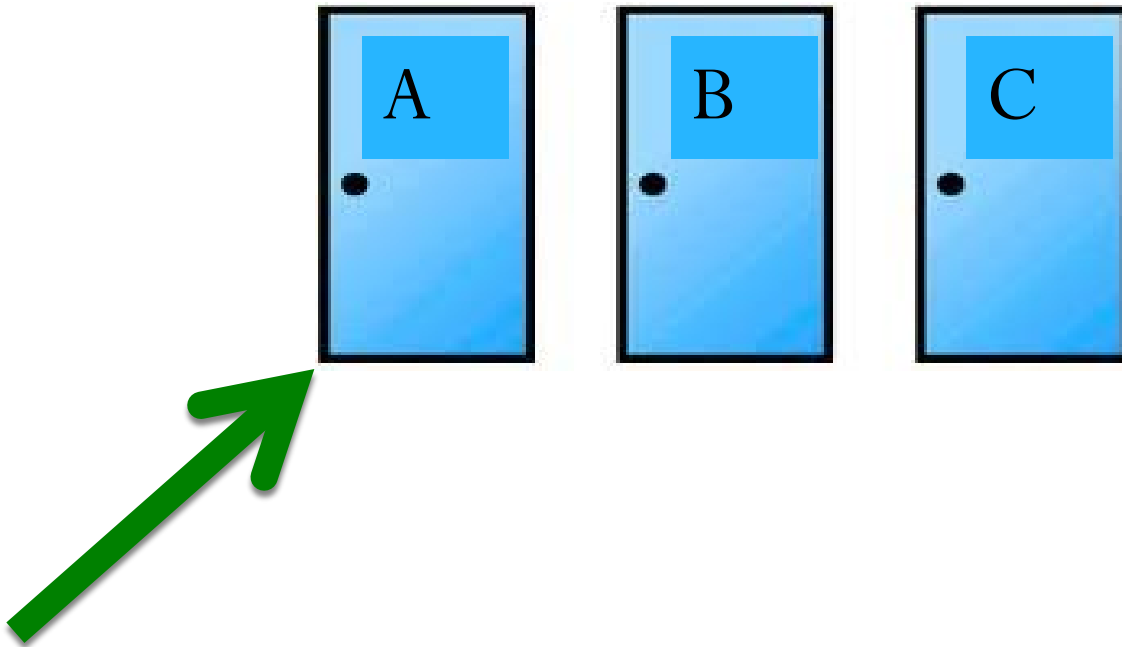
Behind one door is a car; behind the others, goats. You pick a door, say door A [but the door is not opened], and the host, who knows what's behind the doors, opens another door, say door B, which has a goat. He then says to you, "Do you want to pick door C?" Is it to your advantage to switch your choice?



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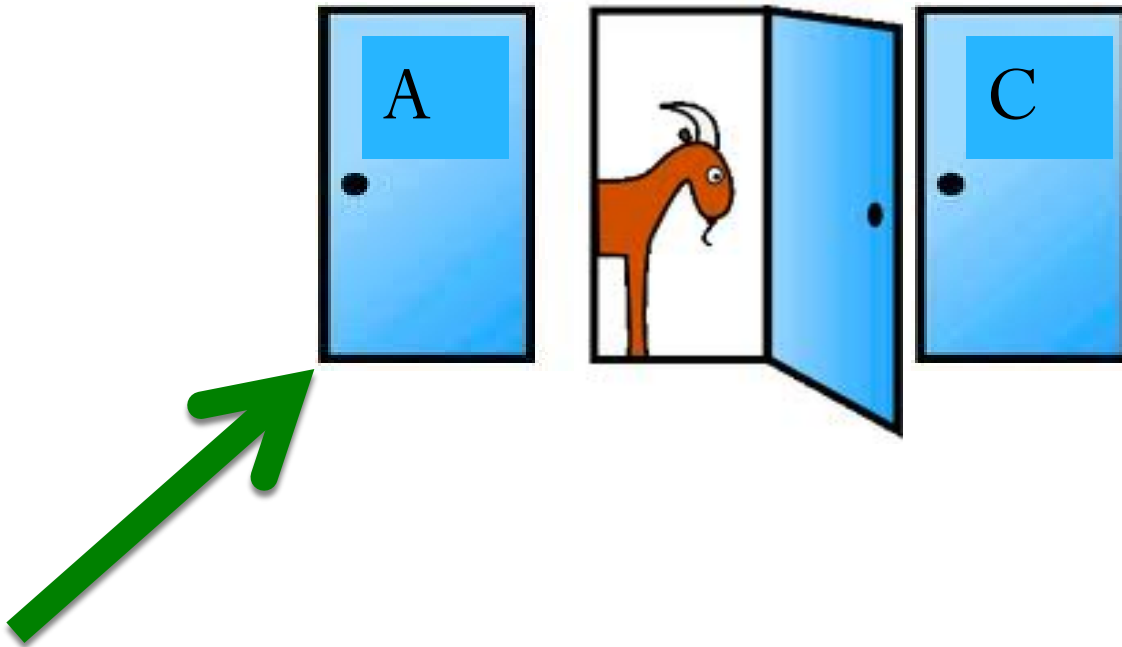
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Bayes' Theorem

$$P(A | B) = \frac{P(B | A)P(A)}{P(B)}$$

$$P(\text{car is behind A} | \text{door B opened}) = \frac{P(\text{door B opened} | \text{car is behind A})P(\text{car is behind A})}{P(\text{door B opened})}$$

$$P(\text{door B opened} | \text{car is behind A}) = 0.5$$

$$P(\text{car is behind A} | \text{door B opened}) = \frac{0.5 * 0.33}{0.5} = 0.33$$

Why is door C different?

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Why is door C different?

$$P(\text{door B opened} | \text{car is behind C}) = 1$$

Updating an Initial Guess

- A cab was involved in a hit-and-run accident at night
 - 85% of the cabs in the city are Green and 15% are Blue
 - A witness identified the cab as Blue. Tests show that this witness correctly identifies the color of the cab 80% of the time

What is the probability the cab was involved in the accident was Blue?



Set-up from Thinking, Fast and Slow (Kahneman)

Combine observation with a priori information

$$P(\text{car is Blue} \mid \text{witnessed Blue}) = \frac{P(\text{witnessed Blue} \mid \text{car is Blue})P(\text{car is blue})}{P(\text{witnessed Blue})}$$

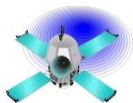
$$P(\text{car is blue}) = 0.15$$

$$P(\text{witnessed Blue} \mid \text{car is Blue}) = 0.8$$

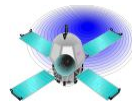
$$P(\text{witnessed Blue}) = 0.8 * 0.15 + 0.2 * 0.85 = 0.29$$

$$P(\text{car is Blue} \mid \text{witnessed Blue}) = \frac{0.8 * 0.15}{0.29} = 0.41$$

Constraining Emissions

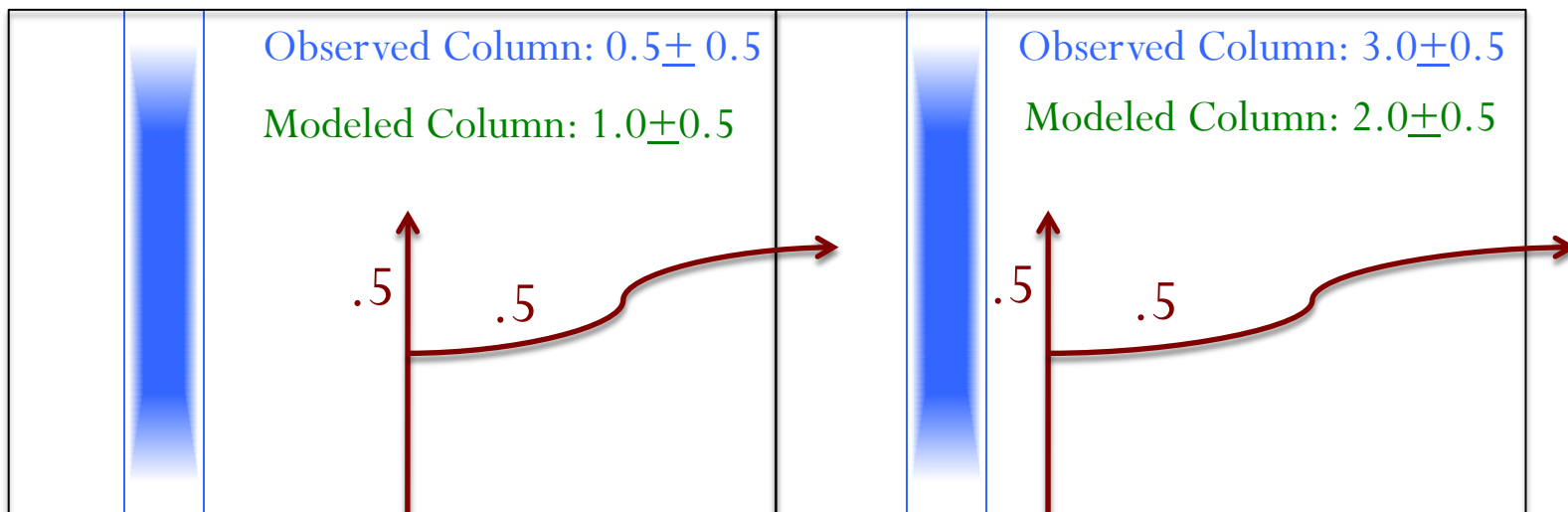


Western US



Eastern US

Column Units:
 10^{18} molec/cm²



Emissions = 1 ± 1

Emissions = 1 ± 1

Emissions units 100 Tg/yr

Modeled Column ~ Western
Emissions

Modeled Column ~ Western
Emissions + Eastern Emissions

Bayesian Inversion

$$P(X(\text{emissions}) | Y(\text{observations})) = \frac{P(Y | X)P(X)}{P(Y)}$$

$$\mathbf{K}^T \mathbf{S}_\varepsilon^{-1} (\mathbf{K} \hat{\mathbf{x}} - \mathbf{y}) + \mathbf{S}_a^{-1} (\hat{\mathbf{x}} - \mathbf{x}_a) = 0$$

$$\mathbf{y} = \mathbf{K}\mathbf{x} = \begin{pmatrix} 1 & 0 \\ 1 & 1 \end{pmatrix} \begin{pmatrix} x_1 \\ x_2 \end{pmatrix} = \begin{pmatrix} x_1 \\ x_1 + x_2 \end{pmatrix}$$

$$\hat{\mathbf{x}} = \mathbf{x}_a + \mathbf{G}(\mathbf{y} - \mathbf{K}\hat{\mathbf{x}}_a)$$

$$\mathbf{G} = \mathbf{S}_a \mathbf{K}^T (\mathbf{K} \mathbf{S}_a \mathbf{K}^T + \mathbf{S}_\varepsilon)^{-1}$$

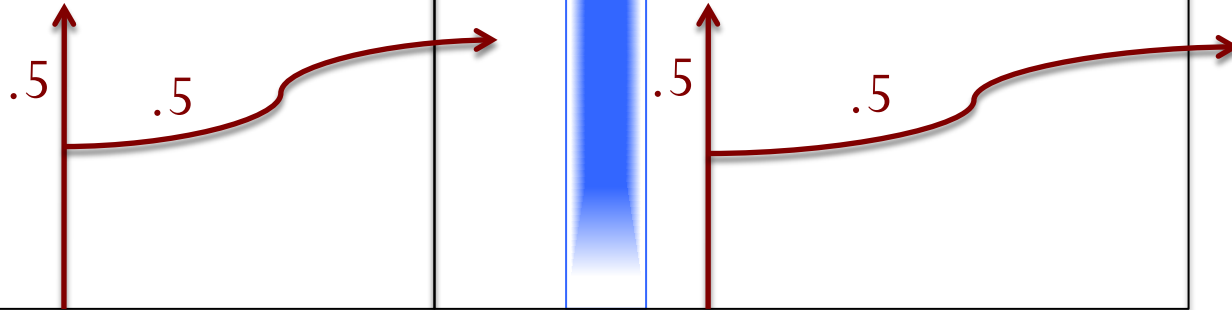
Constraining Emissions

Observed Column: 0.5 ± 0.5

Modeled Column: 1.0 ± 0.5

Observed Column: 3.0 ± 0.5

Modeled Column: 2.0 ± 0.5



Emissions = 1 ± 1

Emissions = 1 ± 1

$$\mathbf{S}_a = \begin{pmatrix} 1^2 & 0 \\ 0 & 1^2 \end{pmatrix}$$

$$\mathbf{S}_\varepsilon = \begin{pmatrix} 2 * 0.5^2 & 0 \\ 0 & 2 * 0.5^2 \end{pmatrix}$$

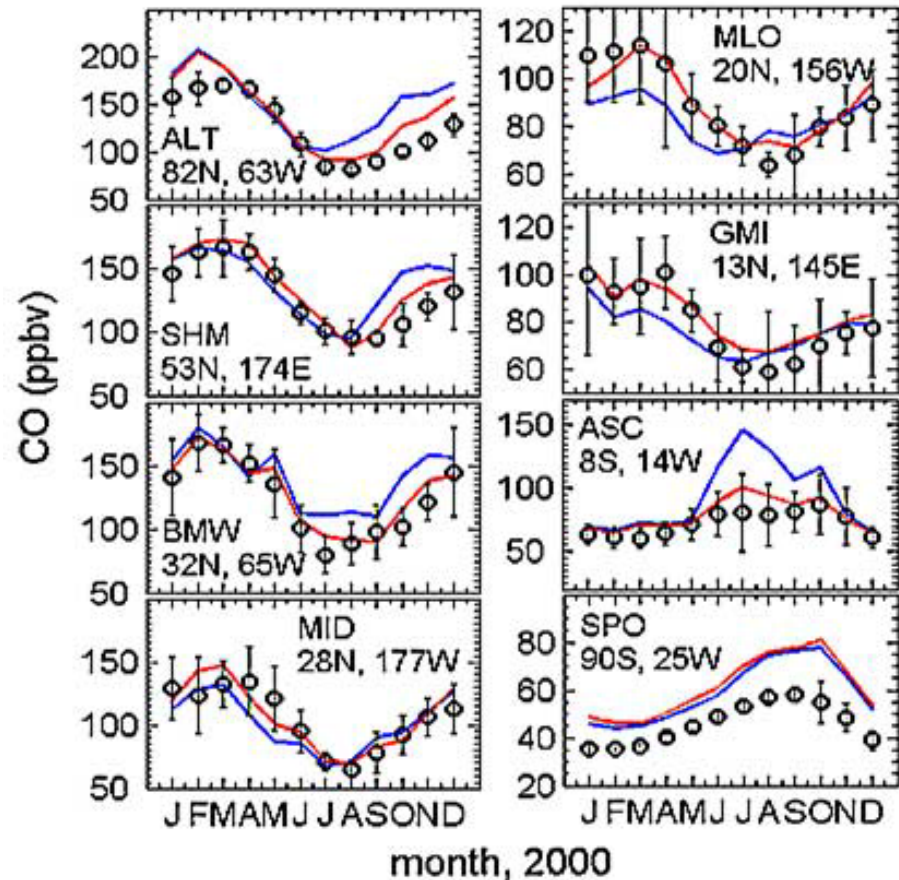
$$\mathbf{G} = \mathbf{S}_a \mathbf{K}^T (\mathbf{K} \mathbf{S}_a \mathbf{K}^T + \mathbf{S}_\varepsilon)^{-1} = \begin{pmatrix} .54 & .18 \\ -.36 & .54 \end{pmatrix}$$

$$(\mathbf{y} - \mathbf{K} \hat{\mathbf{x}}_a) = \begin{pmatrix} .5 \\ 3 \end{pmatrix} - \begin{pmatrix} 1 & 0 \\ 1 & 1 \end{pmatrix} \begin{pmatrix} 1 \\ 1 \end{pmatrix} = \begin{pmatrix} -.5 \\ 1 \end{pmatrix}$$

$$\hat{\mathbf{x}} = \mathbf{x}_a + \mathbf{G}(\mathbf{y} - \mathbf{K} \hat{\mathbf{x}}_a) = \begin{pmatrix} .91 \\ 1.72 \end{pmatrix}$$

Estimating CO Sources

- Why CO?
 - Strong link between free troposphere and regional emissions
 - Relatively easy to observe from space
- Important to compare results using optimized sources to independent data sets
- CO can be used to optimize emissions/concentrations of other species
 - CO₂, ozone



[Arellano et al 2004]