Agua/Tierra



Tomado de: Statistical Rethinking – Richard McElreath, Capítulos 2 y 3

Agua/Tierra

- Hay un planeta con una atmósfera tan densa que no nos permite ver su superficie
- Queremos saber la cobertura de agua en ese planeta
- Enviamos sondas que aterrizan de manera aleatoria sobre la superficie del planeta
- Cada sonda nos informa si aterriza sobre agua o sobre tierra
- ¿Podemos inferir la proporción de cobertura de agua de el planeta con sólo pocas medidas?

Agua/Tierra

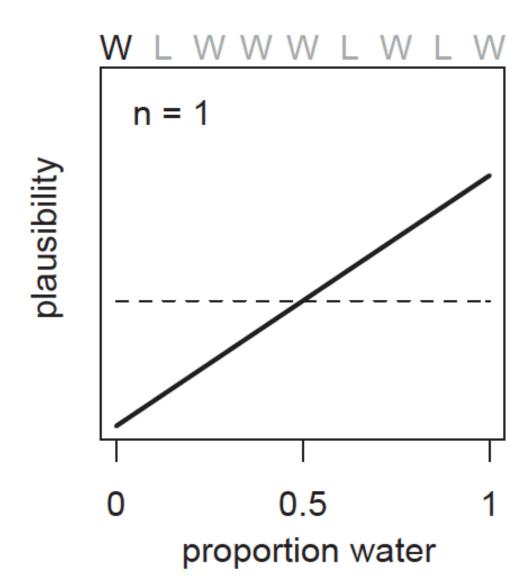


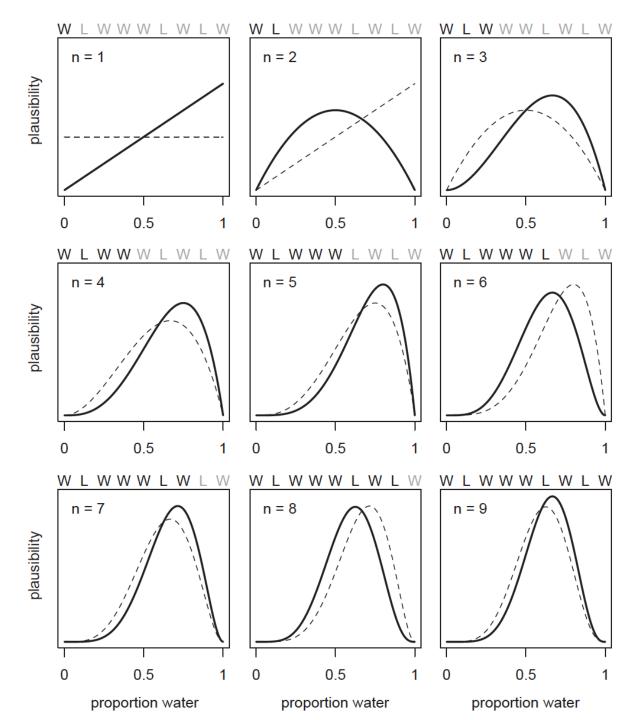
Datos: W L W W W L W L W

Historia para los datos

- (1) La verdadera proporción de superficie cubierta por agua es p
- (2) Cada sonda que aterriza tiene una probabilidad p de producir una observación de agua (W) y una probabilidad 1-p de producir una observación de tierra (L)
- (3) Cada aterrizaje de sonda es independiente de los otros

Actualizar





El proceso de inferencia

- (1) El número de maneras en que cada conjetura puede producir una observación
- (2) El número acumulado de maneras en que cada conjetura puede producir todos los datos
- (3) La plausibilidad inicial para cada conjetura

Verosimilitud

- Escogemos una expresión matemática que pueda explicar (generar) las observaciones
- En este caso hay dos opciones para cada dato
- Cada sonda es independiente de las otras
- La probabilidad p de observar W es la misma en todos los lanzamientos

-> Distribución binomial

Verosimilitud

 Probabilidad de que dado un valor de p, haya un número w de observaciones de W en n lanzamientos

$$\Pr(w|n,p) = \frac{n!}{w!(n-w)!} p^w (1-p)^{n-w}$$

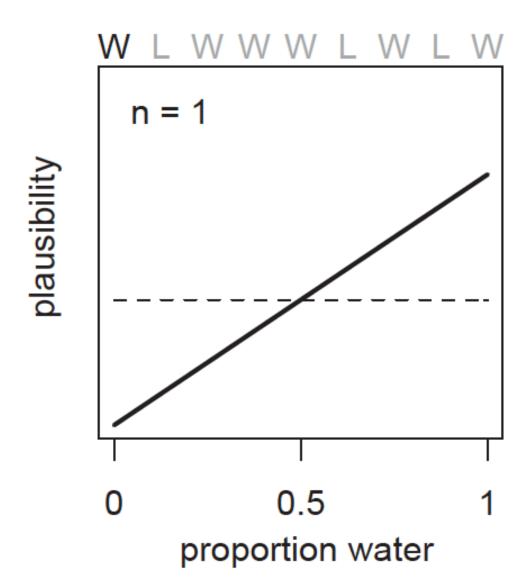
Prior

 Todos los valores en el rango [0,1] son igualmente probables

$$\Pr(p) = \frac{1}{1-0} = 1$$

Prior débilmente informativo/a

Prior: Línea punteada



Posterior

 Objetivo: Dados los datos (!) ¿cuál es la probabilidad de que el parámetro tenga cierto valor?

• Regla de Bayes (obviando n):

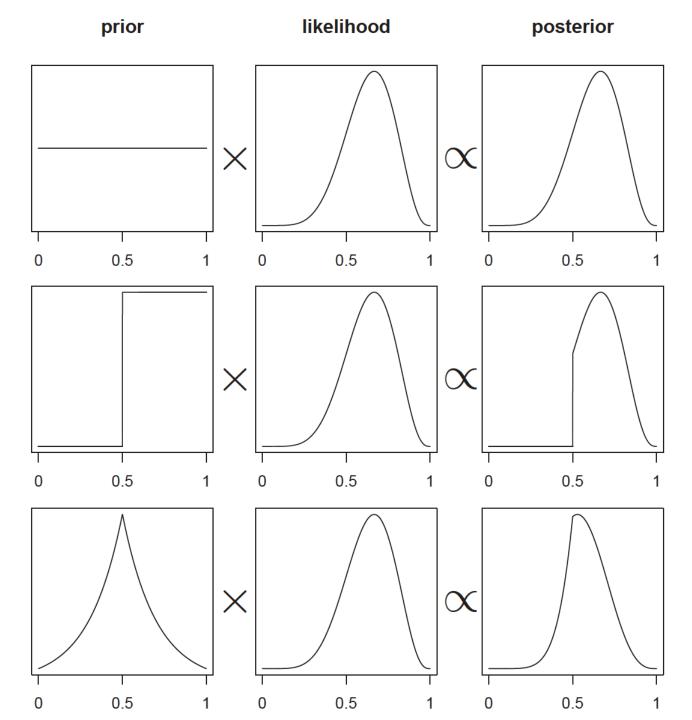
$$Pr(p|w) = \frac{Pr(w|p) Pr(p)}{Pr(w)}$$

Posterior

$$Posterior = \frac{Verosimilitud \times Prior}{Verosimilitud promedio}$$

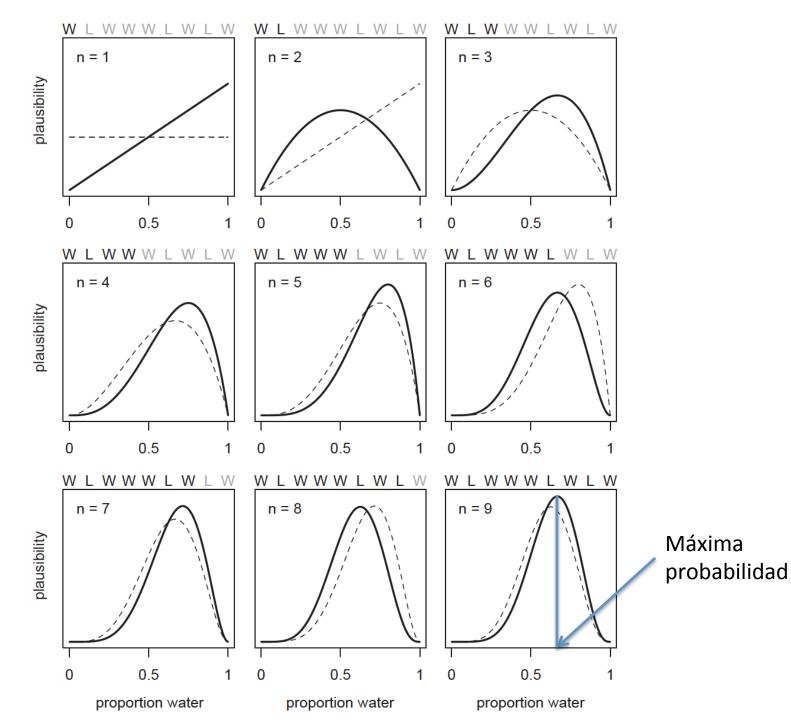
$$Pr(w) = E(Pr(w|p)) = \int Pr(w|p) Pr(p) dp$$

Sirve para que la probabilidad posterior sume 1



Estimación de Posterior

- Grid (fuerza bruta)
- Maximum Likelihood Estimation (frecuentista)
- Monte Carlo (eficiente)
- Markov Chain Monte Carlo (muy eficiente)



Máxima Verosimilitud

- Aproximación no bayesiana (frecuentista)
- Optimización (hallar el máx/min de la posterior)
- Valor óptimo p = 2/3
- ¿Por qué?





build an understanding of statistics and you too can cripple you publication record while also alienating all your colleagues

7:43 PM - 3 Apr 2019



Recommended steps in the statistical analysis of scientific data

- exploration of the data
- careful statement of the scientific problem
- model formulation in mathematical form
- choice of statistical method(s)
- calculation of statistical quantities
- judicious scientific evaluation of the results

Statistics: Some basic definitions

Statistical inference

- Seeking quantitative insight & interpretation of a dataset

Hypothesis testing

To what confidence is a dataset consistent with a previously stated hypothesis?

Estimation

 Seeking the quantitative characteristics of a functional model designed to explain a dataset. An estimator seeks to approximate the unknown parameters based on the data

Probability distribution

 A parametric functional family describing the behavior of a parent distribution of a dataset (e.g. Gaussian = normal)

Nonparametric statistics

- Inference based directly on the dataset without parametric models
 Independent & identically distributed (iid) data point
- A sample of similarly but independently acquired quantitative measurements.

· Frequentist statistics

- Suite of classical inference methods based on simple probability distributions. Fixed hypotheses.

.Bayesian statistics

- Inference methods based on Bayes' Theorem based on likelihoods and prior distributions. Changing hypotheses.

\cdot L₁ and L₂ methods

- 19th century methods for estimation based on minimizing the absolute or squared deviations between a sample and a model.

Maximum likelihood methods

- 20th century methods for parametric estimation based on the likelihood that a dataset fits the model (often like L_2).
- · Gibbs sampling, Metropolis-Hastings algorithm, Markov chain Monte Carlo, ...
 - New computational methods useful for integrations over hypothesis space in Bayesian statistics.

- · Robust (nonparametric) methods
 - Statistical procedures that are insensitive to data outliers or distributions
- Model selection & validation
 - Procedures for estimating the goodness-of-fit and choice of parametric model. (Nested vs. non-nested models, model misspecification)
- Statistical power, efficiency & bias
 - Mathematical evaluation of the effectiveness of a statistical procedure to achieve its desired goals
- · Two-sample & k-sample tests
- Statistical tests giving probabilities that k samples are drawn from the same parent sample
- Independent & identically distributed (i.i.d.) data point
 - -A sample of similarly but independently acquired quantitative measurements.
- ·Heteroscedasticity
 - A failure of i.i.d. due to differently weighted data points, common in astronomy due to measurement errors with known variances

Applied statistics methods

- Multivariate analysis
 - Establishing the structure of a table of rows & columns
 - Analysis of variance, regression, principal component analysis, discriminant analysis, factor analysis
- Multivariate classification
 - Dividing a multivariate dataset into distinct classes
- Correlation & regression
 - Establishing the relationships between variables in a sample
- Time series analysis
 - Studying data measured along a time-like axis
- Spatial analysis
 - Studying point or continuous processes in 2-3-dimensions
- Survival analysis
 - Studying data subject to censoring (e.g. upper limits)
- Data mining
 - Studying structures in mega-datasets