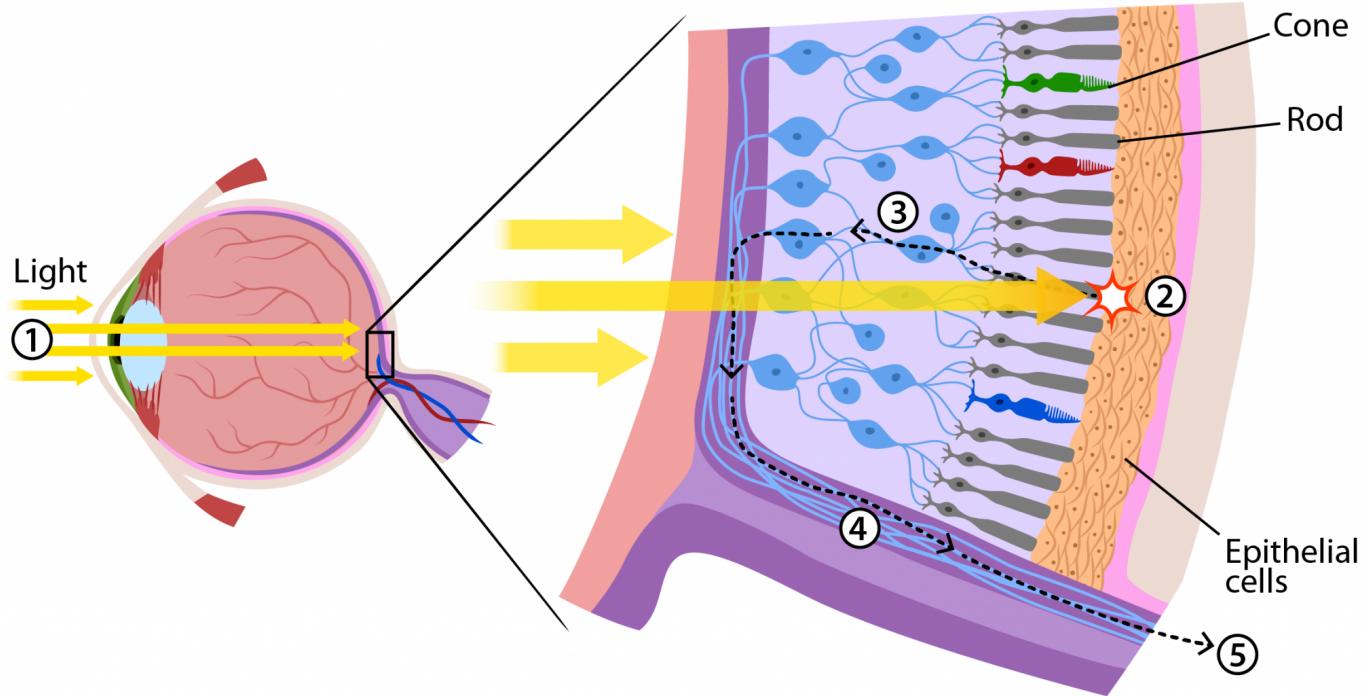


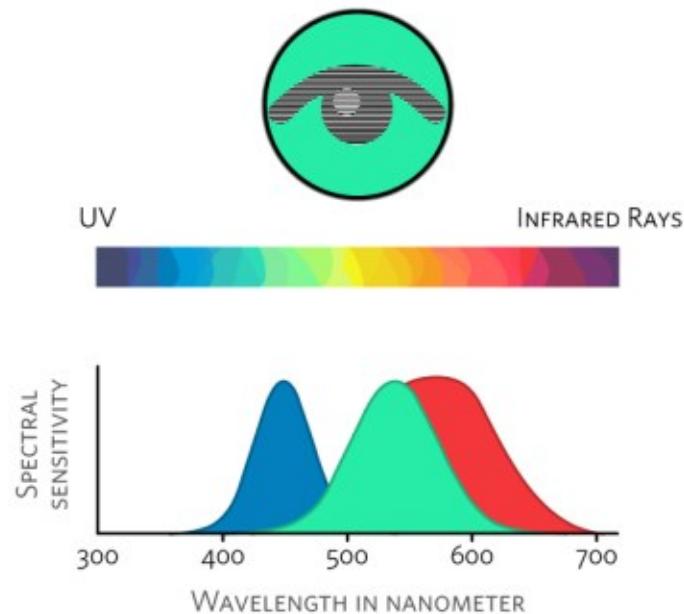
Percepción



Visión



Visión



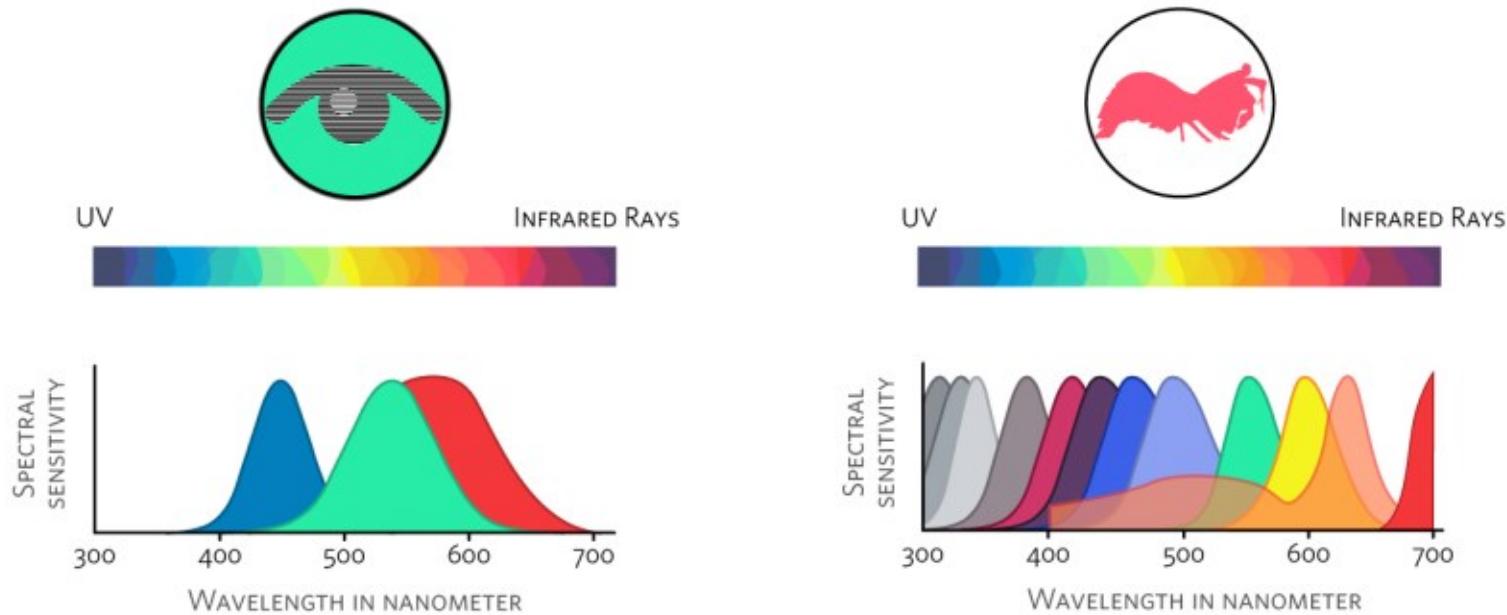
Camarón mantis



Camarón mantis



Camarón mantis



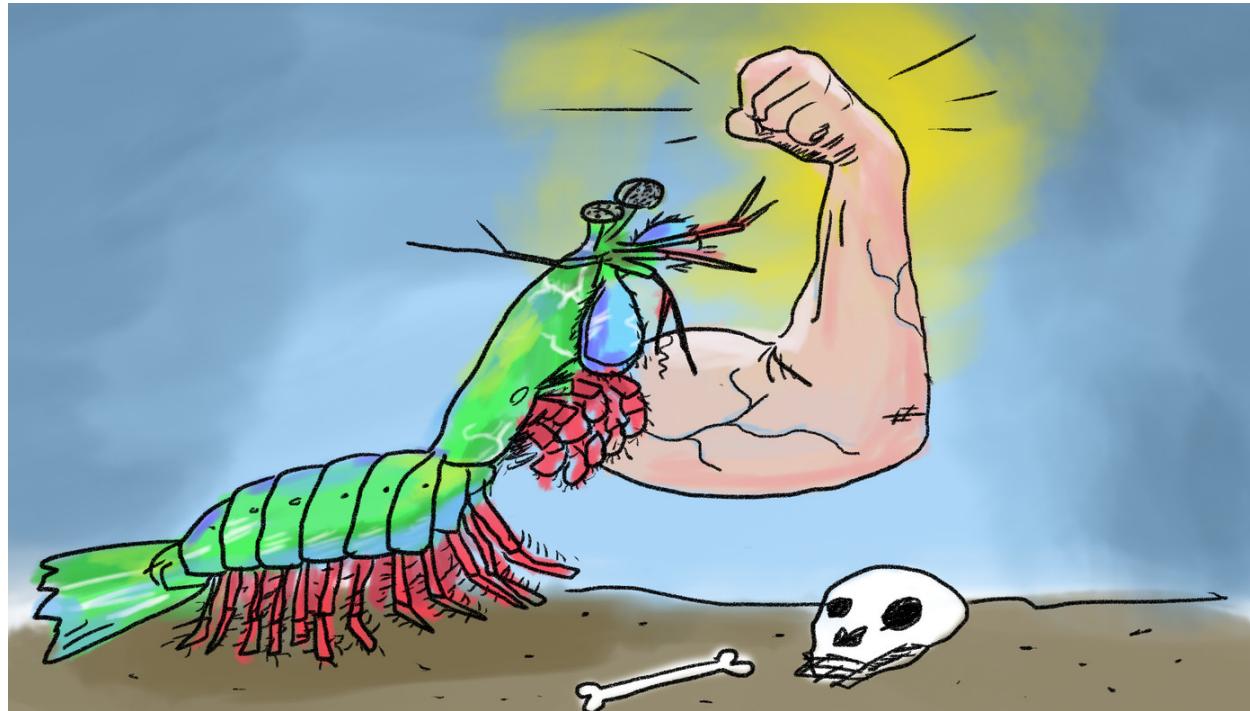
Camarón mantis



Camarón mantis



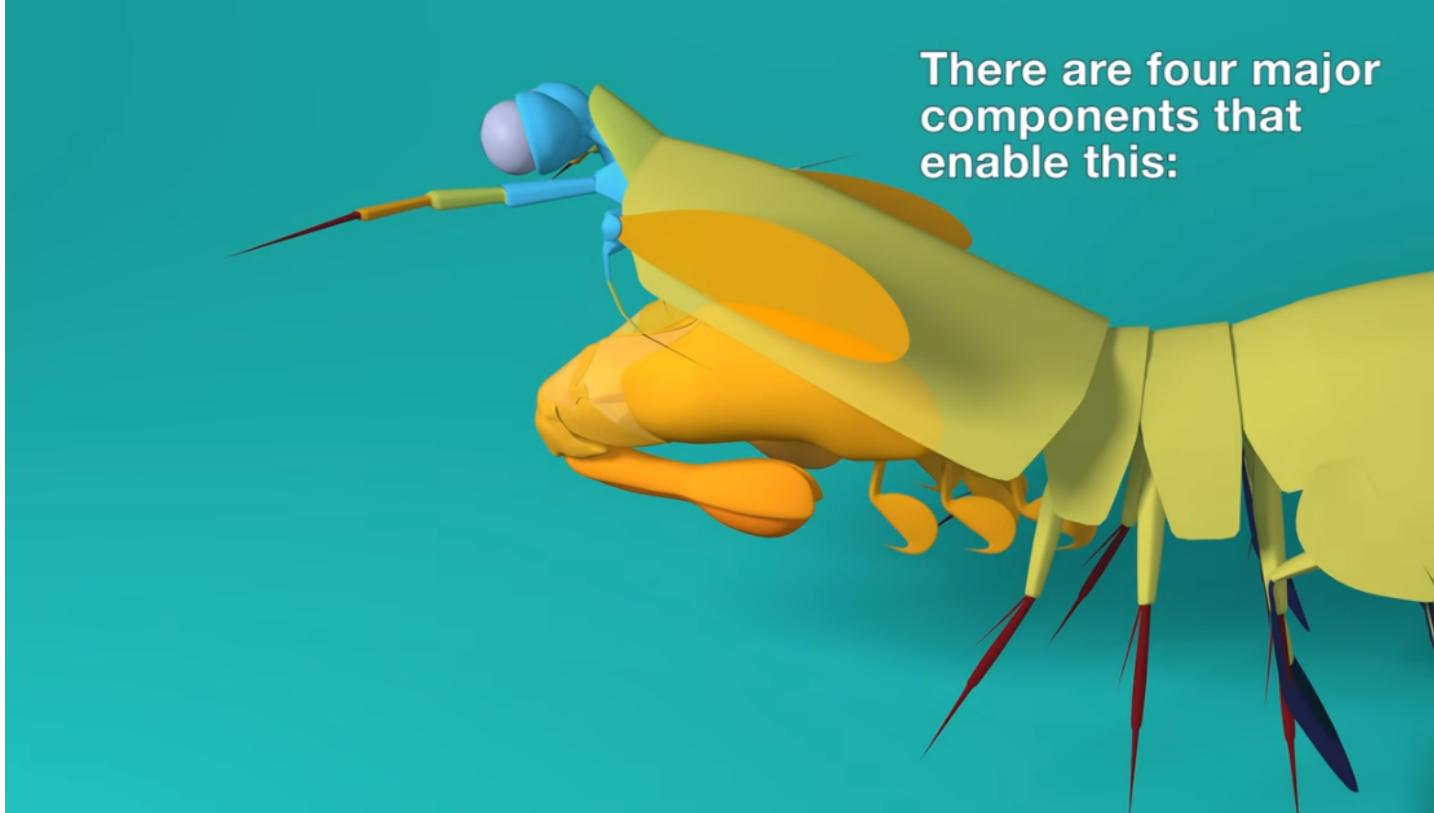
Camarón mantis



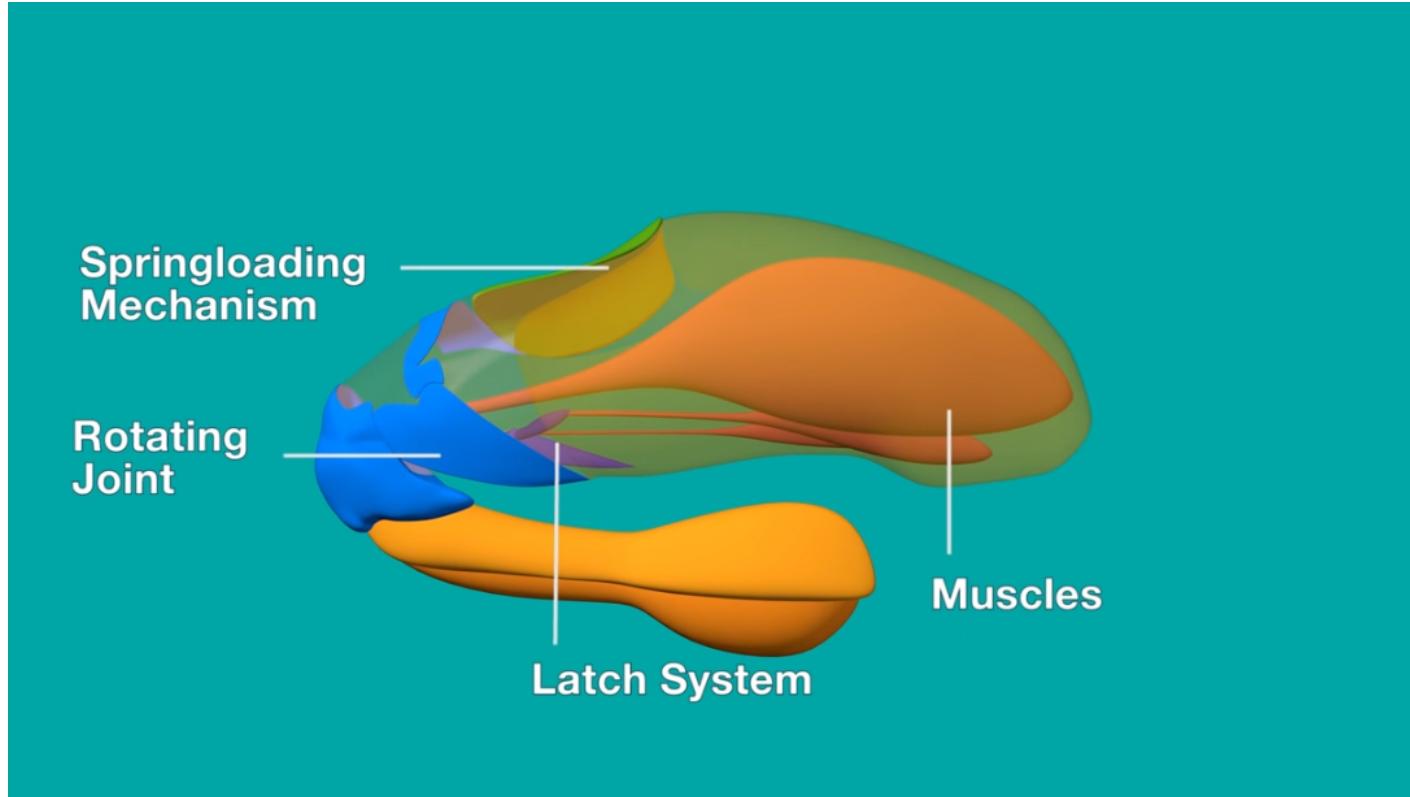
Camarón mantis

<https://www.youtube.com/watch?v=aabCOzFzMxU>

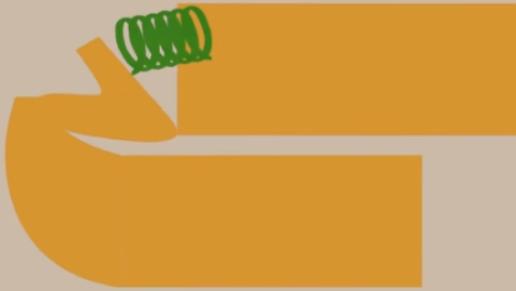
Camarón mantis



Camarón mantis



Camarón mantis

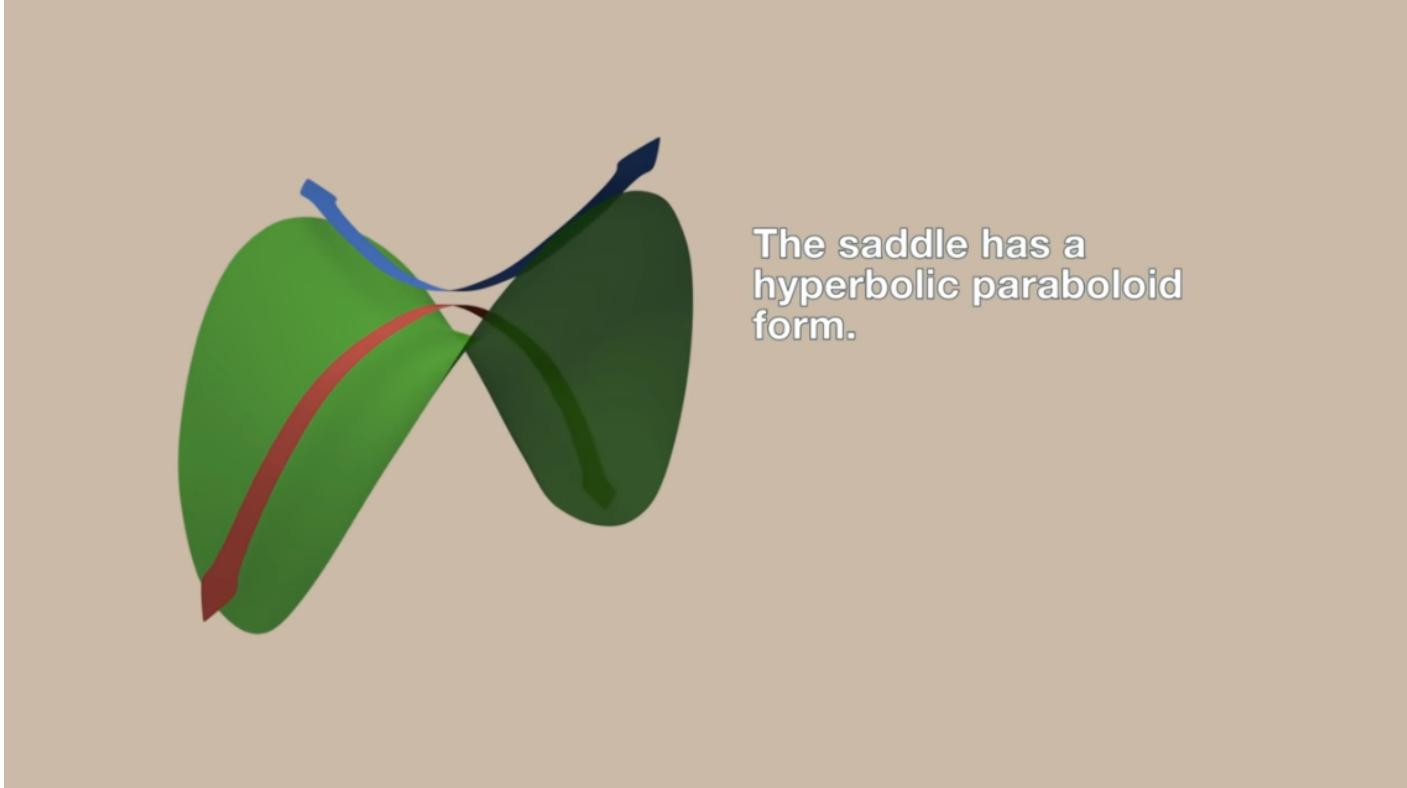


A spring is compressed,
storing energy and then
releasing it explosively.

Camarón mantis



Camarón mantis

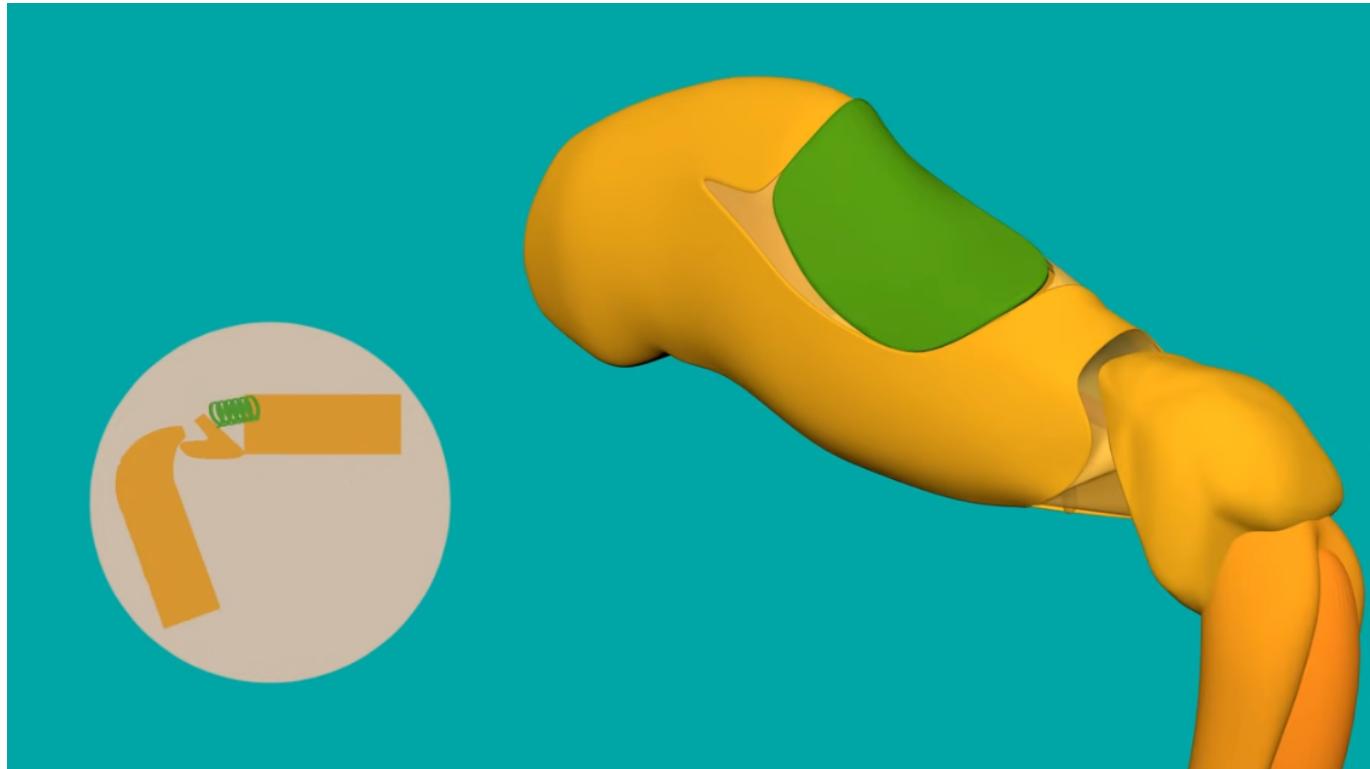


The saddle has a
hyperbolic paraboloid
form.

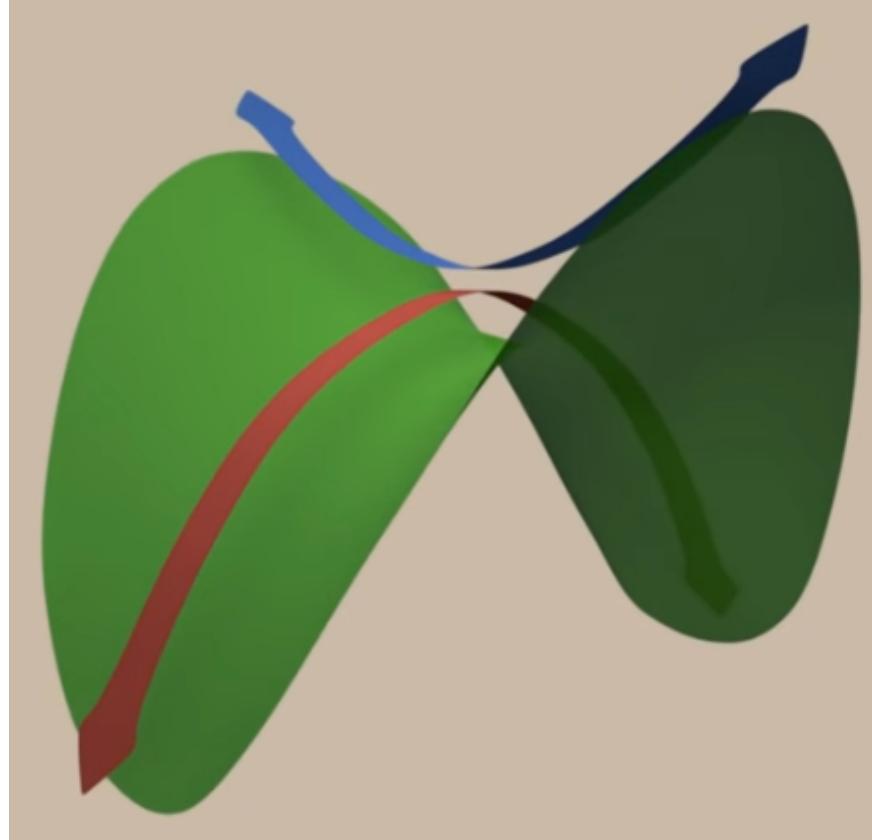
Camarón mantis

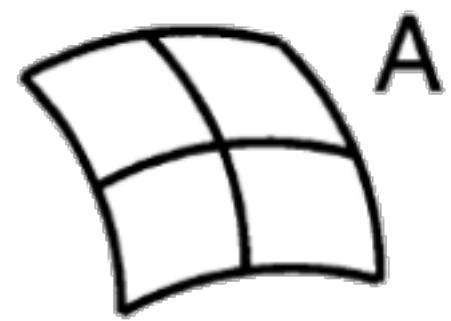


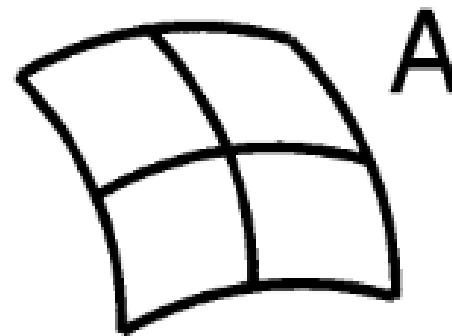
Camarón mantis

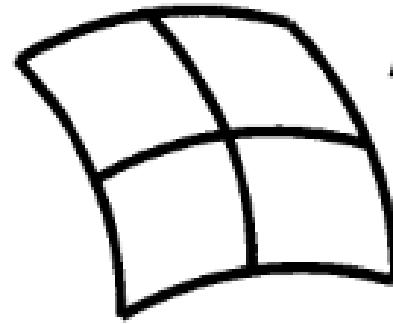


Camarón mantis



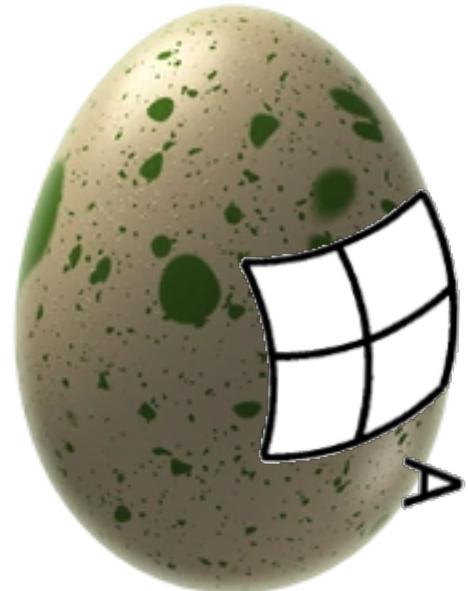


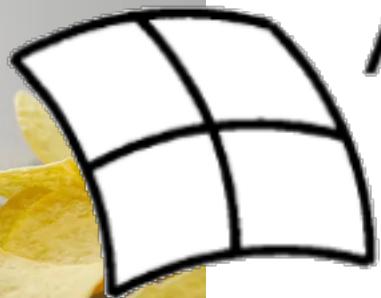




A

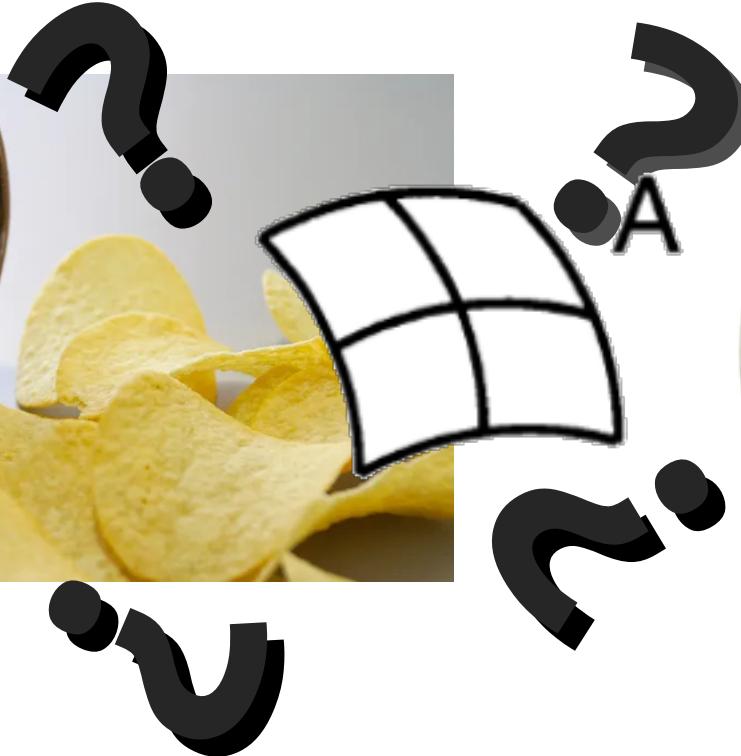






A





















¿Cómo decidimos cuál percepto es correcto?



Annu. Rev. Psychol. 2004. 55:10.1–10.32
doi: 10.1146/annurev.psych.55.090902.142005
Copyright © 2004 by Annual Reviews. All rights reserved
First published online as a Review in Advance on October 7, 2003

OBJECT PERCEPTION AS BAYESIAN INFERENCE

Daniel Kersten

*Department of Psychology, University of Minnesota, Minneapolis, Minnesota 55455;
email: kersten@umn.edu*

Pascal Mamassian

Department of Psychology, University of Glasgow, Glasgow G12 8QB, Scotland

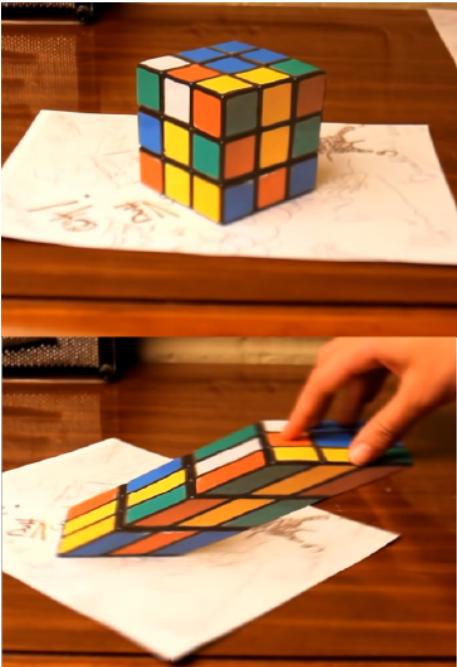
Alan Yuille

*Departments of Statistics and Psychology, University of California, Los Angeles,
Los Angeles, California 90095-1554*

Key Words shape, material, depth, vision, neural, psychophysics, fMRI,
computer vision

■ Abstract We perceive the shapes and material properties of objects quickly and reliably despite the complexity and objective ambiguities of natural images. Typical images are highly complex because they consist of many objects embedded in background clutter. Moreover, the image features of an object are extremely variable and ambiguous owing to the effects of projection, occlusion, background clutter, and illumination. The

Inferencia Bayesiana



(a) Una fotografía elongada de un cubo puede verse como un cubo real desde una perspectiva particular.



(b) El mismo edificio visto en condiciones y perspectivas distintas produce imágenes distintas en la retina.

Inferencia Bayesiana



“Helmholtz proposed that the visual system resolves ambiguity through built-in knowledge of the scene and how retinal images are formed and uses this knowledge to automatically and unconsciously infer the properties of objects.”

Inferencia Bayesiana

Estima la probabilidad de una cierta hipótesis con base en la evidencia actual y el conocimiento previo.

Inferencia Bayesiana

$$P(H|E)$$

$P(H|E)$ = Probabilidad de la hipótesis dado un evento

Inferencia Bayesiana

$$P(H|E) = \frac{P(E|H)}{\text{_____}}$$

$P(H|E)$ = *Probabilidad de la hipótesis dado un evento*
 $P(E|H)$ = *Probabilidad de un evento dada la hipótesis*

Inferencia Bayesiana

$$P(H|E) = \frac{P(E|H) \cdot P(H)}{\text{ }}$$

- $P(H|E)$ = Probabilidad de la hipótesis dado un evento
 $P(E|H)$ = Probabilidad de un evento dada la hipótesis
 $P(H)$ = Probabilidad a priori de la hipótesis

Inferencia Bayesiana

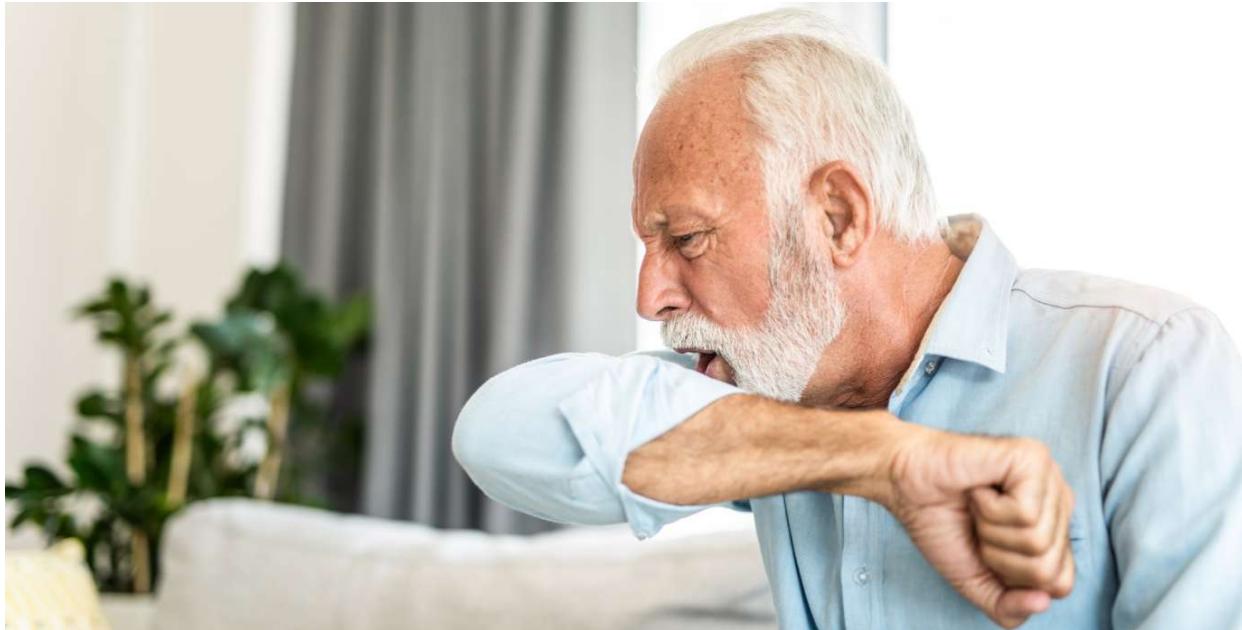
$$P(H|E) = \frac{P(E|H) \cdot P(H)}{P(E)}$$

- $P(H|E)$ = Probabilidad de la hipótesis dado un evento
 $P(E|H)$ = Probabilidad de un evento dada la hipótesis
 $P(H)$ = Probabilidad a priori de la hipótesis
 $P(E)$ = Probabilidad total del evento

EJEMPLO



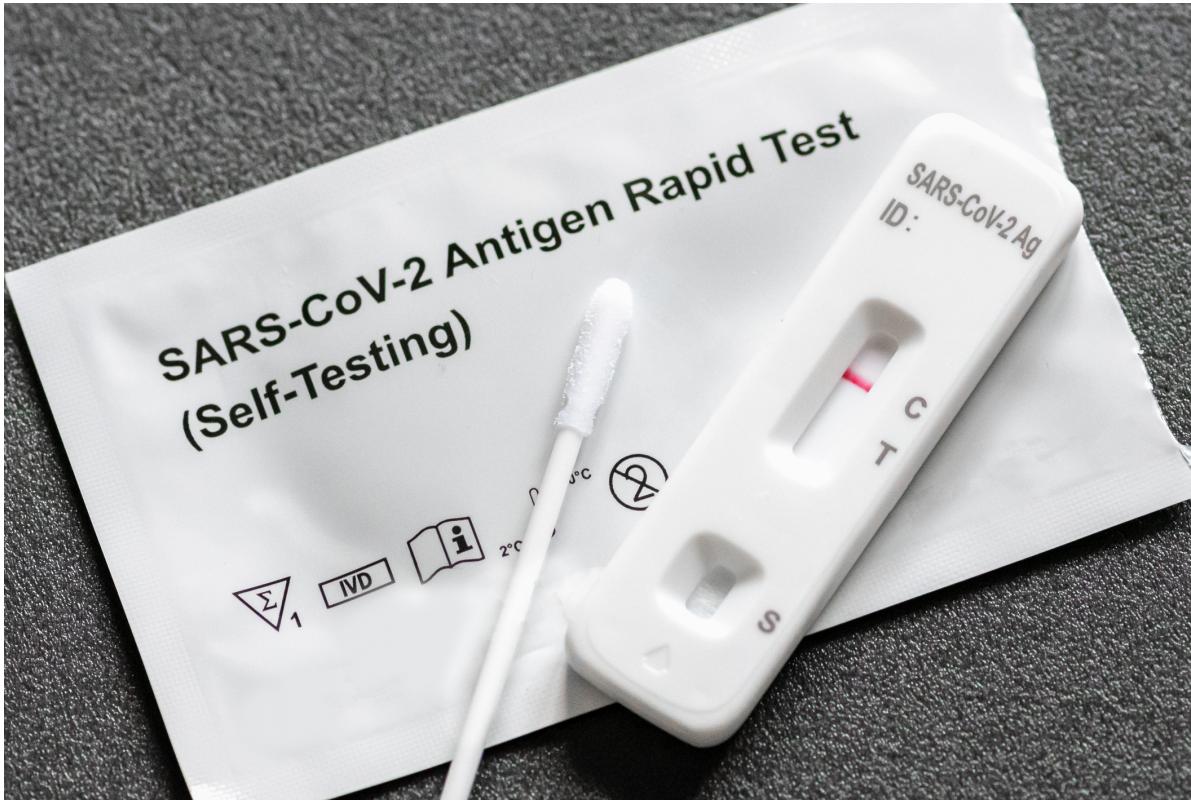
Inferencia Bayesiana





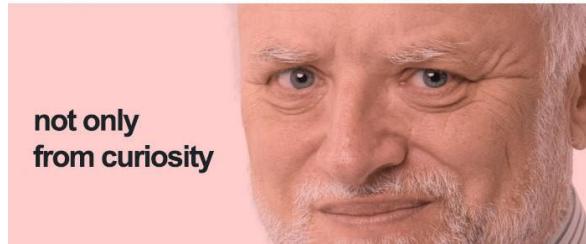
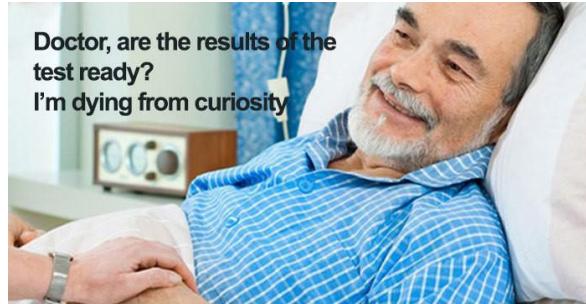
Hola, Dios, soy yo de nuevo

Inferencia Bayesiana



Precisión = 99%
Incidencia de 1 en 1000

Inferencia Bayesiana



Inferencia Bayesiana

$$P(H|E) = \frac{P(E|H) \cdot P(H)}{P(E)}$$

- $P(H|E)$ = Probabilidad de la hipótesis dado un evento
 $P(E|H)$ = Probabilidad de un evento dada la hipótesis
 $P(H)$ = Probabilidad a priori de la hipótesis
 $P(E)$ = Probabilidad total del evento

Inferencia Bayesiana

$$P(\text{enfermo}|\text{prueba}^+) = \frac{P(\text{prueba}^+|\text{enfermo}) \cdot P(\text{enfermo})}{\underbrace{P(\text{prueba}^+)}_{[P(\text{enfermo}) \times P(\text{prueba}^+|\text{enfermo})] + [P(\text{enfermo}^-) \times P(\text{prueba}^+|\text{enfermo}^-)]}}$$

Inferencia Bayesiana

$$P(\text{enfermo}|\text{prueba}^+) = \frac{P(\text{prueba}^+|\text{enfermo}) \cdot P(\text{enfermo})}{\underbrace{P(\text{prueba}^+)}_{[P(\text{enfermo}) \times P(\text{prueba}^+|\text{enfermo})] + [P(\text{enfermo}^-) \times P(\text{prueba}^+|\text{enfermo}^-)]}}$$

$$P(\text{enfermo}|\text{prueba}^+) = \frac{.99 \times .001}{.001 \times .99 + .999 \times .01}$$

Inferencia Bayesiana

$$P(\text{enfermo}|\text{prueba}^+) = \frac{P(\text{prueba}^+|\text{enfermo}) \cdot P(\text{enfermo})}{\underbrace{P(\text{prueba}^+)}_{[P(\text{enfermo}) \times P(\text{prueba}^+|\text{enfermo})] + [P(\text{enfermo}^-) \times P(\text{prueba}^+|\text{enfermo}^-)]}}$$

$$P(\text{enfermo}|\text{prueba}^+) = \frac{.99 \times .001}{.001 \times .99 + .999 \times .01}$$

$$P(\text{enfermo}|\text{prueba}^+) = \frac{.00099}{.00099 + .00999}$$

Inferencia Bayesiana

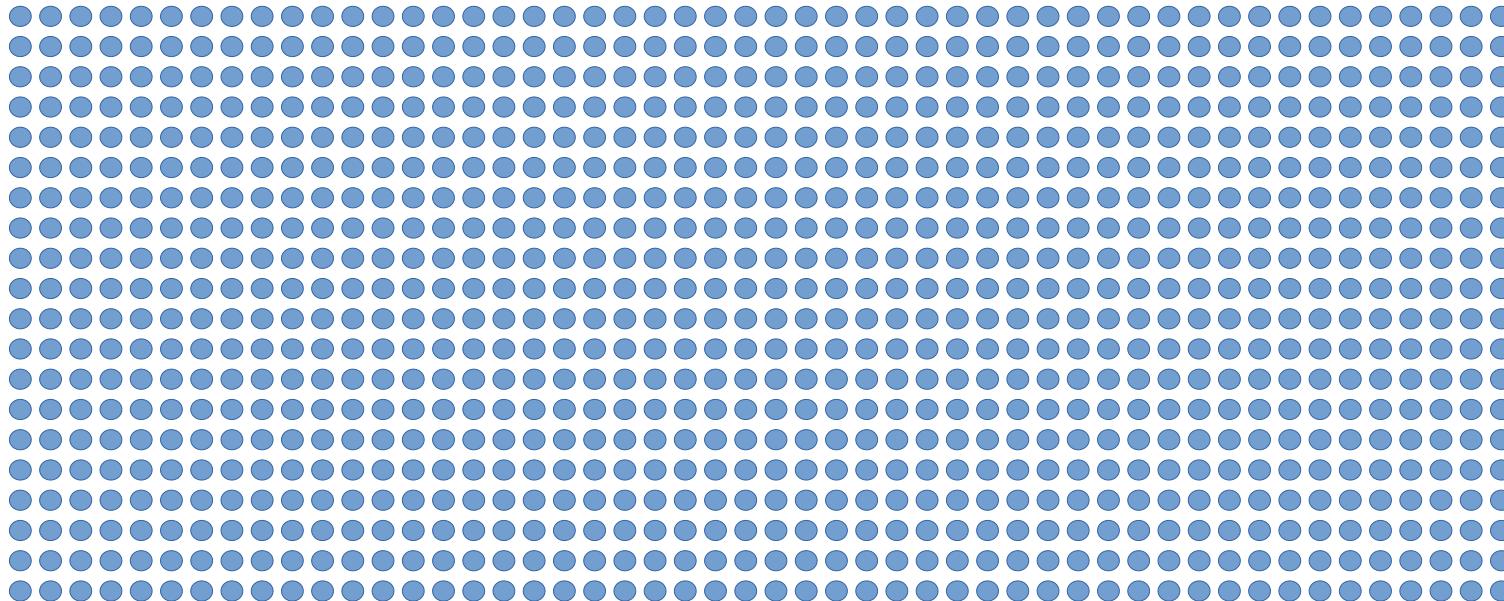
$$P(\text{enfermo}|\text{prueba}^+) = \frac{P(\text{prueba}^+|\text{enfermo}) \cdot P(\text{enfermo})}{\underbrace{P(\text{prueba}^+)}_{[P(\text{enfermo}) \times P(\text{prueba}^+|\text{enfermo})] + [P(\text{enfermo}^-) \times P(\text{prueba}^+|\text{enfermo}^-)]}}$$

$$P(\text{enfermo}|\text{prueba}^+) = \frac{.99 \times .001}{.001 \times .99 + .999 \times .01}$$

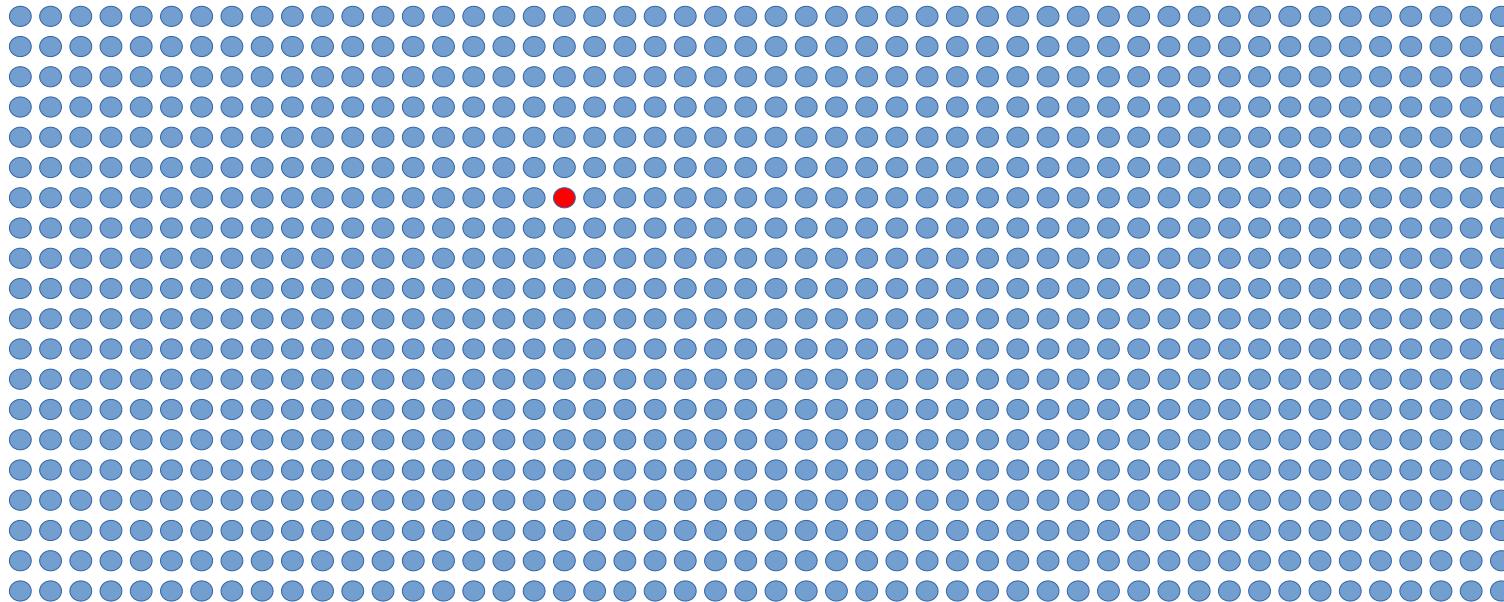
$$P(\text{enfermo}|\text{prueba}^+) = \frac{.00099}{.00099 + .00999}$$

$$P(\text{enfermo}|\text{prueba}^+) = .09$$

Inferencia Bayesiana

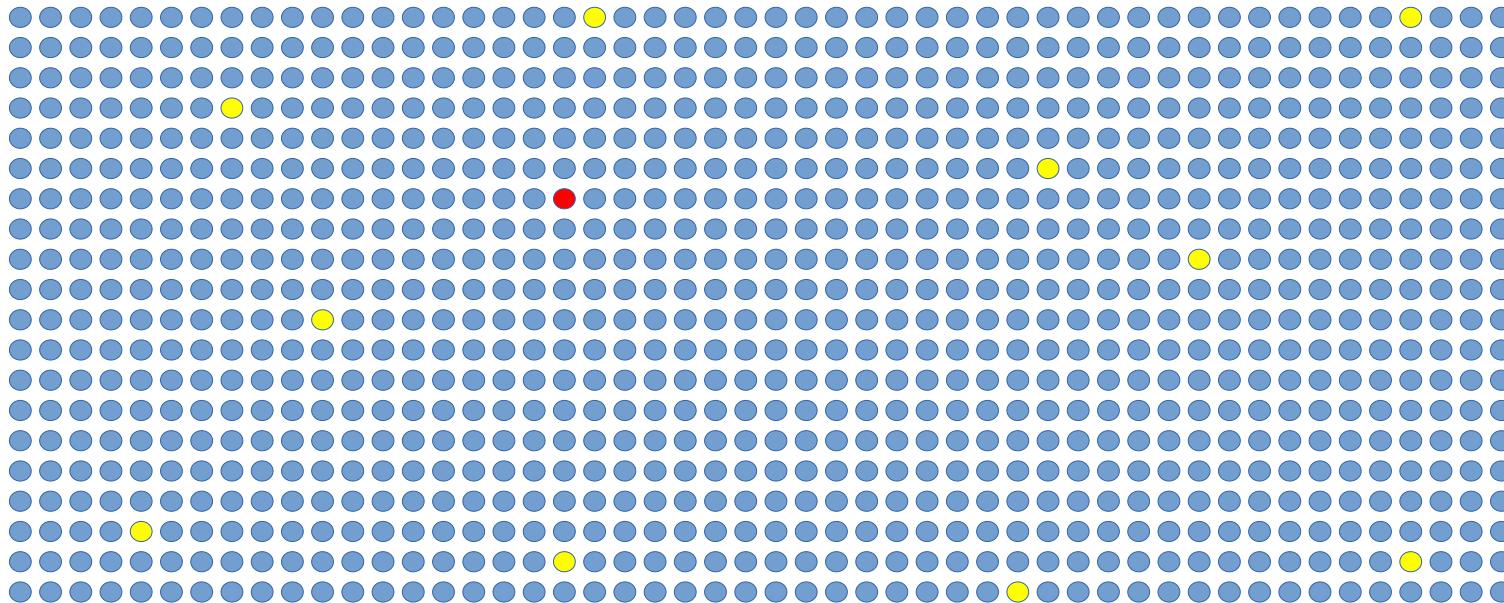


Inferencia Bayesiana



$$p(\text{enfermo}) = \frac{1}{1000} = .001$$

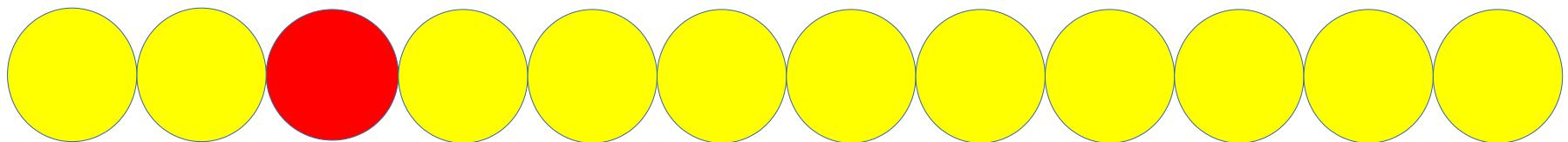
Inferencia Bayesiana



$$p(\text{enfermo}) = \frac{1}{1000} = .001$$

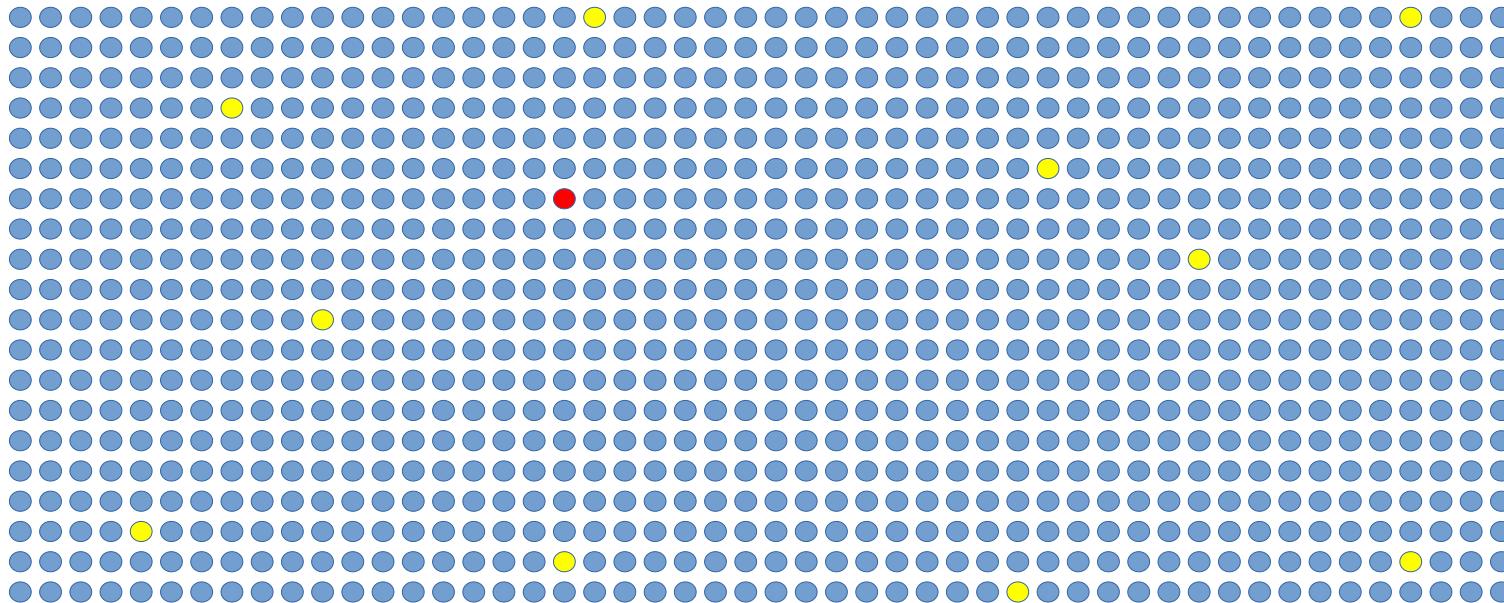
Precisión = .99

Inferencia Bayesiana



$$p(\text{enfermo}) = \frac{1}{11} = 0.09$$

Inferencia Bayesiana



$$P(\text{enfermo} | \text{prueba}^+) = .09$$

Inferencia Bayesiana

$$P(\text{enfermo}|\text{prueba}^+) = \frac{P(\text{prueba}^+|\text{enfermo}) \cdot P(\text{enfermo})}{\underbrace{P(\text{prueba}^+)}_{[P(\text{enfermo}) \times P(\text{prueba}^+|\text{enfermo})] + [P(\text{enfermo}^-) \times P(\text{prueba}^+|\text{enfermo}^-)]}}$$

$$P(\text{enfermo}|\text{prueba}^{++}) = \frac{.99 \times .001}{.001 \times .99 + .999 \times .01}$$

Inferencia Bayesiana

$$P(\text{enfermo}|\text{prueba}^+) = \frac{P(\text{prueba}^+|\text{enfermo}) \cdot P(\text{enfermo})}{\underbrace{P(\text{prueba}^+)}_{[P(\text{enfermo}) \times P(\text{prueba}^+|\text{enfermo})] + [P(\text{enfermo}^-) \times P(\text{prueba}^+|\text{enfermo}^-)]}}$$

$$P(\text{enfermo}|\text{prueba}^{++}) = \frac{.99 \times .001}{.001 \times .99 + .999 \times .01}$$

$$P(\text{enfermo}|\text{prueba}^{++}) = \frac{.99 \times .09}{.09 \times .99 + .91 \times .01}$$

Inferencia Bayesiana

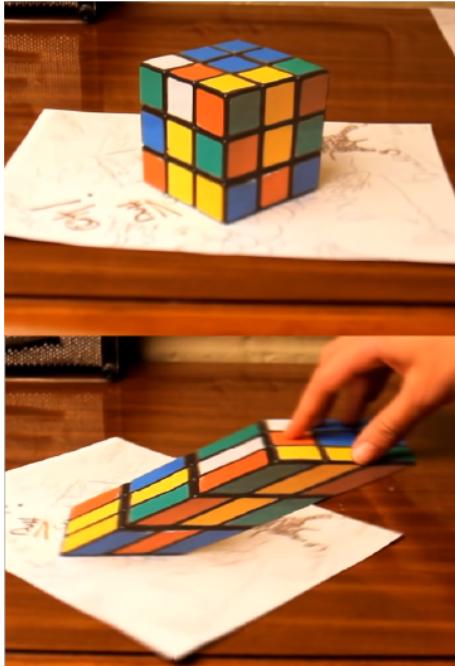
$$P(\text{enfermo}|\text{prueba}^+) = \frac{P(\text{prueba}^+|\text{enfermo}) \cdot P(\text{enfermo})}{\underbrace{P(\text{prueba}^+)}_{[P(\text{enfermo}) \times P(\text{prueba}^+|\text{enfermo})] + [P(\text{enfermo}^-) \times P(\text{prueba}^+|\text{enfermo}^-)]}}$$

$$P(\text{enfermo}|\text{prueba}^{++}) = \frac{.99 \times .001}{.001 \times .99 + .999 \times .01}$$

$$P(\text{enfermo}|\text{prueba}^{++}) = \frac{.99 \times .09}{.09 \times .99 + .91 \times .01}$$

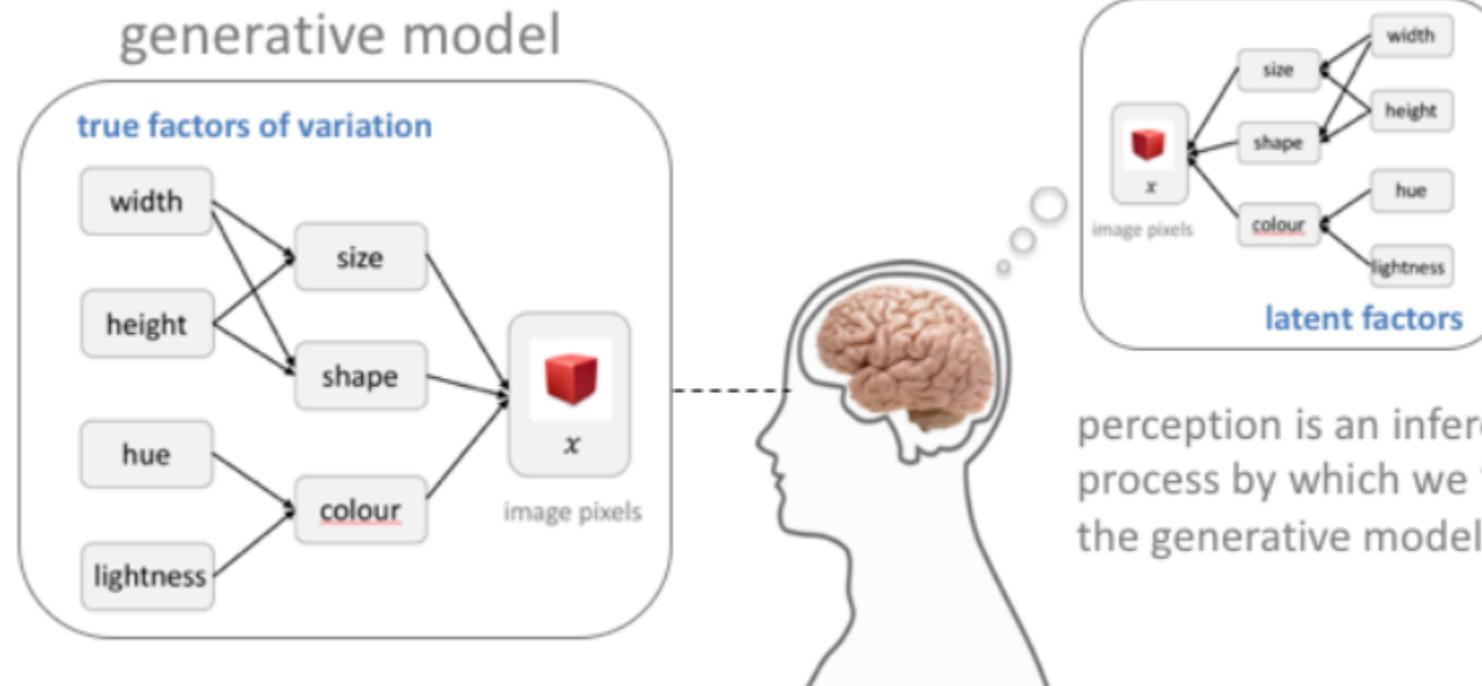
$$P(\text{enfermo}|\text{prueba}^{++}) = 0.907$$

Percepción visual

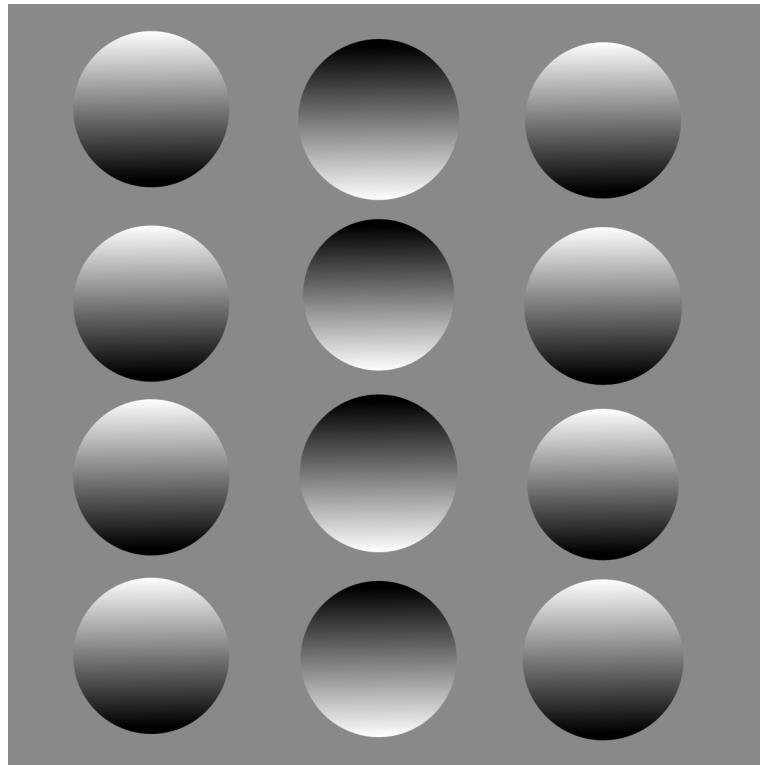


(a) Una fotografía elongada de un cubo puede verse como un cubo real desde una perspectiva particular.

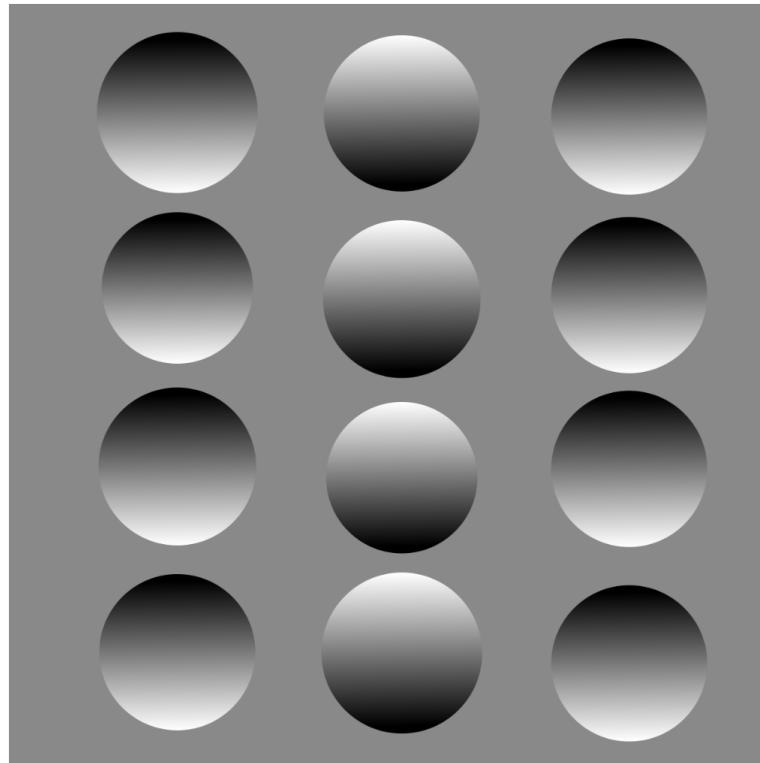
Percepción visual



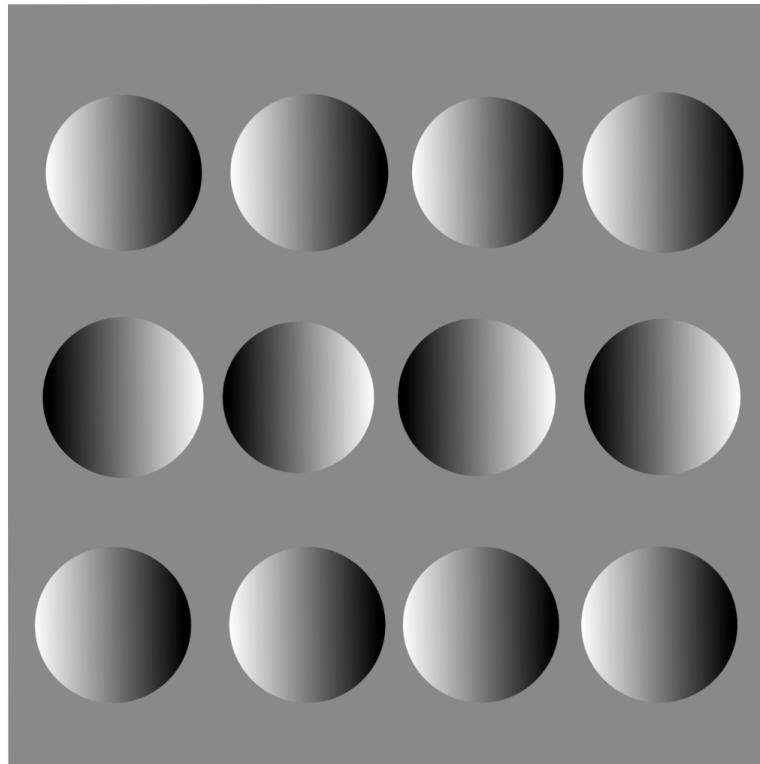
Percepción visual



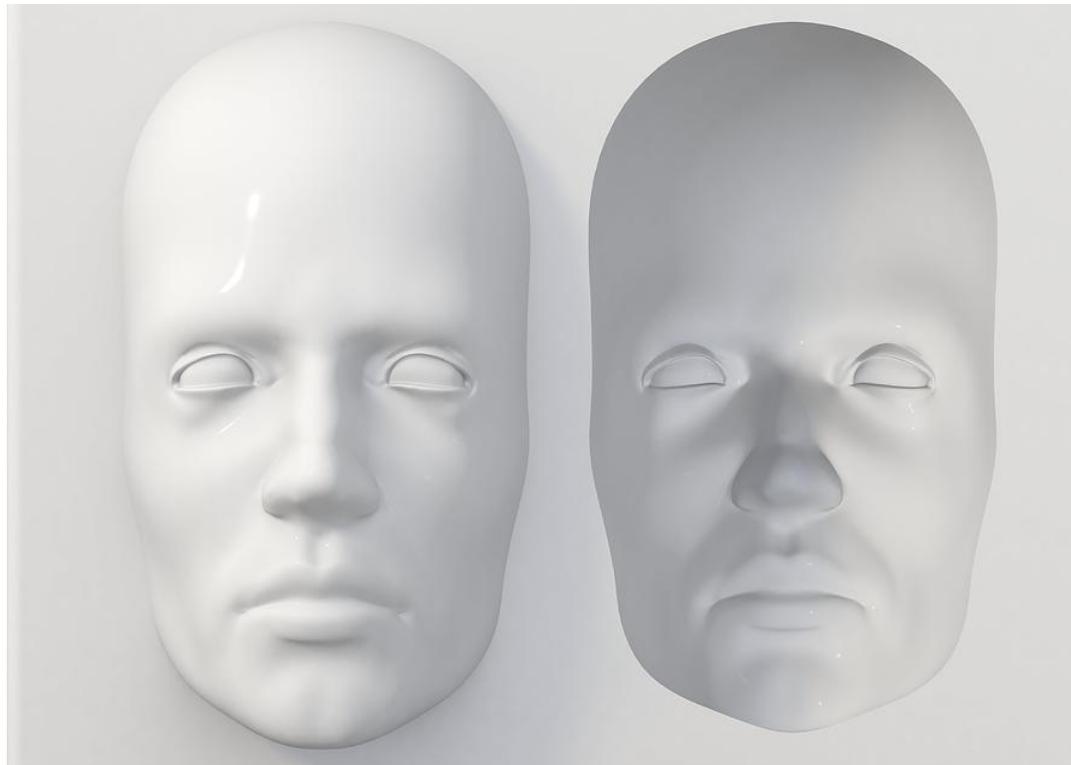
Percepción visual



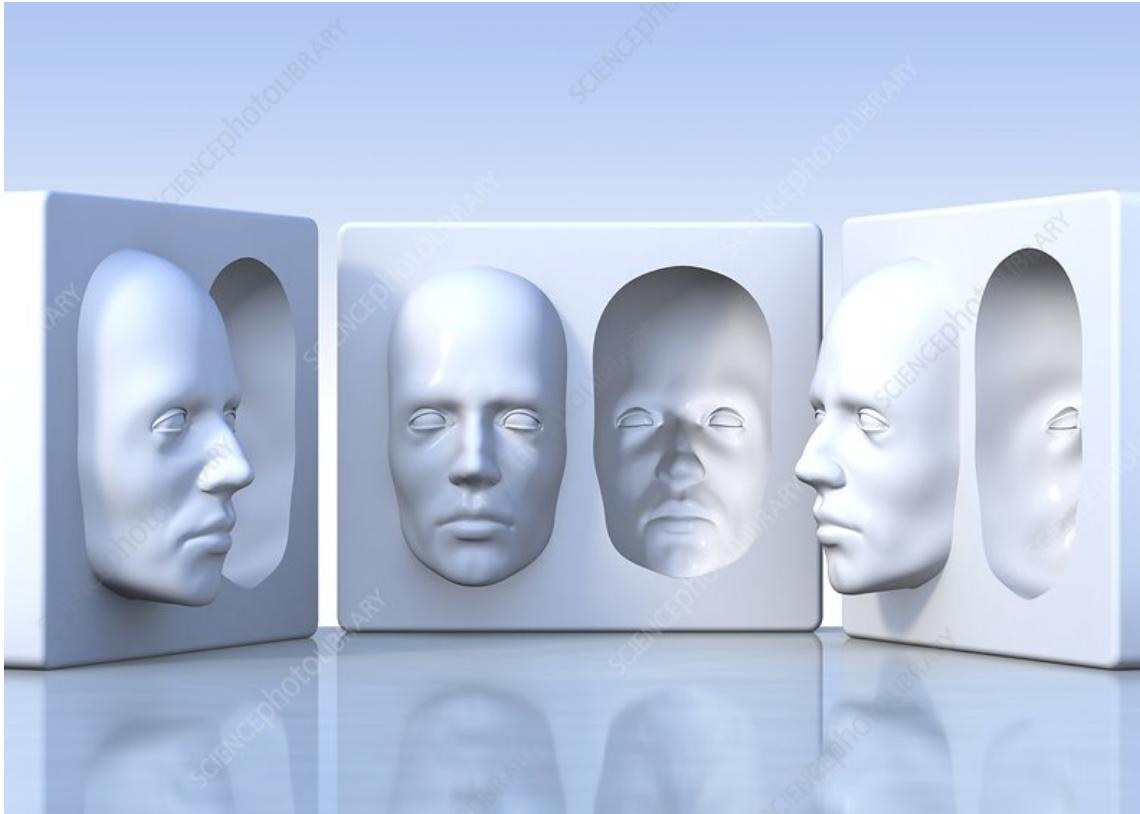
Percepción visual



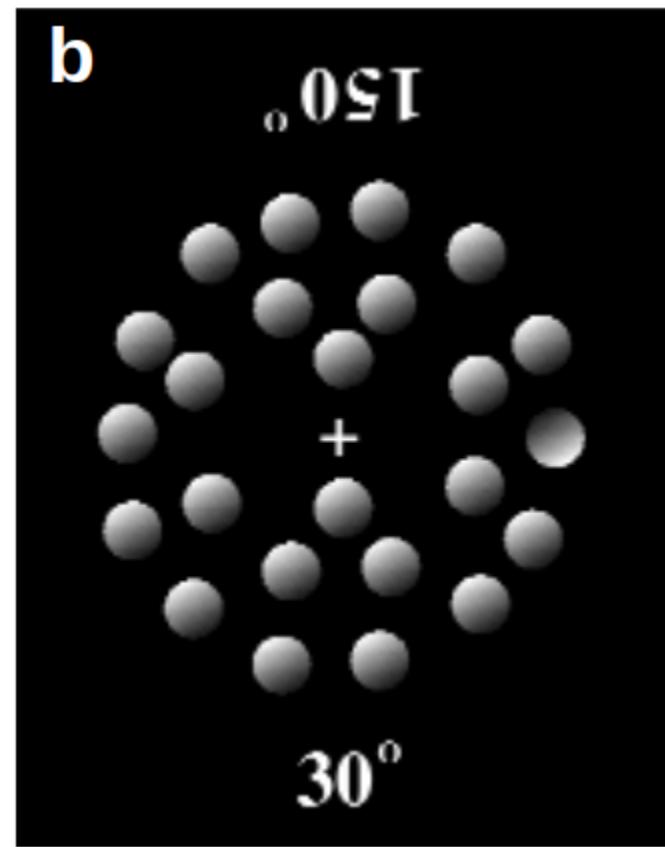
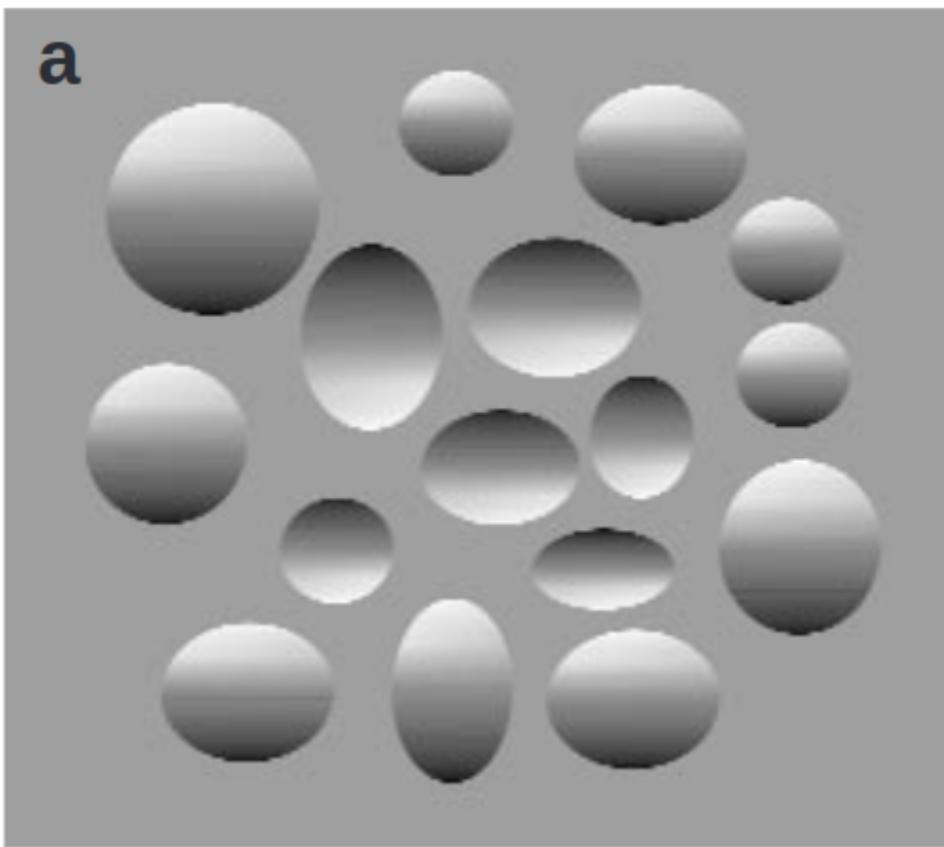
Percepción visual

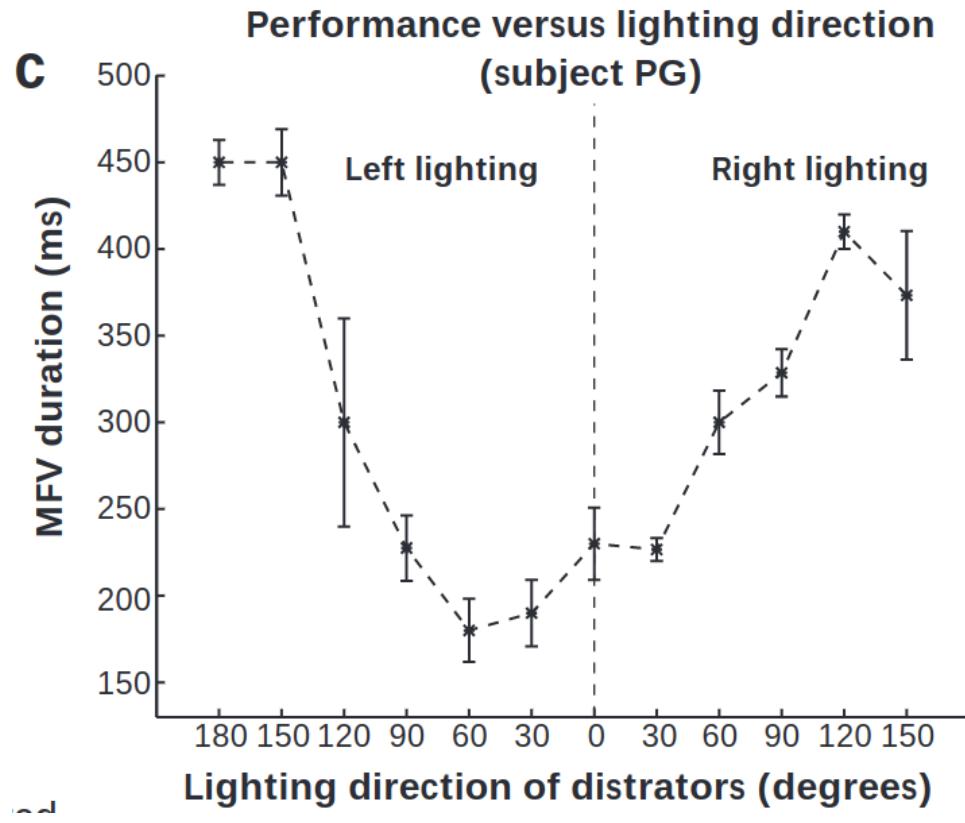


Percepción visual

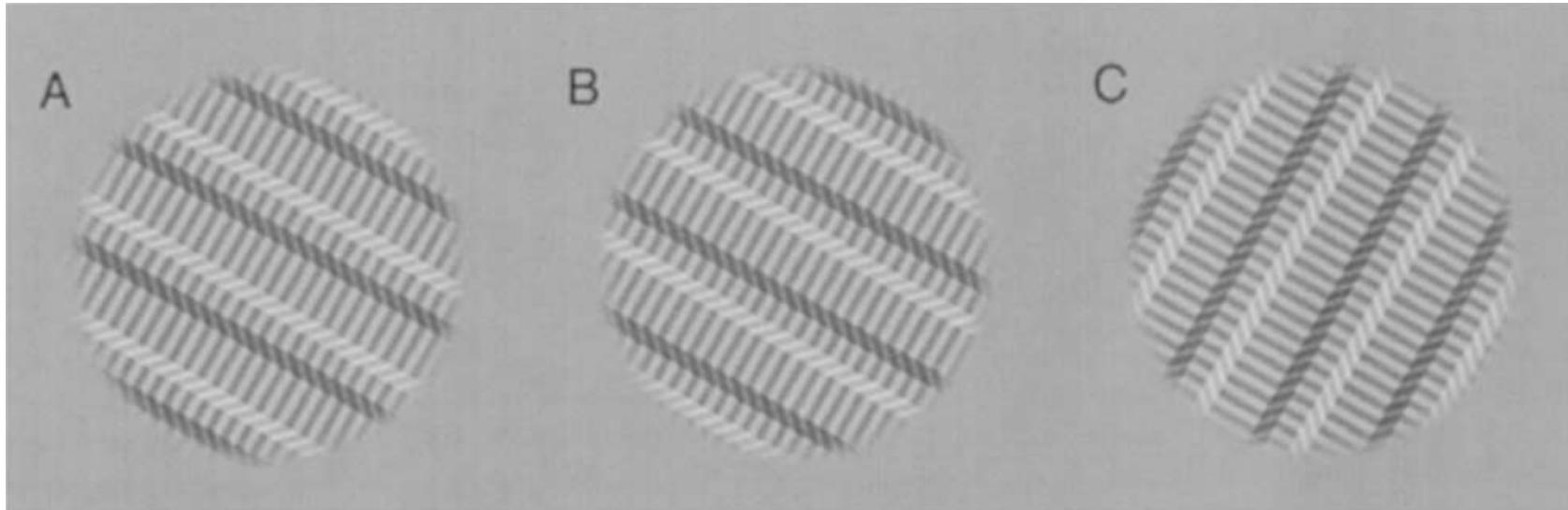


Percepción visual

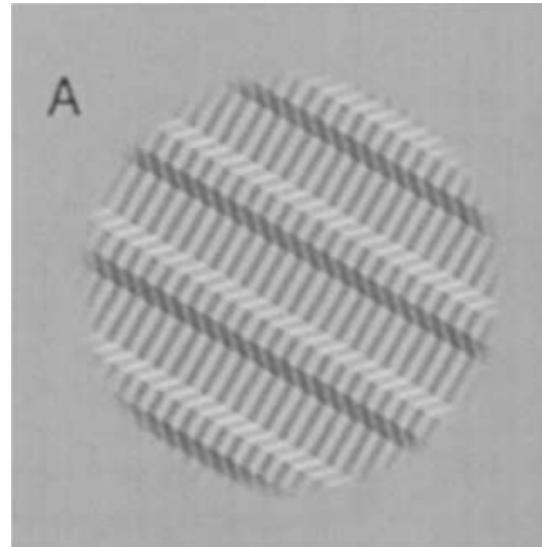




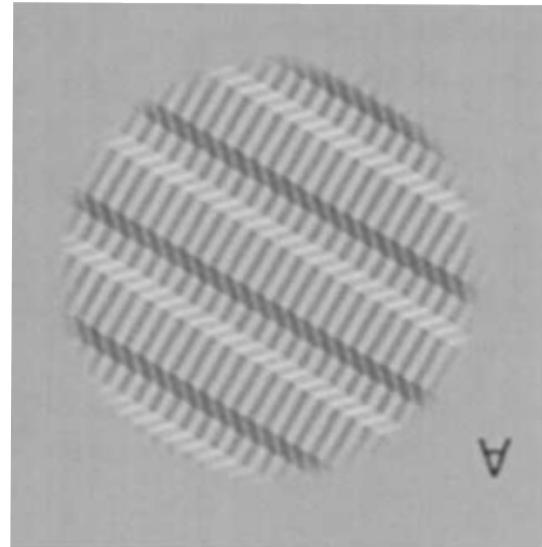
Percepción visual



Percepción visual



Percepción visual



Percepción visual

