

Physical Principles in Biology
Biology 3550
Fall 2016

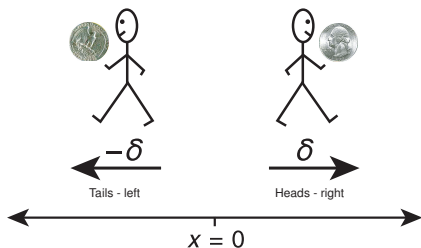
Lecture 10

Random Walks in Two Dimensions

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A Random Walk in One Dimension



- 1 Start at position $x = 0$.
- 2 Flip coin.
 - Heads, take step of length δ to the right.
 - Tails, take step of length δ to the left.
- 3 Repeat 2 another $(n - 1)$ times.
- 4 Final position is $x(n)$.

Symbols Used in the Treatment of the 1-dimensional Random Walk

For a single random walk

- n : number of steps in the random walk.
- $x(i)$: position along the x -axis at the end of step i ($1 \leq i \leq n$)
- δ : length of each step (a positive value)
- δ_i : change in x for step i , either $+\delta$ or $-\delta$
- $p(+\delta)$: probability of a step to the right (towards larger x)
- $p(-\delta)$: probability of a step to the left, equal to $1 - p(+\delta)$
- $\langle \delta_i \rangle$: mean value of δ_i , over many steps (exp. value $E(\delta_i)$).
- $\langle \delta_i^2 \rangle$: mean-square value of δ_i , over many steps (exp. value $E(\delta_i^2)$).

Symbols Used in the Treatment of the 1-dimensional Random Walk

For a large number of random walks

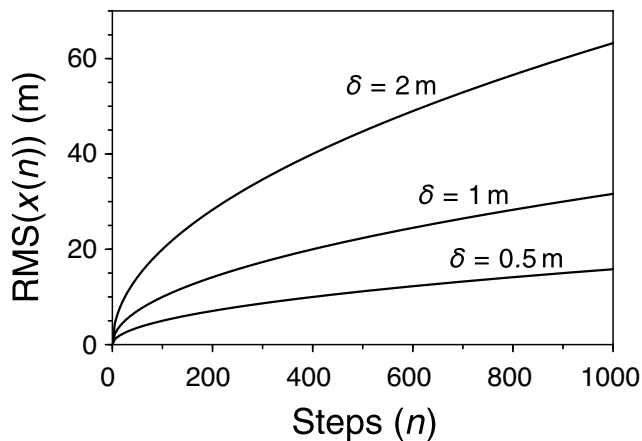
- N : total number of random walks
- $x(i)_j$: x -position of random walk j ($1 \leq j \leq N$), after i ($1 \leq i \leq n$) steps
- $\langle x(i) \rangle$: mean x -value after i steps, equal to $\left(\sum_{j=1}^N x(i)_j \right) / N$
- $\langle x(i)^2 \rangle$: mean-square x -value after i steps, equal to $\left(\sum_{j=1}^N x(i)_j^2 \right) / N$
- $\text{RMS}(x(i))$: root-mean-square x -value after i steps, equal to $\sqrt{\langle x(i)^2 \rangle}$

Major Results for the 1-dimensional Random Walk

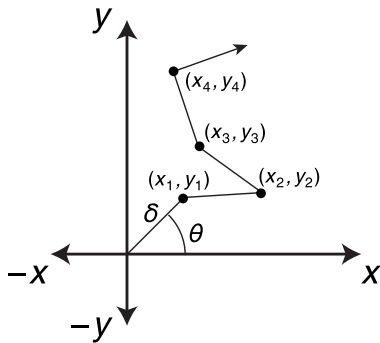
- Without constraints on $E(\delta_i)$
 - $\langle x(n) \rangle = nE(\delta_i)$
- If $E(\delta_i) = 0$ (unbiased choice of direction)
 - $\langle x(n) \rangle = 0$
 - $\langle x(n)^2 \rangle = nE(\delta_i^2) = n\delta^2$
 - $\text{RMS}(x(n)) = \sqrt{\langle x(n)^2 \rangle} = \sqrt{n}\delta$

The Root-mean-square Displacement for a One-dimensional Random Walk

$$\text{RMS}(x(n)) = \sqrt{\langle x(n)^2 \rangle} = \sqrt{n}\delta$$

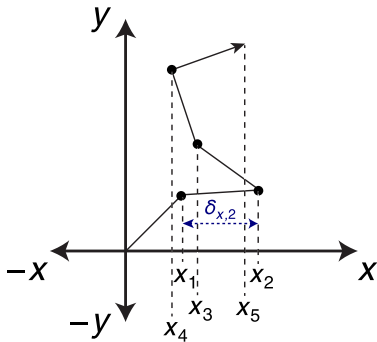


A Random Walk in Two Dimensions



- 1 Start at (x, y) coordinates $(0,0)$.
- 2 Choose a random direction, defined by the angle θ from the x -axis.
- 3 Move distance δ in the chosen direction.
- 4 Repeat 2 and 3 another $(n - 1)$ times.

A Random Walk in Two Dimensions

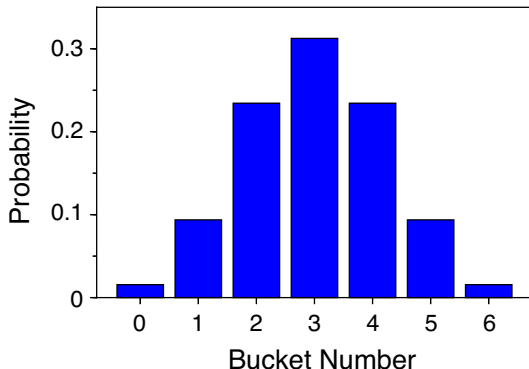


- x -coordinates represent a random walk along the x -axis.
- Can also describe a random walk along the y -axis (or any other direction).
- What are $\langle x_n \rangle$, $\langle x_n^2 \rangle$ and $\text{RMS}(x_n)$?
- The change in x with each step, δ_x is not discrete!

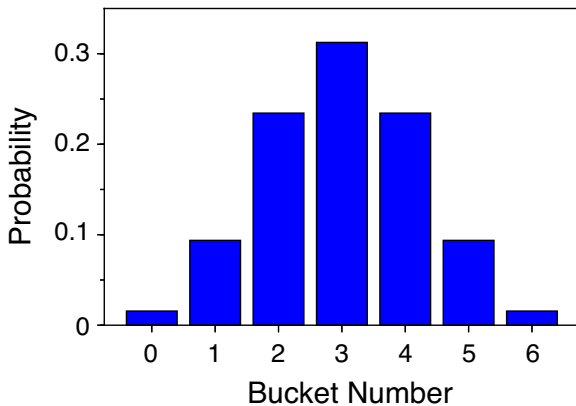
Discrete Probability Distribution Functions

- So far, we have dealt with events with a finite number of discrete outcomes and random variables with discrete values.
- The probability distribution functions can be viewed as tables or bar graphs

Bucket No.	Probability
0	$1/64$
1	$6/64$
2	$15/64$
3	$20/64$
4	$15/64$
5	$6/64$
6	$1/64$



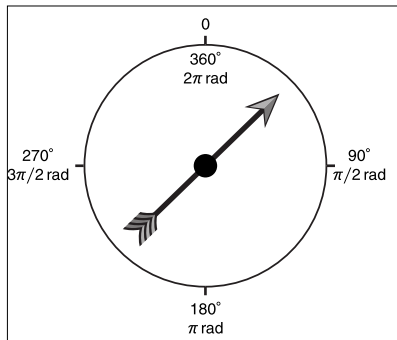
Graph For Discrete Probability Distribution Function



- Sum of probabilities = 1
- If the total width of bars is defined as 1, the total area of the bars is 1.

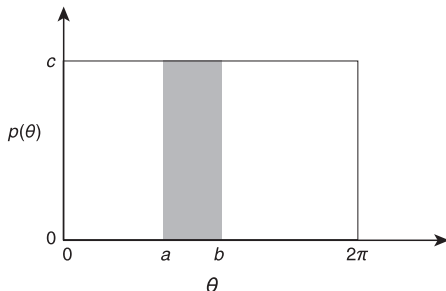
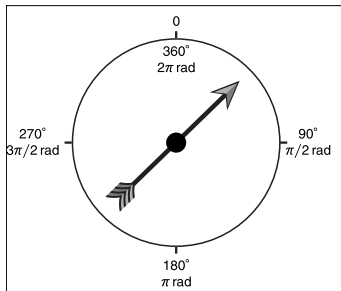
Introducing Continuous Probability Distribution Functions

- A spinner to choose directions for the 2-dimensional random walk



- We could divide up the circle into a finite number of sectors.
 - Two sectors: Like flipping a coin
 - Six sectors: Like throwing a die
 - Lots of other possibilities
- OR, we can treat the result as a continuous variable from 0 to 2π rad

A Continuous Probability Distribution Function for the Spinner

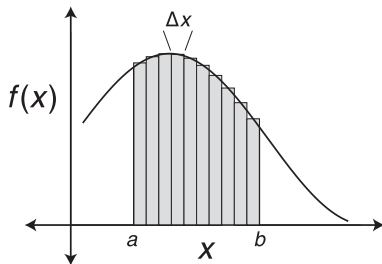


- θ is a continuous variable, with values from 0 to 2π .
- $p(\theta)$ is a function of θ , with a constant value, c , for all values of θ .
- Interpretation of $p(\theta)$: The integral

$$\int_a^b p(\theta) d\theta$$

is the probability that the spinner lands between the values a and b .

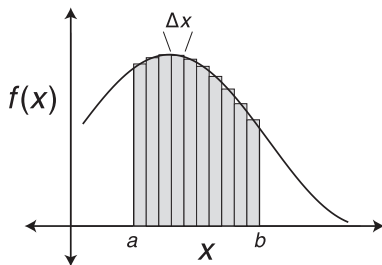
A Quick Refresher on Integrals (as “area under the curve”)



- To approximate the area between the x -axis and the function $f(x)$, between $x = a$ and $x = b$:
 - Divide up the range $a \leq x \leq b$ into n segments $\Delta x = (b - a)/n$ wide.
 - Draw n rectangles Δx wide and $f(x_i)$ high.
 - Sum the areas of the rectangles

$$\text{area} = \sum_{i=1}^n f(x_i) \Delta x$$

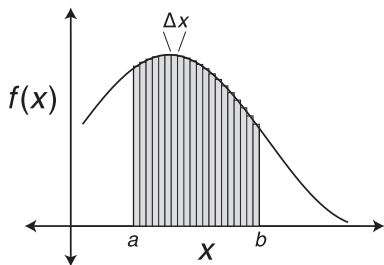
A Quick Refresher on Integrals (as “area under the curve”)



- Improve approximation by making Δx smaller (and n larger).
- If the function is “well behaved”, Δx can be made infinitesimally small.
- The definite integral, from a to b with respect to x , is defined as:

$$\int_a^b f(x) dx = \lim_{\Delta x \rightarrow 0} \sum_{i=1}^n f(x_i) \Delta x$$

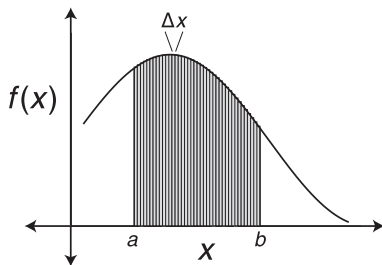
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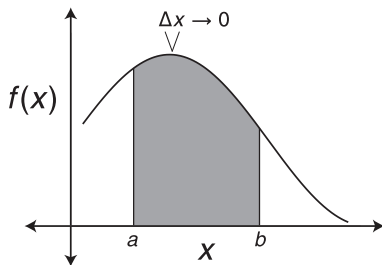
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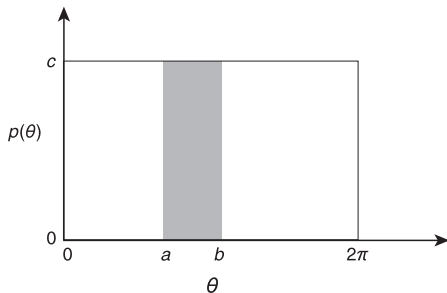
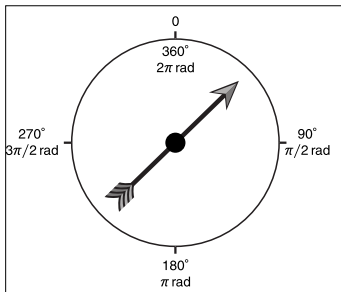


- Improve approximation by making Δx smaller (and n larger).
- If the function is “well behaved”, Δx can be made infinitesimally small.
- The definite integral, from a to b with respect to x , is defined as:

$$\int_a^b f(x) dx = \lim_{\Delta x \rightarrow 0} \sum_{i=1}^n f(x_i) \Delta x$$

Back to the Continuous Probability Distribution Function (PDF) for the Spinner

- $p(\theta)$ is a function of θ , with a constant value, c , for all values of θ .

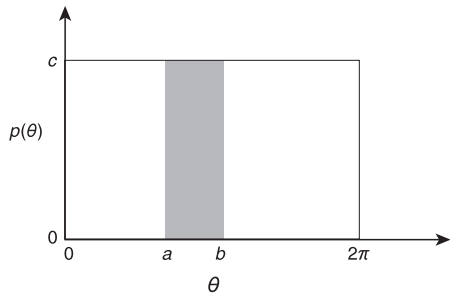


- The integral

$$\int_a^b p(\theta) d\theta$$

is the probability that the spinner lands between the values a and b .

An Important Constraint on a Continuous PDF



- To be properly “normalized”, the integral

$$\int_0^{2\pi} p(\theta) d\theta$$

must equal 1.

- Equivalent to the requirement for a discrete PDF that the sum of all probabilities be equal to 1.

Normalizing the Spinner PDF

- $p(\theta) = c$ (a constant)

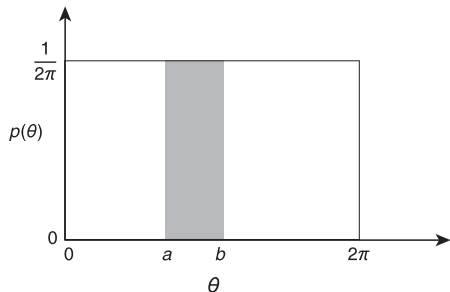
$$\int_0^{2\pi} p(\theta) d\theta = \int_0^{2\pi} c d\theta = 1$$

$$\int_0^{2\pi} c d\theta = c\theta \Big|_0^{2\pi} = (2\pi - 0)c = 1$$

- Solving for c

$$c = \frac{1}{2\pi}$$

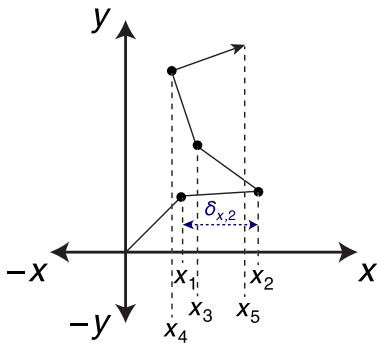
The Normalized Spinner PDF



- The function: $p(\theta) = 1/(2\pi)$
- The probability that θ lies between a and b :

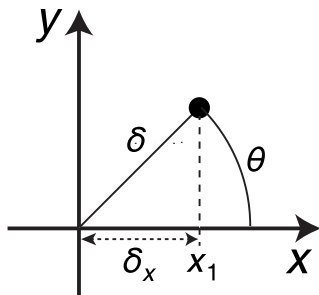
$$\int_a^b p(\theta) d\theta = \int_a^b \frac{1}{2\pi} d\theta = \frac{\theta}{2\pi} \Big|_a^b = \frac{b-a}{2\pi}$$

Back to the 2-dimensional Random Walk



- The first step is to describe the 1-dimensional walk along the x -axis.
- What are $\langle x_n \rangle$, $\langle x_n^2 \rangle$ and $\text{RMS}(x_n)$?
- We need to know the expected value for each change in x , δx .
- The change in x with each step, δx is not discrete.

A Continuous PDF for δ_x

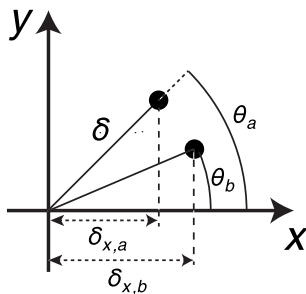


- δ_x is a continuous random variable.
- δ_x is related to θ according to:

$$\cos(\theta) = \frac{\delta_x}{\delta}$$

$$\delta_x = \delta \cos(\theta)$$

A Change of Coordinates



- The probability that δ_x lies between $\delta_{x,a}$ and $\delta_{x,b}$ is the same as the probability that θ lies between θ_b and θ_a .

$$\int_{\delta_{x,a}}^{\delta_{x,b}} p(\delta_x) d\delta_x = \int_{\theta_b}^{\theta_a} p(\theta) d\theta = \int_{\theta_b}^{\theta_a} \frac{1}{2\pi} d\theta$$

The Expected Value of δ_x

- For a discrete random variable, x , with discrete PDF, $p(x)$, the expected value is:

$$E(x) = \sum_{i=1}^n xp(x)$$

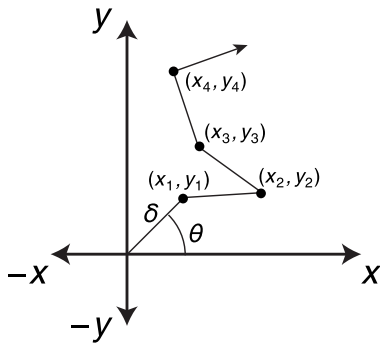
- For a continuous random variable, x , with range $x_1 \leq x \leq x_2$ and continuous PDF, $p(x)$, the expected value is:

$$E(x) = \int_{x_1}^{x_2} xp(x)dx$$

- For δ_x :

$$E(\delta_x) = \int_0^{2\pi} \delta \cos(\theta)p(\theta)d\theta = \int_0^{2\pi} \delta \cos(\theta)\frac{1}{2\pi}d\theta = \frac{\delta \sin(\theta)}{2\pi} \Big|_{\theta=0}^{\theta=2\pi} = 0$$

The Expected Value of δ_x



- For each step, δ_x is as likely to have the value $-a$ as the value $+a$.
- It makes sense that $E(\delta_x) = 0$.