

Recall: $\Gamma(\alpha) := \int_0^{\infty} x^{\alpha-1} e^{-x} dx, \alpha > 0.$

$\Gamma(n) = (n-1)!$ in particular $\Gamma(1) = 1.$

Gamma(α, λ) distribution density function:

$$f(x) = \begin{cases} \frac{\lambda^{\alpha} x^{\alpha-1} e^{-\lambda x}}{\Gamma(\alpha)} & \text{if } x > 0 \\ 0 & \text{otherwise} \end{cases}$$

Beta(a, b) distribution

Let $a, b > 0$. The Beta(a, b) distribution is a continuous distribution on the interval $[0, 1]$. First we define

$$B(a, b) := \frac{\Gamma(a+b)}{\Gamma(a)\Gamma(b)}. \text{ For example:}$$

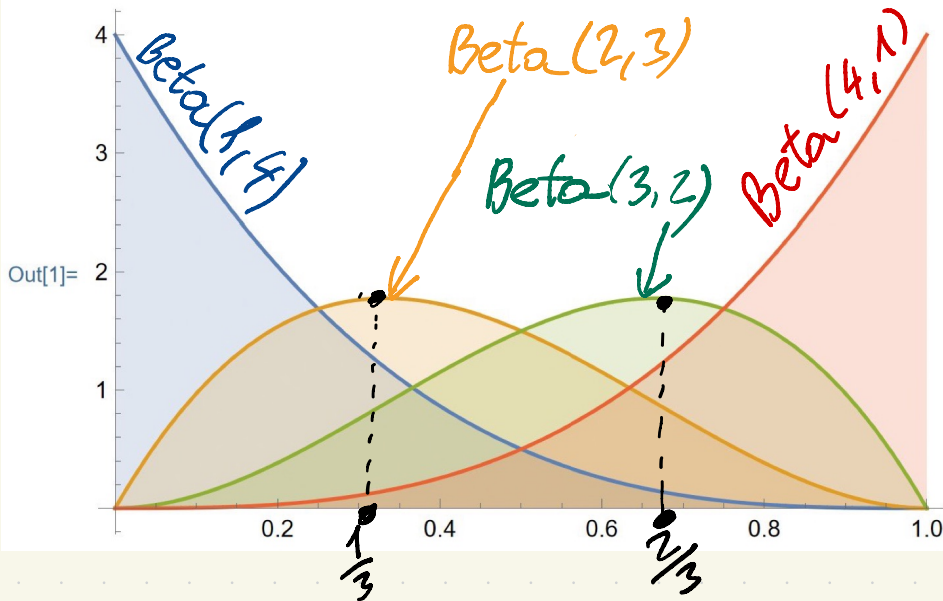
$$B(1, n) = \frac{\Gamma(1+n)}{\Gamma(1)\Gamma(n)} = \frac{n!}{1 \cdot (n-1)!} = n.$$

The density function of the Beta(a, b)

distribution: $f_{a,b}(x) = \frac{1}{B(a,b)} \cdot x^{a-1} \cdot (1-x)^{b-1}$ $x \in [0, 1]$

If $x \notin [0, 1]$ then $f_{a,b}(x) := 0$.

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In[1]:= Plot[Table[PDF[BetaDistribution[a, 5 - a], x], {a, {1, 2, 3, 4}}]
// Evaluate, {x, 0, 1}, Filling -> Axis]
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$$X \sim \text{Beta}(a, b); E[X] = \frac{\alpha}{\alpha + \beta}; \text{Var}(X) = \frac{ab}{(a+b)^2(1+a+b)}$$

The mode of Beta(a, b) Let $a > 1, b > 1$.

The mode is the most likely value (corresponding to the peak of the distribution function). This is $\frac{\alpha - 1}{\alpha + \beta - 2}$.

The median of Beta (a, b)

$$F_{a,b}(y) = \int_0^y x^{a-1} (1-x)^{b-1} dx.$$

Median = $F_{a,b}^{-1}\left(\frac{1}{2}\right)$ no closed formula

For $a=1, b>0$, Median = $1 - 2^{-\frac{1}{b}}$.

Let

X_1, \dots, X_n i.i.d. $X_i \sim \text{Uniform}(0, 1)$.

Let $X_{(j)}$ be the j -th smallest of

X_1, \dots, X_n . Then $X_{(1)} \leq X_{(2)} \leq \dots \leq X_{(n)}$.

Then

$X_{(j)} \sim \text{Beta}(j, n+1-j)$.

5. Let X_1, X_2, \dots i.i.d. random variables taking values in the integers such that $\mathbb{E}[X_i] = 0$ for all i . Let $S_0 := 0$ and $S_n := X_1 + \dots + X_n$.

(a) Let $G_n(x) := \sum_{j=0}^n \mathbb{1}_{\{S_j=x\}}$. That is $G_n(x)$ is the expected number of visits to x in the first n steps. ← $\mathbb{E}[\sum_{j=0}^n \mathbb{1}_{\{S_j=x\}}]$

Show that for all n and x , $G_n(0) \geq G_n(x)$. (Hint: consider the first j with $S_j = x$.)

(b) Note that the Law of Large Numbers implies that for each $\varepsilon > 0$ we have: $\lim_{n \rightarrow \infty} \mathbb{P}(|S_n| \leq n\varepsilon) = 1$.
Using this prove that for each $\varepsilon > 0$ we have $\lim_{n \rightarrow \infty} \frac{1}{n} \sum_{|x| \leq \varepsilon \cdot n, x \in \mathbb{Z}} G_n(x) = 1$

(c) Using (a) and (b), show that for each $M < \infty$ there is an n such that $G_n(0) \geq M$.

(d) Now prove that S_n is a recurrent Markov chain.

Solution (a) $G_n(x) = \mathbb{E}[\sum_{j=0}^n \mathbb{1}_{\{S_j=x\}}]$. $G_n(0) \geq G_n(x)$.

Clearly, for $\forall x \in \mathbb{Z}$: $\textcircled{*} G_k(x) \leq G_l(x)$ if $k \leq l$

Let $S \subseteq \mathbb{Z}$ be the state space. If $x \in \mathbb{Z} \setminus S$ then $G_n(x) = 0 \forall n \in \{0, 1, 2, \dots\}$. So, we may assume that $x \in S$. Let $T_x := \min\{k \geq 1 : S_k = x\}$.

$$G_n(x) = \mathbb{E}[\sum_{j=0}^n \mathbb{1}_{\{S_j=x\}}] = \sum_{j=1}^n \mathbb{E}[\sum_{k=j}^n \mathbb{1}_{\{S_k=x\}} \mid T_x = j] \cdot \mathbb{P}(T_x = j)$$

$$= \sum_{j=1}^{\infty} G_{n-j}(0) \cdot \mathbb{P}(T_x = j) \stackrel{\textcircled{*}}{\leq} G_n(0). \text{ This is so since}$$

$\sum_{j=1}^{\infty} \mathbb{P}(T_x = j) \leq 1$ & $G_{n-j}(0) \leq G_n(0)$. This completes the proof of (a).

Law of Large Numbers

(b) By L.L.N. for every $\varepsilon > 0$ we have

$$\textcircled{**} \lim_{n \rightarrow \infty} \mathbb{P}(|S_n| \leq n\varepsilon) = 1. \text{ Namely, } \mathbb{E}[X_n] = 0 \forall n.$$

Hence by the L.L.N. we have

$$\lim_{n \rightarrow \infty} \frac{\overbrace{X_1 + \dots + X_n}^{S_n}}{n} = E[X_1] = 0 \text{ a.s. Hence, for } \varepsilon > 0:$$

$$\lim_{n \rightarrow \infty} P\left(\left|\frac{S_n}{n}\right| > \varepsilon\right) = 0. \text{ This verifies that } (**)\text{ holds.}$$

Now we prove:

$$\lim_{n \rightarrow \infty} \frac{1}{n} \sum_{|X| \leq \varepsilon n} G_n(X) = 1. \quad (***)$$

$$\begin{aligned} \lim_{n \rightarrow \infty} \frac{1}{n} \sum_{|X| \leq \varepsilon n} \underbrace{E\left[\sum_{j=0}^n \mathbb{1}_{\{S_j = X\}}\right]}_{G_n(X)} &= \\ &= \lim_{n \rightarrow \infty} \frac{1}{n} \sum_{j=0}^n \sum_{|X| \leq \varepsilon n} \underbrace{E[\mathbb{1}_{\{S_j = X\}}]}_{P(S_j = X)} \\ &= \lim_{n \rightarrow \infty} \frac{1}{n} \sum_{j=0}^n \underbrace{P(|S_j| \leq \varepsilon n)}_{P(S_j = X)} \end{aligned}$$

$$= \lim_{n \rightarrow \infty} \frac{1}{n} \sum_{j=0}^n P(|S_j| \leq \varepsilon n). \text{ Now we prove that}$$

$$\lim_{n \rightarrow \infty} \frac{1}{n} \sum_{j=0}^n P(|S_j| \leq \varepsilon n) = 1. \quad (****) \quad \text{Let } a_{j,n} = P(|S_j| \leq \varepsilon n).$$

We know that $\lim_{n \rightarrow \infty} a_{n,n} = 1$ & $a_{j,n} \leq 1$.
(from (**))

$$\lim_{n \rightarrow \infty} \frac{1}{n} \sum_{j=0}^n a_{j,j} = 1$$

$$a_{j,j} = P(|S_j| \leq \varepsilon j) \leq P(|S_j| \leq \varepsilon n) = a_{j,n} \text{ if } j \leq n$$

$$1 = \lim_{n \rightarrow \infty} \frac{1}{n} \sum_{j=0}^n a_{j,j} \leq \lim_{n \rightarrow \infty} \frac{1}{n} \sum_{j=0}^n a_{j,n} \leq 1.$$

$$\text{Hence } \lim_{n \rightarrow \infty} \frac{1}{n} \sum_{j=0}^n a_{j,n} = \lim_{n \rightarrow \infty} \frac{1}{n} \sum_{j=0}^n P(|S_j| \leq \varepsilon n) = 1.$$

This proves that $(***)$ holds. Hence $(**)$ also holds. This completes the proof of part (b).

(c) We want to prove that $\forall M > 0, \exists n$ s.t. $G_n(0) \geq M$.

Recall that we have proved that $\forall \varepsilon > 0$:

$$\lim_{n \rightarrow \infty} \frac{1}{n} \sum_{|x| \leq \varepsilon n} G_n(x) = 1. \quad (***)$$

$$1 = \lim_{n \rightarrow \infty} \frac{1}{n} \sum_{|x| \leq \varepsilon n} G_n(x) \leq \lim_{n \rightarrow \infty} \frac{1}{n} \cdot 2\varepsilon n G_n(0) = 2\varepsilon \lim_{n \rightarrow \infty} G_n(0)$$

That is $\forall \varepsilon, \exists n_0$ s.t. for all $n \geq n_0$, $G_{n_0}(0) \geq \frac{1}{2\varepsilon}$.

We choose $\varepsilon > 0$ s.t. $\frac{1}{2\varepsilon} > M$. Then we have proved the assertion of part (c).

(d) It is obvious that S_n is irreducible.

Hence it is enough to prove that 0 is recurrent. We have seen in part (c) that the expected number of visits to 0 in the first n steps tends to infinity as $n \rightarrow \infty$.

Reflecting Random Walk

This is a Discrete time MC.

$\{X_n\}_{n=0}^{\infty}$ takes values from $\mathbb{N} = \{0, 1, 2, \dots\}$.

$$\begin{cases} p(i, i+1) = p & \text{for } i \geq 0 \\ p(i, i-1) = 1-p & \text{for } i \geq 1 \\ p(0, 0) = 1-p \end{cases}$$

This is a birth & death chain. So, the stationary distribution (if exists)

can be found by the detailed balance equation:

$$p\pi(i) = (1-p)\pi(i+1) \quad \text{for } i \geq 0.$$

This yields $\pi(i) = C \cdot \left(\frac{p}{1+p}\right)^i$ for $C = \pi(0)$.

(a) $p < \frac{1}{2}$ Then $\frac{p}{1+p} < 1$, $\sum_{k=0}^{\infty} \left(\frac{p}{1+p}\right)^k = \frac{1}{1 - \frac{p}{1+p}}$

$$\pi(i) = \frac{1-2p}{1-p} \left(\frac{p}{1+p}\right)^i$$

the first return time to 0

Then $P(X_n = j) \rightarrow \pi(j)$; $E_0[T_0] = \frac{1}{\pi(0)} = \frac{1-p}{1-2p}$

This is so since the chain is aperiodic.

(b) $p > \frac{1}{2}$ The series above is divergent no stationary distribution

① $p = \frac{1}{2}$ The chain is null recurrent.
See Durrett's book p. 55.

$p < \frac{1}{2}$ positive recurrent

$p = 0$ null recurrent

$p > \frac{1}{2}$ transient.

Theorem (see Durrett's book p. 55 Theorem 1.23)

For an irreducible chain the following are equivalent:

- (i) There is a positive recurrent state
- (ii) The stationary distribution exists
- (iii) All states are positive recurrent.