Sums

- If 0 < a < 1 then  $\sum_{k=j}^{\infty} a^k = \frac{a^j}{1-a}$ .
- $e^x = 1 + x + \frac{x^2}{2!} + \frac{x^3}{3!} + \dots = \sum_{k=0}^{\infty} \frac{x^k}{k!}.$ •  $(a+b)^n = \sum_{k=0}^n \binom{n}{k} a^k b^{n-k}.$

**Expected Value**: If the discrete random variable Z takes values 0, 1, ..., then  $E[Z] = \sum_{k=1}^{\infty} Pr\{Z \ge k\}.$ 

**Vocabulary** A state j is said to be **accessible** from i if  $P_{ij}^{(n)} > 0$  for some n. Two states that are accessible to one another are said to **communicate**, and we write  $i \leftrightarrow j$ . All the states that communicate with one another are grouped together in an **equivalence class**. When the state space has only one equivalence class, it is said to be **irreducible**. The **period** of a state i, written d(i), is the greatest common divisor of all the integers  $n \geq 1$  such that  $P_{ii}^{(n)} > 0$ . If  $P_{ii} > 0$ , then d(i) = 1. All states in an equivalence class have the same period.

For any state i,  $f_{ii}^{(n)} = Pr\{X_n = i | X_0 = i, X_1 \neq i, X_2 \neq i, \dots, X_{n-1} \neq i\}$  is the probability that, starting in state i, the first return to i is at the *n*th transition. We define  $f_{ii}^{(0)} = 0$ .

 $f_{ii} = \sum_{n=1}^{\infty} f_{ii}^{(n)}$  is the probability of *ever* returning to state *i*. If  $f_{ii} = 1$ , the state *i* is said to be **recurrent**. If  $f_{ii} < 1$ , then *i* is said to be **transient**. Let *i* be a recurrent state; the *return time* is  $R_i = \min\{n \ge 1 : X_n = i\}$ , with  $Pr\{R_i = n | X_0 = i\} = f_{ii}^{(n)}$ . If  $E[R_i|X_0 = i] = \sum_{n=1}^{\infty} n f_{ii}^{(n)} < \infty$ , the state *i* is called **positive recurrent**. Otherwise, *i* is called **null recurrent**.

**Theorem 3.0**  $P_{ii}^{(n)} = \sum_{k=0}^{n} f_{ii}^{(k)} P_{ii}^{(n-k)}$ . This is Eq. (3.2) in our text.

**Theorem 3.1** State *i* is transient if and only if  $\sum_{n=1}^{\infty} P_{ii}^{(n)} < \infty$ .

**Corollary 3.1** If  $i \leftrightarrow j$ , then states *i* and *j* are either both transient or both recurrent. **Theorem 4.1** For an irreducible, recurrent, aperiodic, stationary Markov chain,  $\lim_{n\to\infty} P_{jj}^{(n)} = \lim_{n\to\infty} P_{ij}^{(n)} = \frac{1}{m_i}$ , where  $m_i = \sum_{n=1}^{\infty} n f_{ii}^{(n)} = E[R_i|X_0 = i]$ . **Theorem 4.2** For an irreducible, positive recurrent, aperiodic, stationary Markov chain,

**Theorem 4.2** For an irreducible, positive recurrent, aperiodic, stationary Markov chain,  $\pi_j = \lim_{n \to \infty} P_{ij}^{(n)}$  is uniquely determined by  $\boldsymbol{\pi} = \boldsymbol{\pi} \mathbf{P}$  and  $\sum_{j=0}^{\infty} \pi_j = 1$ .

**Binomial**:  $X \sim B(n,p)$  means  $Pr\{X = k\} = \binom{n}{k} p^k (1-p)^{n-k}$  for k = 0, ..., n. Note: E[X] = np, V[X] = np(1-p).

**Poisson**:  $X \sim P(\mu)$  means  $Pr\{X = k\} = \frac{e^{-\mu}\mu^k}{k!}$  for  $k = 0, 1, \dots$  Note:  $E[X] = V[X] = \mu$ 

**Geometric**:  $X \sim \text{Geometric}(a)$  means  $Pr\{X = n\} = a(1-a)^{n-1}$  for  $n = 1, \dots$ . Note: E[X] = 1/a

**Exponential**:  $X \sim \exp(\lambda)$  means  $f_X(x) = \lambda e^{-\lambda x} \mathbf{1}\{x \ge 0\}$ . Note  $E[X] = 1/\lambda$ , and  $F_X(x) = (1 - e^{-\lambda x}) \mathbf{1}\{x \ge 0\}$ .

**Gamma**:  $X \sim G(n, \lambda)$  means  $f_X(x) = \frac{\lambda^n}{(n-1)!} e^{-\lambda x} x^{n-1} \mathbf{1}\{x \ge 0\}$ . Note  $E[X] = \frac{n}{\lambda}$ . The sum of *n* independent exponential random variables is Gamma.