Section 3.8: Sums, Products, and Quotients of Independent Random Variables

Gene Quinn

Theorem: (3.8.1 - discrete case) Suppose *X* and *Y* are independent discrete random variables with pdfs $p_X(x)$ and $p_Y(y)$, respectively.

Define a random variable W to be the sum of X and Y.

$$p_W(w) = \sum_{\text{all } x} p_X(x) \cdot p_Y(w - x)$$

Example: Suppose two dice are thrown, and the random variables X and Y represent the numbers showing on the faces of the first and second die, respectively, so: $p_X(x) = p_Y(y) = 1/6$.

We are primarily interested in the sum of the two faces, so define a new random variable W = X + Y.

Example: Suppose two dice are thrown, and the random variables X and Y represent the numbers showing on the faces of the first and second die, respectively, so: $p_X(x) = p_Y(y) = 1/6$.

We are primarily interested in the sum of the two faces, so define a new random variable W = X + Y.

Theorem 3.8.3 says that the pdf of W will be

$$p_W(w) = \sum_{\text{all } x} p_X(x) \cdot p_Y(w - x)$$

Example: Suppose two dice are thrown, and the random variables X and Y represent the numbers showing on the faces of the first and second die, respectively, so: $p_X(x) = p_Y(y) = 1/6$.

We are primarily interested in the sum of the two faces, so define a new random variable W = X + Y.

Theorem 3.8.3 says that the pdf of W will be

$$p_W(w) = \sum_{\text{all } x} p_X(x) \cdot p_Y(w - x)$$

Suppose we want to find $p_W(4)$:

 $p_W(4) = p_X(1) \cdot p_Y(3) + p_X(2) \cdot p_Y(2) + p_X(3) \cdot p_Y(1)$

$$=\frac{1}{36} + \frac{1}{36} + \frac{1}{36} = \frac{3}{36}$$

Theorem: (3.8.1 - continuous case) Suppose *X* and *Y* are independent continuous random variables with pdfs $f_X(x)$ and $f_Y(y)$, respectively.

Define a random variable W to be the sum of X and Y.

$$f_W(w) = \int_{-\infty}^{\infty} f_X(x) \cdot f_Y(w - x) \, dx$$

Theorem: (3.8.1 - continuous case) Suppose *X* and *Y* are independent continuous random variables with pdfs $f_X(x)$ and $f_Y(y)$, respectively.

Define a random variable W to be the sum of X and Y.

Then the pdf of *W* is:

$$f_W(w) = \int_{-\infty}^{\infty} f_X(x) \cdot f_Y(w - x) \, dx$$

The above integral is called the **convolution** of the functions $f_X(x)$ and $f_Y(y)$.

Products of Independent Random Variables

Theorem: (3.8.3) Suppose *X* and *Y* are independent continuous random variables with pdfs $f_X(x)$ and $f_Y(y)$, respectively.

Define a random variable W to be the product XY.

$$f_W(w) = \int_{-\infty}^{\infty} \frac{1}{|x|} f_X(w/x) \cdot f_Y(x) \, dx$$

Quotients of Independent Random Variables

Theorem: (3.8.2) Suppose *X* and *Y* are independent continuous random variables with pdfs $f_X(x)$ and $f_Y(y)$, respectively. Assume that *X* is zero for at most a set of isolated points.

Define a random variable W to be the quotient Y/X.

$$f_W(w) = \int_{-\infty}^{\infty} |x| f_X(x) \cdot f_Y(wx) \, dx$$

Theorem: (variation of 3.8.1 - discrete case) Suppose *X* and *Y* are discrete random variables with joint pdf $p_{XY}(x, y)$.

Define a random variable W to be the sum of X and Y.

$$p_W(w) = \sum_{\text{all } x} p_{XY}(x, w - x)$$

Theorem: (variation on 3.8.1 - continuous case) Suppose *X* and *Y* are independent continuous random variables with joint pdf $f_{XY}(x, y)$.

Define a random variable W to be the sum of X and Y.

$$f_W(w) = \int_{-\infty}^{\infty} f_{XY}(x, w - x) \, dx$$

Products of Non-Independent Random Variables

Theorem: (variation on 3.8.3) Suppose *X* and *Y* are independent continuous random variables with joint pdf $f_{XY}(x, y)$.

Define a random variable W to be the product XY.

$$f_W(w) = \int_{-\infty}^{\infty} \frac{1}{|x|} f_{XY}(w/x, x) \, dx$$

Quotients of Non-Independent Random Variables

Theorem: (variation of 3.8.2) Suppose *X* and *Y* are independent continuous random variables with joint pdf $f_{XY}(x, y)$. Assume that *X* is zero for at most a set of isolated points.

Define a random variable W to be the quotient Y/X.

$$f_W(w) = \int_{-\infty}^{\infty} |x| f_{XY}(x, wx) \, dx$$