IB Paper 7: Probability How do Moment Generating Functions Work?

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Abstract

This note shows how the definition of moment generating functions leads to short cuts to calculating mean, variance and higher moments of a distribution. Expanding the generating function as a power series, is demonstrated that if it is differentiated *n* times, the leading order term includes a component in $\mathbb{E}[X^n]$ which can be used to calculate the *n*th moment.

Continuous generating functions

Start with the continuous case because the algebra is tidier. We have a continuous random variable X. Define a function f of a 'transform variable' s which takes the form,

$$f(s) = \exp(-sX).$$

The exponential can also be written as a power series (see Maths Data Book),

$$f(s) = 1 + (-sX) + \frac{(-sX)^2}{2!} + \frac{(-sX)^3}{3!} + \mathcal{O}(s^4).$$

Now define the continuous generating function g(s) to be the expectation of f, across all possible values of X with s fixed,

$$g(s) = \mathbb{E}[f(s)] = 1 - s\mathbb{E}[X] + \frac{s^2}{2}\mathbb{E}[X^2] - \frac{s^3}{6}\mathbb{E}[X^3] + \mathcal{O}(s^4).$$

Calculate the derivatives of g, and evaluate at s = 0 to keep only the leading order term,

$$g'(s) = -\mathbb{E}[X] + s\mathbb{E}[X^2] - \frac{s^2}{2}\mathbb{E}[X^3] + \mathcal{O}(s^3) \quad \Rightarrow \quad g'(0) = -\mathbb{E}[X],$$

$$g''(s) = \mathbb{E}[X^2] - s\mathbb{E}[X^3] + \mathcal{O}(s^2), \quad \Rightarrow \quad g''(0) = \mathbb{E}[X^2].$$

So the mean and variance are,

$$\mathbb{E}[X] = -g'(0), \text{ and } \mathbb{V}[X] = \mathbb{E}[X^2] - \mathbb{E}[X]^2 = g''(0) - g'(0)^2$$

This procedure works for any *n*th moment, because if the generating function is differentiated *n* times, the leading order term includes a component in $\mathbb{E}[X^n]$.

Discrete generating functions

We have a discrete random variable X. Define a function f of a 'transform variable' z which for a particular value of X takes the form,

$$f(z) = z^X.$$

Differentiating with respect to z, with X fixed,

$$f'(z) = Xz^{X-1},$$

 $f''(z) = X(X-1)z^{X-2}.$

Write f(z) as a Taylor series about z = 1. Using the Maths Data Book notation, x = 1 and h = z - 1. We have,

$$f(z) = f(1) + (z-1)f'(1) + \frac{(z-1)^2}{2!}f''(1) + \mathcal{O}((z-1)^3).$$

Substituting for the derivatives of f evaluated at z = 1,

$$f(z) = 1 + (z-1)X + \frac{(z-1)^2}{2}X(X-1) + \mathcal{O}((z-1)^3).$$

Now define the discrete generating function g(z) to be the expectation of f, across all values of X with z fixed,

$$g(z) = \mathbb{E}[f(z)] = 1 + (z-1)\mathbb{E}[X] + \frac{(z-1)^2}{2}\mathbb{E}[X(X-1)] + \mathcal{O}((z-1)^3).$$

Calculate derivatives of g and then evaluate at z = 1 to keep the leading order term,

$$g'(z) = \mathbb{E}[X] + (z-1)\mathbb{E}[X(X-1)] + \mathcal{O}((z-1)^2) \implies g'(1) = \mathbb{E}[X],$$

$$g''(z) = \mathbb{E}[X(X-1)] + \mathcal{O}((z-1)) \implies g''(1) = \mathbb{E}[X(X-1)] = \mathbb{E}[X^2 - X] = \mathbb{E}[X^2] - \mathbb{E}[X].$$

So the mean and variance are, in these terms,

$$\begin{split} \mathbb{E}[X] &= g'(1), \end{split} \\ \mathbb{V}[X] &= \mathbb{E}[X^2] - \mathbb{E}[X]^2 = \left(\mathbb{E}[X^2] - \mathbb{E}[X]\right) + \mathbb{E}[X] - \mathbb{E}[X]^2, \cr \mathbb{V}[X] &= g''(1) + g'(1) - g'(1)^2. \end{split}$$

This procedure works for any *n*th moment, because if the generating function is differentiated *n* times, the leading order term includes a component in $\mathbb{E}[X^n]$.