

The Kelly Criterion

The Kelly Criterion was first introduced in 1956 (*“A New Interpretation of Information Rate”*) in solving the St. Petersburg Paradox (a coin flip game). It was later extended by Edward Thorp (*“Beat the Dealer”*) for card games. A recent writing by Thorp went through a great length of the treatment (*“The Kelly Criterion in Blackjack, Sport Betting, and the Stock Market”*) but a few key details were omitted. This article follows Jane Hung’s treatment (*“Betting with the Kelly Criterion”*) with focus on using the binomial distribution (see Section 7.1 of Thorp’s Thesis).

The mathematical technique to approximate the random variable X_i using a binomial distribution with two distinct values at $\mu + \sigma$ and $\mu - \sigma$ each with equal probability of $\frac{1}{2}$, *i.e.*,

$$P(X_i = \mu + \sigma) = P(X_i = \mu - \sigma) = \frac{1}{2} \quad (1)$$

This binomial distribution has a mean of μ and variance of σ^2 .

Assuming that the initial asset is Y_0 , risk-free annual interest rate r , and betting fraction (position sizing) f . At the n -th compounding cycle within a year, the total asset Y_n grows into,

$$Y_n = Y_0 \prod_{i=1}^n \left(1 + (1-f)\frac{r}{n} + fX_i \right) \quad (2)$$

The expected logarithmic growth of the asset is:

$$E \left[\ln \left(\frac{Y_n}{Y_0} \right) \right] = E \left[\sum_{i=1}^n \ln \left(1 + (1-f)\frac{r}{n} + fX_i \right) \right] \quad (3)$$

Accounting for the binomial distribution described in equation (1) for X_i

$$\begin{aligned}
E \left[\ln \left(\frac{Y_n}{Y_0} \right) \right] &= \sum_{i=1}^n \frac{1}{2} \ln \left(1 + (1-f) \frac{r}{n} + f \left(\frac{\mu}{n} + \frac{\sigma}{\sqrt{n}} \right) \right) \\
&\quad + \frac{1}{2} \ln \left(1 + (1-f) \frac{r}{n} + f \left(\frac{\mu}{n} - \frac{\sigma}{\sqrt{n}} \right) \right) \\
&= \frac{n}{2} \ln \left(\left(1 + (1-f) \frac{r}{n} + \frac{f\mu}{n} \right)^2 - \left(\frac{f\sigma}{\sqrt{n}} \right)^2 \right)
\end{aligned} \tag{4}$$

Carrying out the Taylor expansion of the above equation at $f = 0$ and taking n to infinite (continuously compounding),

$$E \left[\ln \left(\frac{Y_\infty}{Y_0} \right) \right] = r + (\mu - r)f - \frac{(f\sigma)^2}{2} \tag{5}$$

Solving the quadratic equation of f to maximize the logarithmic growth, the Kelly Criterion for the optimal betting fraction f^* is

$$f^* = \frac{\mu - r}{\sigma^2} \tag{6}$$