

# To Bet or Not To Bet: The Modified Kelly Criterion

Dani Chu, Yifan Wu & Tim B. Swartz



## Sports Gambling

### ► Point Spread:

Golden State Warriors	+ 6.5	1.909
-----------------------	-------	-------

Sacramento Kings	- 6.5	1.909
------------------	-------	-------

### ► Over/Under:

Capitals vs Penguins	Over	5.5 goals	1.90
----------------------	------	-----------	------

Capitals vs Penguins	Under	5.5 goals	1.90
----------------------	-------	-----------	------

### ► Other Props:

Belichick Hoodie Colour	Blue	1.92
-------------------------	------	------

Belichick Hoodie Colour	Grey	1.92
-------------------------	------	------

## Profitable Systems

- ▶ A gambling system (often found in sports) is profitable with
  - ▷ Wager of size  $\$x$
  - ▷ System win probability  $p$
  - ▷ Return of  $\$x \cdot \theta$  on a win and 0 on a loss

If

$$(-x)(1 - p) + (x\theta - x)p > 0 \rightarrow p > 1/\theta$$

## The Kelly Criterion

- ▶ The Kelly criterion (Kelly 1956) provides a gambler an optimal fraction of a bankroll for wagering given probability  $p$  of winning a bet.

$$k(p) = \begin{cases} \frac{p\theta-1}{\theta-1} & p > 1/\theta \\ 0 & p \leq 1/\theta \end{cases}$$

- ▶ **Problem:** Experienced gamblers claim  $k(p)$  is too large
- ▶ **Reason:**  $p$  is not known and often overestimated with data
- ▶ **The Fix:** Model the unknown parameter  $p$  and estimate the unknown  $k(p)$  with the estimator  $f = f(x)$



## Modified Kelly Criterion

- ▶ To assess the quality of  $f$  we use  $l_i(f, p)$  as loss function  $i$
- ▶ Use a Bayes estimator  $f$  which minimizes the Bayes risk
  - ▷ Will minimize expected posterior loss

$$G(f) = \int_0^1 l_0(f, p) \pi(p | x) dp$$

- ▶ Posterior distribution of  $p$  is defined by
  - ▷ historical data  $x \sim \text{Binomial}(n, p)$  from historical win/loss data
  - ▷ prior distribution  $p \mid \sim \text{Beta}(a, b)$
- ▶ For different loss functions (see poster) we get different  $f$ 
  - ▷ Can be solved for directly or by estimating the integral through computation

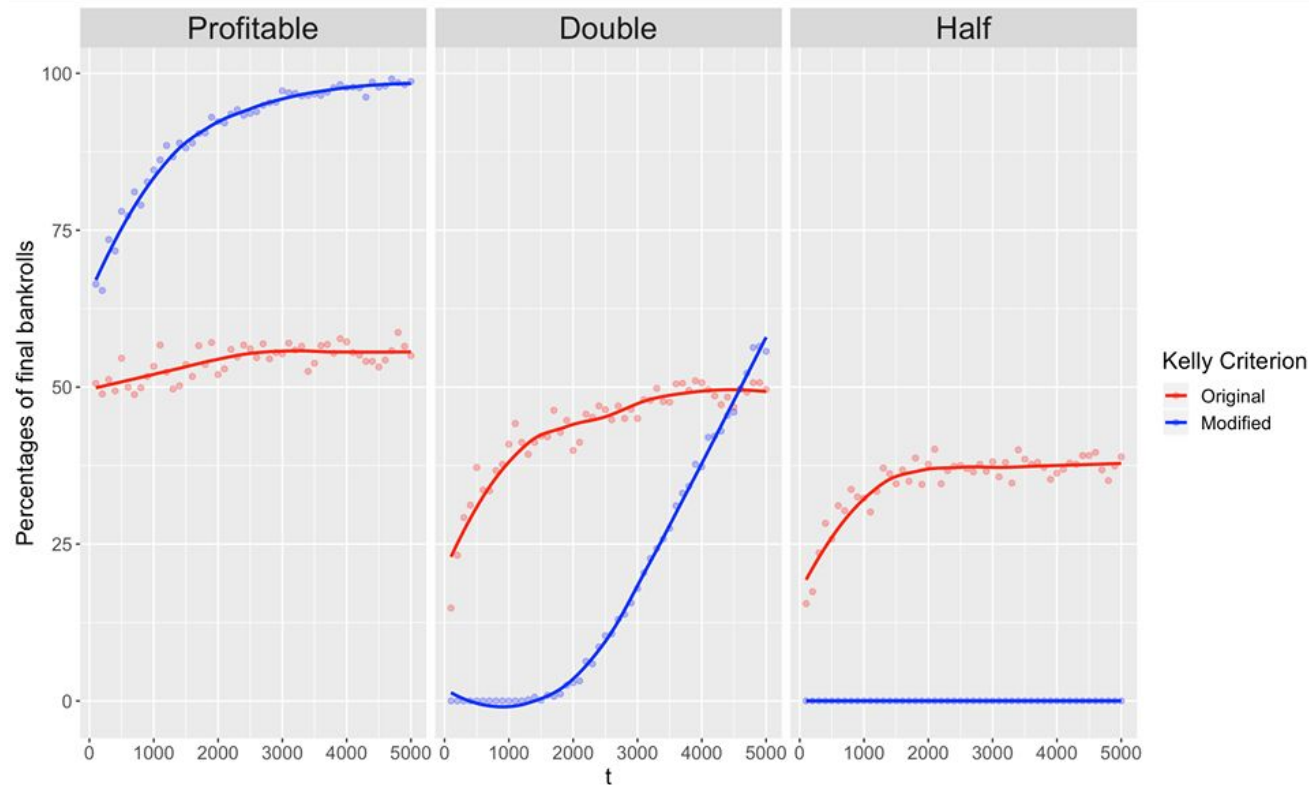


## In Practice

- ▶ **Find a Winning Gambling System**
  - ▷ Professor MJ's push-back phenomenon in NBA playoff games point spread system with  $271/484 = 0.56$  win rate.
- ▶ **Define Beta distribution with  $a$  and  $b$ :**
  - ▷  $a = b = 50$ 
    - ▷ Centered at 0.5, and  $\sim 95\%$  of the prior probability in the interval (0.4, 0.6)
- ▶ **Choose loss function:**
  - ▷  $l_{3b}(f, p) = (l_{f > k(p)} + 2) |f - k(p)|^{1.5}$  - lies halfway between absolute error loss and squared error loss. Penalty of overestimation is 1.5 times worse than underestimation
- ▶ Starting with an \$1000 bankroll would have produced a final bankroll of
  - ▷ Original Kelly: \$668.34, using 7.6% of initial bankroll for each bet
  - ▷ Modified Kelly: \$794.89, using 4.7% of initial bankroll for each bet



# In Simulation



**THANKS!**

# Any Questions?

Please find me at the E-Poster Session, Poster 11!

Or see our paper in [JQAS](#)!

**Dani Chu**

dani\_chu@sfu.ca

@chuurveg

danichusfu.github.io

**Yifan (Lucas) Wu**

yifan\_wu@sfu.ca

**Tim B. Swartz**

tim@stat.sfu.ca

people.stat.sfu.ca/~tim/

Are you a student interested in Sports Analytics?

Enjoy Vancouver and want to come back on September 22nd?

Check out the Vancouver Whitecaps Datathon at [www.VanSASH.com](http://www.VanSASH.com)!



**NSERC  
CRSNG**