zrml-neo-swaps Documentation

0.4.3 (October 30, 2023)

1 Introduction

This document provides the mathematical and technical details for zrml-neo-swaps. The automatic market maker (AMM) implemented by zrml-neo-swaps is a variant of the Logarithmic Market Scoring Rule (LMSR; [2]) which was first developed by Gnosis (see https://docs.gnosis.io/conditionaltokens/docs/introduction3/). We often refer to it as AMM 2.0.

Unlike the typical implementation using a cost function (see [1]), this implementation of LMSR is a constant-function market maker (CFMM), similar to the classical constant product market maker, which allows us to implement dynamic liquidity. In other words, liquidity providers (LPs) can come and go as they please, allowing the market to self-regulate how much price resistance the AMM should provide.

2 The Trading Function

We consider a prediction market with n outcomes, denoted by $1, \ldots, n$ for simplicity. Every complete set of outcome tokens is backed a unit of collateral, denotes by \$. The AMM operates on a *liquidity pool* (or just *pool*), which consists of a *reserve* (r_1, \ldots, r_n) of outcome tokens and a *liquidity parameter b*. The trading function is defined as

$$\varphi(b,r) = \sum_{i} e^{-r_i/b}.$$

In fact, $\varphi(b,r)$ must always equal 1. This means that a trader may change the reserve from r to r' and receive the delta provided that $\varphi(b,r')=1$. We denote such a trade by $r \to r'$. We call these outcome-to-outcome (O2O) swaps.

However, we do not allow users to execute these types of trades. Instead, we only allow buys (exchange collateral for outcome tokens) and sells (exchange outcome tokens for collateral).

3 Buying and Selling

3.1 Buying

Buying and selling is implemented by combining complete set operations (exchange x dollars for x units of every outcome) and O2O swaps.

Alice wants to swap x dollars for units of outcome i. This is done by exchanging x dollars for x complete sets and then selling all outcomes $k \neq i$ for more i using an O2O swap $r \to r'$, which yields y(x) additional units of i. Ignoring swap fees, this modifies the reserve to r', where $r'_k = r_k + x$ for $k \neq i$ and $r'_i = r_i - y(x)$. As trades don't change the invariant, we have $1 = \sum_k e^{-r'_k/b}$. Thus, using $1 = \varphi(b, r) = \sum_k e^{-r_k/b}$,

$$1 = \sum_{k} e^{-r'_{k}/b}$$

$$= \sum_{k \neq i} e^{-(r_{k}+x)/b} + e^{-(r_{i}-y(x))/b}$$

$$= e^{-x/b} \sum_{k \neq i} e^{-r_{k}/b} + e^{y(x)/b} e^{-r_{i}/b}$$

$$= e^{-x/b} (1 - e^{-r_{i}/b}) + e^{y(x)/b} e^{-r_{i}/b}.$$

Rearranging these terms gives

$$e^{y(x)/b} = e^{r_i/b}(1 - e^{-x/b}(1 - e^{-r_i/b})),$$

and, thus,

$$y(x) = b \ln(e^{r_i/b}(1 - e^{-x/b}(1 - e^{-r_i/b})))$$

= $b \ln(1 - e^{-x/b}(1 - e^{-r_i/b})) + r_i$
= $b \ln(e^{x/b} - 1 + e^{-r_i/b}) + r_i - x$.

Note that the total amount of outcome i that Alice receives is y(x) from the O2O trade and x from the complete set operation. We denote this by z(x) = y(x) + x.

This allows us to calculate the *spot price* of outcome i

$$p_i(b,r) = \lim_{x \to 0} \frac{x}{z(x)} = \frac{1}{z'(0)} = \frac{1}{y'(0) + 1}.$$

Calculating the derivative of y yields

$$y'(x) = \frac{e^{x/b}}{e^{x/b} - 1 + e^{-r_i/b}} - 1$$

and thus $y'(0) = e^{r_i/b} - 1$, which yields $p_i(b, r) = e^{-r_i/b}$.

Note that this means

$$1 = \varphi(b, r) = \sum_{i} p_i(b, r).$$

In particular, (p_1, \ldots, p_n) always maps to a probability distribution.

Trading fees are specified as fractional (a fee of f = .01 means that 1% are charged) and deducted from the amount of collateral before the complete set operations are executed. In other words, the liquidity providers receive fx dollars (fees are distributed pro rata amongst the liquidity providers) and Alice goes through the entire process described above with $\tilde{x} = (1 - f)x$ in place of x. The spot price taking the fees into account is (as expected)

$$p_i(b, r, f) = (1 - f)^{-1} e^{-r_i/b}.$$

3.2 Selling

Alice wants to swap x units of i for dollars. This is done by selling x' < x units of i for v(x) = x - x' units of all outcomes $k \neq i$ and then selling v(x) units of complete sets, which yields v(x) dollars. Ignoring swap fees, this modifies the reserve from r to r', where $r'_k = r_k - v(x)$ and $r_i = r_i + x'$. Using $1 = \varphi(b, r')$ and x' = x - v(x), we get

$$1 = \sum_{k} e^{-r'_{k}/b}$$

$$= \sum_{k \neq i} e^{-(r_{k} - v(x))/b} + e^{-(r_{i} + x')/b}$$

$$= e^{v(x)/b} \sum_{k \neq i} e^{-r_{i}/b} + e^{-x'/b} e^{-r_{i}/b}$$

$$= e^{v(x)/b} (1 - e^{-r_{i}/b}) + e^{-x/b} e^{v(x)/b} e^{-r_{i}/b}$$

$$= e^{v(x)/b} (1 - e^{-r_{i}/b} + e^{-(r_{i} + x)/b}).$$

Thus, we get

$$e^{-v(x)/b} = 1 - e^{-r_i/b} + e^{-(r_i+x)/b}$$
.

which in turn yields

$$v(x) = -b \ln(1 - e^{-r_i/b} + e^{-(x+r_i)/b})$$

$$= -b \ln(e^{r_i/b} - 1 + e^{-x/b}) + r_i$$

$$= -b \ln(e^{(x+r_i)/b} - e^{x/b} + 1) + r_i + x.$$

Trading fees are deducted from the amount of collateral received from the complete set operation. In other words, the liquidity providers receive fv(x) dollars and Alice receives (1-f)v(x). The selling price (the amount of collateral received for each unit of x), is then (as expected)

$$\lim_{x \to 0} \frac{(1-f)v(x)}{x} = (1-f)v'(0) = (1-f)e^{-r_i/b} = (1-f)p_i(b,r).$$

This leads to a typical bid-ask spread between buy and sell price.

4 Dynamic Liquidity

Liquidity may be added or removed dynamically to regulate the market's price resistance. Each LP's share of the pool is tracked using pool share tokens, which record their *pro rata* share of the pool.

We consider a pool with liquidity parameter b, reserve r and a total issuance of q pool shares.

4.1 Adding Liquidity

Alice wants to add liquidity to the pool. She's willing to pay x dollars. To implement this, she first spends x dollars to buy x complete sets.

Now let i be so that $r_i = \max_k r_k$. Let $\lambda = x/r_i$ and $\mu = 1 + \lambda$. For each k, Alice moves λr_k units of k into the pool and receives λq pool shares. The liquidity parameter changes from b to $b' = \mu b$. Alice's transfers change the reserve from r to $r' = \mu r$.

The new total issuance of pool shares is μq and Alice's share of the pool now is λ/μ . Note that Alice retains the balance $(x)^n - \lambda r$ of "left-over tokens".

4.2 Withdrawing Liquidity

Alice wants to withdraw liquidity from the pool. She's willing to burn p pool shares.

Let $\lambda = p/q$ and $\mu = 1 - \lambda$. For each k, Alice receives λr_k units of k from the pool. The liquidity parameter changes from b to $b' = \mu b$. Alice's transfers change the reserve from r to $r' = \mu r$.

Alice could now convert $x = \min_i r_i$ complete sets of her newly received funds into x dollars. The remainder of the funds will remain in her wallet until the market resolves or she opts to sell them.

4.3 Fee Distribution

Fees are distributed pro rata amongst the liquidity providers. These funds are completely separate from the reserve used for trading. Transferring the fees into the pool (like the constant product market maker does) wouldn't make any sense here as collateral is not directly traded on the pool.

5 Creating Pools

Creating a pool is straightforward. The initial odds are defined by adding different amounts of each outcome to the pool. If Alice wants to deposit liquidity worth x units of collateral with initial probability p, then she starts off by buying x complete sets. The following algorithm is used to calculate how many units of each outcome go into the pool. Alice retains the other tokens as "left-overs".

Let b = 1, and let $r_i = -b \ln p_i$ for all i. Now let $y = x/\max_i r_i$. Then $yr_i \le x$ for all i and there exists i_0 so that $yr_{i_0} = x$. Set $\tilde{r}_i = yr_i$ and $\tilde{b} = yb$. Then

$$p_i(\tilde{r}) = e^{-\tilde{r}_i/\tilde{b}} = e^{-r_i/b} = p_i$$

and $\max_i \tilde{r}_i = x$ (so Alice uses up at least one of her outcome balances).

In pseudocode:

```
Procedure CalculateBalances(p[1...n], x)

b ← 1 // Initialize b, larger values may be picked for

numerical stability

For i from 1 to n do

r[i] ← -b * log(p[i])

End For

y ← x / max(r[1...n])

For i from 1 to n do

r[i] ← y * r[i]
```

Listing 1: Procedure to Calculate Balances

6 Additional Formulas

6.1 Estimated Price After Execution

After executing a buy for x units of collateral for outcome i, the new reserve of i is

$$r'_i = r_i - y((1-f)x) = -b\ln(1 - e^{-(1-f)x/b}(1 - e^{-r_i/b})).$$

Thus, the new price is

$$p_i(f, b, r') = \frac{1}{1 - f} (1 - e^{-(1 - f)x/b} (1 - e^{-r_i/b})).$$

After executing a sell of x units of outcome i for collateral, the new reserve of i is

$$r'_i = r_i + x' = r_i + x - v(x) = b \ln(e^{(x+r_i)/b} - e^{x/b} + 1).$$

The new price is therefore

$$p_i(f, b, r') = \frac{1}{1 - f} (e^{(x+r_i)/b} - e^{x/b} + 1).$$

7 Numerical Issues

Special care must be taken to avoid over- and underflows when calculating expressions like

$$y(x) = b \ln(e^{x/b} - 1 + e^{-r_i/b}) + r_i - x,$$

$$v(x) = -b \ln(e^{r_i/b} - 1 + e^{-x/b}) + r_i.$$

The magnitude of y(x) is the same as x, but the exponentials $e^{x/b}$ and $e^{-r_i/b}$ over- or underflow easily.

What makes matters more is that the logarithm's derivative approaches infinity near zero, so any rounding errors when calculating the argument of ln in that area will amplify.

Let A=10. Python calculates $e^A=22026.465794806718$ and $e^{-A}=4.5399929762484854\cdot 10^{-5}$. The fixed crate (see https://crates.io/crates/fixed) can represent these using FixedU128<U80> without considerable loss of precision or risk of over- or underflow. Let M=Ab, $\varepsilon=e^{-A}$ and $c=\frac{1}{10}$. The following restrictions will prevent numerical issues.

- If $p_i(b,r) < \varepsilon$ (or, equivalently, $r_i > M$), then selling is not allowed.
- If $p_i(b,r) \geq \varepsilon$ (or, equivalently, $r_i \leq M$), then the amount x must satisfy $x \leq M$ when calculating v(x) (to avoid overflows).

- Sells that push the price below ε are not allowed.
- When buying, then the ln argument must satisfy $e^{x/b} 1 + e^{-r_i/b} \ge c$ and $x \le M$ (to avoid overflows).

The last restriction may need some elaboration. We actually allow $e^{-r_i/b}$ to underflow to zero during calculation, but only provided that the remaining term $e^{x/b} - 1$ is large enough to avoid this to cause any damage. In fact, on the interval $[c, \infty)$, the logarithm has derivative $\leq c^{-1} = 10$, so all rounding errors in the argument of ln cause (only) ten times the error in the calculation of ln.

How bad are these restrictions? The restriction $x \leq M$ is completely irrelevant: Suppose Alice executes a trade of y(x) units of outcome i for x = M dollars, the maximum allowed value. Let $q = 1 - e^{-r_i/b} \in (0, 1)$. Then

$$\ln(M) - \ln(M - q) = \ln\left(\frac{M}{M - q}\right)$$

$$\leq \ln\left(\frac{M}{M - 1}\right)$$

$$\approx 4.540096037046513 \cdot 10^{-5}$$

$$\leq 10^{-4} = \varepsilon.$$

Then we have

$$y(x) = b \ln(M - q) + r_i - x$$

$$\geq b(\ln(M) - \varepsilon) + r_i - x$$

$$= M - b\varepsilon + r_i - x$$

$$= r_i - b\varepsilon.$$

Thus, Alice receives all funds from the pool except $b\varepsilon$, which is very small unless the pool contains an inordinate amount of liquidity.

The restriction that sells can't push prices below ε is liable to cause some confusion for the users. Nevertheless, it's unclear if this causes them any real inconvenience.

Despite the rule that sells can't push prices below ε , prices can be moved below ε . The issue in a market with three or more outcomes A, B, C, \ldots is that if C is a clear underdog and most of the trading happens between the favorites A and B, then the price of C might drop below the allowed threshold. This is what the last rule handles.

The last rule is made to ensure that even if the price is very small (below ε) or even underflows (both of which *shouldn't* ever happen thanks to trading fees), buying the outcome is still possible, allowing users to push the price back up again. The restriction is that there's essentially a minimum amount. In fact, in the pessimistic case (price underflows to zero), the restriction becomes

$$x \ge b \ln(c+1) \approx bc = \frac{b}{10}.$$

That's certainly a tall order if the market is very liquid. If this continues to be a problem, we can test how much smaller c can be made without causing trouble.

References

- [1] Yiling Chen and Jennifer Wortman Vaughan, A new understanding of prediction markets via no-regret learning, EC '10: Proceedings of the 11th ACM conference on Electronic commerce, June 2010, Pages 189–198. https://doi.org/10.1145/1807342.1807372
- [2] Robin Hanson, Logarithmic Market Scoring Rules for Modular Combinatorial Information Aggregation, The Journal of Prediction Markets, 1(1), May 2003. https://doi.org/10.5750/jpm.v1i1.417