Composing Networks of Automated Market Makers

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ABSTRACT

Automated market makers (AMMs) are automata that trade electronic assets at rates set by mathematical formulas. AMMs are usually implemented by smart contracts on blockchains. In practice, AMMs are often *composed*: and outputs from AMMs can be directed into other compatible AMMs. This paper proposes a mathematical model for AMM composition. We define sequential and parallel composition operators for AMMs in a way that ensures that AMMs are closed under composition, in a way that works for "higher-dimensional" AMMs that manage more than two asset classes, and so the composition of AMMs in "stable" states remains stable.

1 INTRODUCTION

Decentralized finance (or "DeFi") has become a booming area of applied distributed computing. For example, between June 2020 and October 2020, the total value of assets locked in decentralized finance (DeFi) protocols surged from \$1 billion to \$7.7 billion [21]. An automated market maker ("AMM") is an automaton that has custody of several pools of assets, sets prices for those assets according to a mathematical formula, and is always willing to trade those assets at those prices. Unlike traditional "order book" traders, AMMs do not need to match up (and wait for) compatible buyers and sellers. Today, AMMs such as Uniswap [6], Bancor [16], Balancer [18], and others have become one of the most popular ways to trade electronic assets.

Here is an example of a *constant-product* AMM loosely based on Uniswap v1 [6]. The AMM has state (x, y) if it has custody of x units of asset X, and y units of asset Y. The AMM's state is subject to the invariant that the product xy is constant. The AMM's states thus lie on the hyperbolic curve xy = c, for x, y > 0 and some constant c > 0. If a trader transfers dx units of X to the AMM, the AMM will transfer dy units of Y back to the trader, preserving the invariant (x + dx)(y - dy) = xy = c. The client profits if the value of dy units of Y exceeds the value of dx units of X in the current market (or at another AMM).

The *price* of asset Y in units of X at state (x, y) is the curve's slope at that point. Trades move the AMM's state along the curve: trading X for Y makes X cheaper and Y more expensive, a phenomenon known as *slippage*. Usually, an AMM's state reflects current market conditions: if a constant-product AMM is in state (x, y), then the market value of x

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units of X should be y units of Y, because otherwise a trader can make an arbitrage profit by buying the undervalued asset. Note that a constant-product AMM can adjust to any (non-zero) market rate between X and Y, and every AMM state matches some market valuation.

It is natural to compose AMMs. If AMM A trades assets X and Y, and B trades assets Y and Z, then a trader can buy Z with X by transferring X to A, feeding A's Y output to B, and pocketing B's Z output. Existing systems of AMMs encourage exactly this kind of composition: for Uniswap V1, the intermediate asset *Y* would be ether (ETH), and for Bancor, it would be Bancor network token (BNT). Here is the question at the heart of this paper: can we treat the result of composing these two constant-product AMMs as a "black box" AMM for trading X for Z? Note that this composition is not itself a constant-product AMM, so the class of constant-product AMMs is not closed under composition for any reasonable notion of composition. Can we instead pick a broader AMM definition that does support common-sense notions of composition? What constraints on pricing formulas yield AMMs that behave reasonably under composition? What should composition mean when AMMs trade more than two kinds of assets? In short, what notions of AMM composition make sense?

This paper makes the following contribution: we propose novel operators for sequential and parallel composition of AMMs. For these notions of composition to make sense, it will be necessary to make small, but critical adjustments to prior mathematical models for AMMs. In particular, to define a sensible notion of composition for "higher-dimensional" AMMs that manage more than two asset classes, we will need to introduce intermediate projection and virtualization operators to arrive at well-structured notions of composition.

This paper is organized as follows. Section 2 describes the problem and the model of computation, Section 3 presents in axiomatic form the properties a practical AMM should satisfy, Section 4 shows that all such AMMs have a common underlying topological structure, and Section 5 presents basic mathematical operators useful for defining composition. Section 6 defines *sequential composition*, where the outputs of one trade becomes the inputs to another. Some careful choices are needed to impose order on composition of AMMs that manage more than two assets. Section 7 defines *parallel composition*, where a trader decides how to split a trade

among alternative AMMs. Section 8 surveys related work, and Section 9 discusses future directions and open problems.

2 MODEL

We use the following notation and terminology. Vectors are in bold face (x) and scalars in italics (x). Variables, scalar or vector, are usually taken from the end of the alphabet (x,y,z), and constants from the beginning (a,b,c). If $a=(a_1,\ldots,a_n)$ and $b=(b_1,\ldots,b_m)$, then (a,b) is the vector $(a_1,\ldots,a_n,b_1,\ldots,b_m)$. Vector comparisons are componentwise: for vectors of the same dimension, $a\leq b$ means that each $a_i\leq b_i$. Constant vectors $1=(1,\ldots,1)$ and $0=(0,\ldots,0)$ have dimension that will be clear from context. $\mathbb{R}_{>0}=\{x\in\mathbb{R}|x>0\}$, and $\mathbb{R}_{<0}=\{x\in\mathbb{R}|x<0\}$. We use $x\not\subseteq x'$ to mean that $x\leq x'$ but $x\neq x'$, so at least one inequality is strict.

A function $f: \mathbb{R}^n \to \mathbb{R}$ is *strictly convex* if for all $t \in (0, 1)$ and distinct $x, x' \in \mathbb{R}^n$, f(tx+(1-t)x') < tf(x)+(1-t)f(x'). The function is *strictly concave* if the inequality is reversed. A set X is *strictly convex* if for all distinct $x, x' \in X$ and $t \in (0, 1)$, tx + (1 - t)x' lies in the interior of X.

For $A: \mathbb{R}^n_{>0} \to \mathbb{R}$, the set $\{x \in \mathbb{R}^n_{>0} | A(x) = c\}$ is called a *level set* at c, and the set $\{x \in \mathbb{R}^n_{>0} | A(x) \ge c\}$ is called the *upper contour set* at c.

We use L1 and L2 norms: if $\boldsymbol{v}=(v_1,\dots,v_n)$ is a vector, then $\|\boldsymbol{v}\|_1=\sum_{i=1}^n|v_i|$ and $\|\boldsymbol{v}\|_2=\sqrt{\sum_{i=1}^nv_i^2}$.

2.1 AMM State Spaces are Manifolds

An *asset* might be a cryptocurrency, a token, an electronic deed to property, and so on. Assets can fluctuate in value, and participants may want to trade assets to respond to past price changes or to anticipate future price changes.

An n-dimensional AMM trades across assets X_1, \ldots, X_n . Each AMM state has the form $\mathbf{x} = (x_1, \ldots, x_n) \in \mathbb{R}^n_{>0}$, where each x_i is the number of units of asset X_i in the AMM's custody. AMM states are points in a (twice) differentiable manifold, a higher-dimensional generalization of a surface or curve. A trade moves the AMM from state $\mathbf{x} = (x_1, \ldots, x_n)$ to state $\mathbf{x}' = (x_1', \ldots, x_n')$. For each i where $x_i' > x_i$, the trader pays $x_i' - x_i$ units of X_i to the AMM, while if $x_i' < x_i$, the AMM pays $x_i - x_i'$ units of X_i to the trader. We call $\mathbf{x} - \mathbf{x}'$ a profit-loss vector. Requiring the AMM state space to be a differentiable manifold ensures that both prices and slippage change gradually rather than abruptly, although we will see in Section 3 that not every manifold makes sense as an AMM state space.

2.2 System Model

There are two kinds of participants in decentralized finance. (1) *Traders* transfer assets to AMMs, and receive assets back. Traders can compose AMMs into networks in complicated

ways. (2) *Liquidity providers* (or "providers") fund the AMMs by lending assets, and receiving shares, fees, or other profits. Traders and providers play a kind of alternating game: traders modify AMM states by adding and removing assets, and providers can respond by adding or removing assets, reinvesting fees, or adjusting other AMM properties.

Today, AMMs are usually implemented by smart contracts on blockchains. For our purposes, a *blockchain* is a highly-available, tamper-proof distributed ledger that records which participants own various assets. A *smart contract* is a public program that controls how assets are recorded and transferred on a blockchain. Smart contracts typically support *atomic transactions* that allow traders to execute atomic sequences of trades on multiple AMMs. The analysis given in this paper is largely independent of the the particular technology used to implement blockchains and smart contracts.

3 DEFINITIONS

3.1 Common-Sense Axioms

Although any automaton that trades assets can be regarded as an AMM, only those AMMs that satisfy certain properties make sense in practice. To make this presentation self-contained, we list some informal, common-sense axioms any practical AMM should satisfy, and then restate these informal requirements as more precise mathematical properties. These properties mirror prior proposals [5, 8, 17], with some adjustments to encompass higher-dimensional AMMs (those that trade more than two kinds of assets), and to facilitate later introduction of composition operators.

Informal Axiom 1 (Continuity). Every AMM state should define precise rates of exchange between every pair of assets, and trades should change these rates gradually rather than abruptly.

Informal Axiom 2 (Expressivity). An AMM must be able to adapt to any market conditions.

If an AMM is unable to adapt to market conditions that cause one asset to be undervalued with respect to the others, then traders will drain all of the undervalued asset from the AMM at the expense of the providers.

Informal Axiom 3 (Stability). Every AMM state should be the appropriate response to some possible market condition.

If no market condition justifies entering a particular state, then that state is superfluous.

Informal Axiom 4 (Convexity). Slippage should work to the disadvantage of the trader. Buying more of asset X should make X more expensive, not less.

Otherwise, a runaway effect can occur where traders are motivated to buy more and more of an asset until the AMM's supply is exhausted.

A constant-sum AMM such as

$$ax + by = 1, (1)$$

trades between assets *X* and *Y* at a fixed exchange rate. Constant-sum AMMs fail to satisfy expressivity: as long as the exchange rate matches the market rate, a constant-sum AMM trades without slippage, but as soon as the market rate departs from the AMM's exchange rate, arbitrageurs will exhaust the AMM's supply of the undervalued asset, to the detriment of the liquidity providers. For this reason, with very few exceptions [4], constant-sum AMMs are not used in practice. (Note, however, that the continuity axiom implies that any AMM's behavior *approximates* the behavior of a constant-sum market maker when trades are sufficiently small.) Henceforth, except when explicitly noted, we use "AMM" as shorthand for "AMM that satisfies these axioms".

3.2 AMMs and Valuations

Formally, an *n*-dimensional AMM is given by a function $A: \mathbb{R}^n_{>0} \to \mathbb{R}$ such that:

- For all $c \ge 0$, the *upper contour set* $A(x) \ge c$ is closed and strictly convex.
- A(x) is strictly increasing in each coordinate, and
- *A* is twice-differentiable.

We adopt the convention that the AMM's state space is the level set A(x) = 0, though sometimes we replace 0 with another constant. For brevity, when there is no danger of confusion, we use A to refer to the AMM's function, its state space, and the AMM itself. We use upper(A) for A's upper contour set at 0. We sometimes define an AMM by saying A := F(x) = 0 to mean the AMM A's states lie on the curve F(x) = 0

We remark that restricting the domain of an AMM's function to all-positive coordinates is a "without loss of generality" convention, since an AMM's trading behavior does not change if its function is translated by a constant vector.

It is often convenient to express an AMM in an alternative form. The implicit function theorem [9] implies that for any point on the manifold, all but one coordinate can be chosen freely, and the remaining coordinate is a twice-differentiable convex function f_i of the rest:

$$A(x_1, \ldots, x_{i-1}, f_i(x_1, \ldots, x_{i-1}, x_{i+1}, \ldots, x_n), x_{i+1}, \ldots, x_n) = 0.$$

An opinion on the relative values of assets X_1, \ldots, X_n is captured by a *valuation* $v = (v_1, \ldots, v_n)$, where $0 < v_i < 1$ and $\sum_i v_i = 1^1$. A trader who moves an AMM from state x to state x' makes a profit if the dot product $v \cdot (x - x')$ is positive, and otherwise incurs a loss. The current *market value* is a valuation accepted by most participants.

The standard simplex $\Delta^n \in \mathbb{R}^n_{>0}$ is the set of points (v_1, \ldots, v_n) where for $i = 1, \ldots, n, 0 < v_i < 1$, and $\sum_{i=1}^n v_i = 1$. Each valuation forms the barycentric coordinates of a point in the interior of the standard n-simplex²: $int(\Delta^n)$.

A *stable point* for an AMM A and valuation v is a point $x \in A$ that minimizes the dot product $v \cdot x$. If v is the market valuation, then any trader can make an arbitrage profit by moving the AMM from any state to a stable point, and no trader can make a profit by moving the AMM out of a stable point.

Of course, this model is idealized in several ways. Asset pools are not continuous variables: they assume discrete values. Computation is not infinite-precision: round-off errors and numerical instability are concerns. Popular AMM such as Uniswap V2 [2] and V3 [3] perform trades under a more complicated and dynamic model than the one considered here. Nevertheless, the problem of AMM composition remains largely unaddressed, at least in a formal way, and we believe our model and definitions capture enough of the essential properties of AMMs to make useful progress.

We assume that the functions defining an AMM do not change over time. In practice, an AMM's defining function can change over time. For example, AMMs charge fees in a variety of ways, usually adding those fees to the assets managed by the AMM. Nevertheless, an AMM's defining function does not usually change in the course of a single transaction, the duration for which AMM composition is meaningful. Extending our analysis to time-varying AMMs is the subject of future work.

3.3 Formal Axioms

We are now able to restate the common-sense axioms of Section 3.1 in more precise terms.

AXIOM 1 (CONTINUITY). For every AMM A, the function $A: \mathbb{R}^n_{>0} \to \mathbb{R}$ is twice-differentiable.

As far as we know, existing AMMs use smooth (infinitely differentiable) functions.

AXIOM 2 (CONVEXITY). For every AMM A, upper(A) is strictly convex.

The following is a standard result from convex analysis.

LEMMA 1. A function $f: \mathbb{R}^n_{>0} \to \mathbb{R}_{>0}$ is a strictly convex function if and only if its epigraph $epi(f) = \{(x, a) \in \mathbb{R}^n_{>0} \times \mathbb{R}_{>0} : f(x) \leq a\}$ is a strictly convex set.

AXIOM 3 (Expressivity). Every valuation has a unique stable point in A.

 $^{^1\}mathrm{For}$ simplicity, we assume valuations never assign an asset relative value 0 or 1.

²This notation is non-standard but convenient; others define Δ^n to be a subset of $\mathbb{R}^{n+1}_{>0}$.

LEMMA 2. For any AMM A, every valuation $v \in int(\Delta^n)$ has a stable point in upper(A).

PROOF. Pick any x in upper(A). The set

$$S = \left\{ \boldsymbol{x'} \in \mathbb{R}_{\geq 0}^{n} : \boldsymbol{v} \cdot \boldsymbol{x'} \leq \boldsymbol{v} \cdot \boldsymbol{x} \right\}$$

is compact. Since upper(A) is closed, $\tilde{S} = S \cap upper(A)$ is compact. A stable point solves the optimization problem

$$\min_{\mathbf{x}'\in\tilde{S}}\mathbf{v}\cdot\mathbf{x}'$$

This minimum exists since $\mathbf{v} \cdot \mathbf{x}'$ is a continuous function on the compact set \tilde{S} .

LEMMA 3. For any AMM A and any valuation v, the stable point for v in upper(A) is unique.

PROOF. Fix valuation $\mathbf{v} \in int(\Delta^n)$ and let $\mathbf{x}, \mathbf{x'} \in upper(A)$ where $\mathbf{x} \neq \mathbf{x'}$ and $\mathbf{w} = \mathbf{v} \cdot \mathbf{x} = \mathbf{v} \cdot \mathbf{x'}$. Because upper(A) is strictly convex, for all $t \in (0,1)$, $\tilde{\mathbf{x}} = t\mathbf{x} + (1-t)\mathbf{x'}$ is in int(upper(A)). Since int(upper(A)) is open, there is some $\epsilon > 0$ such that the open ϵ -ball $B_{\tilde{\mathbf{x}}}(\epsilon) \subset int(upper(A))$. Now choose \mathbf{x}^* in $B_{\tilde{\mathbf{x}}}(\epsilon)$ such that $\mathbf{x}^* < \tilde{\mathbf{x}}$. Then we have

$$v \cdot x^* < v \cdot \tilde{x}$$

$$= tv \cdot x + (1 - t)v \cdot x'$$

$$= tw + (1 - t)w = w,$$

a contradiction.

For example, for the 2-dimensional constant-product AMM given by (x,c/x), the valuation (v,1-v) has the unique stable point

$$\left(\sqrt{\frac{c(1-v)}{v}}, \sqrt{\frac{v}{c(1-v)}}\right).$$

More generally, for the 2-dimensional constant-product AMM given by (x, f(x)), the valuation (v, 1 - v) has the unique stable point

$$\left(f'^{-1}\left(-\frac{v}{1-v}\right), f\left(f'^{-1}\left(-\frac{v}{1-v}\right)\right)\right).$$

Lemma 4. For all valuations v, the stable point for v in upper(A) lies on the level set A.

PROOF. Fix valuation v and let x^* be its stable point. Suppose that $x^* \notin A$ but in int(upper(A)). As in Lemma 3, we can find $\epsilon > 0$ and an open ball $B_{x^*}(\epsilon)$ in int(upper(A)). Choosing $x \in B_{x^*}(\epsilon)$ such that $x < x^*$, we have $v \cdot x < v \cdot x^*$, a contradiction.

COROLLARY 1. Every AMM satisfies expressivity.

AXIOM 4 (STABILITY). Every $x \in A$ is the stable point for some valuation.

For example, for the 2-dimensional constant-product AMM (x,c/x), the point (x,c/x) is the stable point for the valuation $(\frac{c}{c+x^2},1-\frac{c}{c+x^2})$. More generally, for the 2-dimensional constant-product AMM (x,f(x)), the point (x,f(x)) is the stable point for the valuation $(\frac{f'(x)}{f'(x)-1},\frac{1}{1-f'(x)})$.

LEMMA 5. For every $x \in A$, there is some $w \in \mathbb{R}^n_{<0}$ such that $w \cdot x > w \cdot y$ for all $y \in upper(A), y \neq x$.

PROOF. Pick $x \in A$. Since upper(A) is strictly convex, the supporting hyperplane theorem implies there is a non-zero $w \in \mathbb{R}^n$ such that $w \cdot x > w \cdot y$ for all $y \in upper(A), y \neq x$. We need to show that $w \in \mathbb{R}^n_{<0}$. Say $w \in \mathbb{R}^n_{>0}$. Choose any $\epsilon > 0$ and let $\tilde{x} = x + \epsilon w$ so that $w \cdot \tilde{x} = w \cdot x + \epsilon ||w||^2 > w \cdot x$. By monotonicity, $\tilde{x} \in upper(A)$, a contradiction. Now consider the case where $w \in \mathbb{R}^n \setminus (\mathbb{R}^n_{<0} \cup \mathbb{R}^n_{>0})$, namely w cannot have all strictly positive or all strictly negative entries. We construct \tilde{w} orthogonal to w. Replace all of the non-negative entries of w in \tilde{w} with the sum of the absolute values of the negative coordinates. Replace all the negative entries of w in \tilde{w} with the sum of all of the non-negative entries of w. These replacements guarantee that all coordinates of \tilde{w} are positive, and $\tilde{w} \cdot w = 0$. Pick $\epsilon > 0$ and let $\tilde{x} = x + \epsilon \tilde{w}$ so

$$w \cdot \tilde{x} = w \cdot x + \epsilon w \cdot \tilde{w}$$
$$= w \cdot x$$

and yet $\tilde{x} \in upper(A)$ since A is strictly increasing. By contradiction, $\mathbf{w} \in \mathbb{R}^n_{<0}$.

LEMMA 6. For every $x \in A$, there exists a valuation v for which x is a stable point.

PROOF. Let A be an n-dimensional AMM and let $x \in A$. Choose $\mathbf{w} \in \mathbb{R}^n_{<0}$ as described in the previous lemma for x. We have $\mathbf{w} \cdot \mathbf{x} > \mathbf{w} \cdot \mathbf{y}$ for $\mathbf{y} \in A$, $\mathbf{y} \neq \mathbf{x}$. Negating \mathbf{w} and re-scaling the result so the elements sum to 1 yields $\mathbf{v} \in int(\Delta^n)$. Thus we have $\mathbf{v} \cdot \mathbf{x} < \mathbf{v} \cdot \mathbf{y}$ for all $\mathbf{y} \in A$ where $\mathbf{y} \neq \mathbf{x}$, implying \mathbf{x} is a stable point for \mathbf{v} .

COROLLARY 2. Every AMM satisfies stability.

In short, the goal of this paper is to balance axioms and composition operators so that the class of AMMs satisfying these axioms remains closed under these composition operators.

4 TOPOLOGICAL EQUIVALENCE

How many truly distinct AMMs of a given dimension are there? Considered as a mathematical object, much of an AMM's structure is captured by the link between valuations and their stable points. We say that two AMMs are topologically equivalent if there is a stable-point preserving homeomorphism between their manifolds. By itself, a homeomorphism between manifolds conveys little information,

but a homeomorphism that preserves stable points preserves the AMMs' common underlying structure.

In this section we show that *all* AMMs over the same set of assets, if they satisfy our axioms, are topologically equivalent. More precisely, for any two AMMs $A(x_1, ..., x_n)$ and $B(x_1, ..., x_n)$ over asset types $X_1, ..., X_n$, satisfying our axioms, there is a homeomorphism $\mu: A \to B$ such that x and $\mu(x)$ are the stable states for the same valuation. This proof relies on the uniqueness of stable points: for example it would not hold if AMM functions were convex instead of strictly convex.

Although topological equivalence implies a common mathematical structure, two topologically equivalent AMMs may differ substantially with respect to price slippage, fees, or how expensive it is to move from one valuation's stable state to another's.

Recall from Lemma 3 and Lemma 6 that there is a unique function $\phi: int(\Delta^n) \to A$ carrying each valuation to its unique stable point.

LEMMA 7. For A, an AMM, the stable point map

$$\phi: int(\Delta^n) \to A$$

is a continuous bijection.

PROOF. The map ϕ is surjective by Lemma 6, and injective by Lemma 3. To show continuity, consider the sequence $\{v_n\}_{n=1}^{\infty} \subset int(\Delta^n)$ where $\lim_{n\to\infty} v_n = v$. Let $\tilde{x} = \lim_{n\to\infty} \phi(v_n)$ and $x^* = \phi(v)$. Suppose $\tilde{x} \neq x^*$. Note that $v \cdot x^* < v \cdot \tilde{x}$ by definition of stable point. Letting $\overline{x} = \frac{x^* + \tilde{x}}{2}$ by strict convexity we know $\overline{x} \in int(upper(A))$. We also have that $v \cdot x^* < v \cdot \overline{x} < v \cdot \tilde{x}$. Notice now that $v_n \cdot \overline{x} > v_n \cdot \phi(v_n)$ by definition so taking limits we get $v \cdot \overline{x} > v \cdot \tilde{x}$, a contradiction. Thus $\lim_{n\to\infty} \phi(v_n) = \phi(v)$.

LEMMA 8. For A, an AMM, the stable point map

$$\phi: int(\Delta^n) \to A$$

is a homeomorphism.

PROOF. From Lemma 7, ϕ is both bijective and continuous, so it is enough to show ϕ^{-1} is continuous. For any v with stable point x, the first-order conditions imply that $v = \lambda \nabla A(x)$ for some non-zero (Lagrange multiplier) $\lambda \in \mathbb{R}$. Since A is strictly increasing, $\lambda > 0$. Thus we can think of v as a function of x, written $v(x) = \lambda(x)\nabla A(x)$. Because A is continuously differentiable, $\nabla A(x)$ is continuous, so it is enough to check $\lambda(x)$ is continuous. Because v(x) is a convex combination, and v(x) is unique,

$$\lambda(\mathbf{x}) = \frac{1}{\|\nabla A(\mathbf{x})\|_1},$$

which is continuous because each $\frac{\partial A(x)}{\partial x_i}$ is continuous. It follows that ϕ^{-1} is continuous.

Theorem 3. Let A and B be AMMs over the same set of assets. There is a homeomorphism $\mu: A \to B$ that preserves stable points: if x is the stable point for valuation v in A, then $\mu(x)$ is the stable point for v in B.

PROOF. By Lemma 8, there exist homeomorphisms

$$\phi : int(\Delta^n) \to A,$$

 $\phi' : int(\Delta^n) \to B$

Their composition $\mu = \phi' \circ \phi^{-1}$ is also a homeomorphism. For $v \in int(\Delta^n)$ with stable points $x \in A, x' \in B, \phi(v) = x$ and $\phi'(v) = x'$, so $\mu(x) = \phi'(\phi^{-1}(x)) = \phi'(v) = x'$, implying that μ preserves stable points.

Note that this result requires that *A* have continuous first derivatives.

5 OPERATORS

It is useful to be able to reduce an AMM's dimension, perhaps by ignoring some assets, or by creating "baskets" of distinct assets that can be treated as a unit. In this section we introduce two tools for reducing dimensionality: projection, and asset virtualization.

5.1 Projection

An AMM may provide the ability to trade across a variety of asset types, but traders may choose to restrict their attention to a subset, ignoring the rest. Perhaps the ignored assets are too volatile, or not volatile enough, or there are regulatory barriers to owning them.

Mathematically, the *projection* operator acts on an AMM by fixing some state coordinates to constant values and letting the rest vary. We will show that projecting an AMM in this way yields another AMM of lower dimension. Informally, traders are free to ignore uninteresting assets.

Definition 4. Let $\mathbf{x} = (x_1, \dots, x_n)$, $\mathbf{y} = (y_1, \dots, y_m)$, and a constant $\mathbf{a} = (a_1, \dots, a_n)$. The projection of A onto \mathbf{a} is given by $A_{\mathbf{a}}(\mathbf{y}) = A(\mathbf{a}, \mathbf{y}) = 0$

LEMMA 9. Given an (n+m)-dimensional AMM A(x, y) = 0 and $a \in \mathbb{R}^n_{>0}$, the projection $A_a(y)$ is an m-dimensional AMM.

PROOF. It is enough to check that A_a is twice-differentiable, strictly increasing, and upper(A) is strictly convex. Because A(x, y) is twice-differentiable, so is $A(a, y) = A_a(y)$. To show that A_a is strictly increasing, let $x' \geq x$.

$$A_{a}(x) = A(a, x)$$

$$< A(a, x')$$

$$= A_{a}(x').$$

To show that $upper(A_a)$ is strictly convex, pick distinct x and x' in $upper(A_a)$. Namely $A(a,x) = A_a(x) \ge 0$ and

$$A(a, x') = A_a(x') \ge 0$$
. For $t \in (0, 1)$:
 $A_a(tx + (1 - t)x') = A(ta + (1 - t)a, tx + (1 - t)x')$
 $= A(t(a, x) + (1 - t)(a, x')) > 0$

by the strict convexity of upper(A).

Lemma 10. For index set I, let $v=(v_i|i\in I)$ be a valuation for a sequence of asset types $X=(X_i|i\in I)$. For $J\subset I$, $v'=(\frac{v_j}{1-\sum_{k\in I}v_k}|j\in J)$ is a valuation for $X'=(X_j|j\in J)\subset X$.

We say that X' inherits v' from valuation v of X.

The next lemma states that stable points persist under projection.

LEMMA 11. Let $A(\mathbf{x}, \mathbf{y}) = 0$ be an (n + m)-dimensional AMM, $\mathbf{v} \in int(\Delta^{n+m})$ a valuation on (\mathbf{x}, \mathbf{y}) . If (\mathbf{a}, \mathbf{b}) is the stable point for \mathbf{v} in $A(\mathbf{x}, \mathbf{y})$ then \mathbf{b} is the stable point for the inherited valuation \mathbf{v}' .

PROOF. Suppose the stable point for v' is $b' \neq b$, namely $v' \cdot b' < v' \cdot b$. Scaling both sides by $1 - \sum_{i=1}^{n} v_i$ yields $(v_{n+1}, \ldots, v_{n+m}) \cdot b' \leq (v_{n+1}, \ldots, v_{n+m}) \cdot b$. Adding $(v_1, \ldots, v_n) \cdot a$ to both sides yields $v \cdot (a, b') \leq v \cdot (a, b)$, contradicting the assumption that (a, b) is the stable point for v.

5.2 Virtualization

It is sometimes convenient to create a "virtual asset" from a linear combination of assets. Here we show that replacing a set of assets traded by an AMM with a single virtual asset is also an AMM. This construction works for any linear combination, although the most sensible combination is usually the assets' current market valuation.

Here is a simple example of asset virtualization. Consider an AMM that trades across three asset types, X, Y, Z, defined by the constant-product formula

$$A(x, y, z) = xyz - 8 = 0,$$

initialized in state (2,2,2). A trader believes that 2 units of Y are always worth 1 unit of Z, and that it makes sense to link them in that ratio by creating a virtual asset W worth 2/3 units of Y and 1/3 unit of Z, and to trade in a single denomination of W instead of individual denominations of Y and Z.

Formally, the trader defines W in terms of the valuation $v = (\frac{2}{3}, \frac{1}{3})$ on Y, Z. The virtualized AMM A|v is defined by

$$(A|v)(x, w) = A(x, \frac{2w}{3}, \frac{w}{3} + 1)$$
$$= x\frac{2w}{3}(\frac{w}{3} + 1) - 8$$
$$= 0$$

with initial state (2,3). (The "+1" in the Z co-ordinate appears because 2 Y and 2 Z units are not evenly divisible into W units.)

Let $\mathbf{x} = (x_1, \dots, x_n)$ and $\mathbf{y} = (y_1, \dots, y_m)$. Let $A(\mathbf{x}, \mathbf{y}) := A(x_1, \dots, x_n, y_1, \dots, y_m) = 0$ be an (n + m)-dimensional AMM with initial state $(\mathbf{a}, \mathbf{b}) = (a_1, \dots, a_n, b_1, \dots, b_m)$. Let us create a virtual asset Z from y_1, \dots, y_m , using the valuation $\mathbf{v} = (v_1, \dots, v_m)$.

Let $c \in \mathbb{R}_{>0}$ be the largest value such that $b - cv \ge 0$. The value c is the number of Z assets in b, and r = b - cv is the vector of residues if b is not evenly divisible into Z units. The virtualized AMM is given by

$$(A|v)(x,z) = A(x, v_1z + r_1, \dots, v_mz + r_m)$$

= 0

with initial state (a_1, \ldots, a_n, c) .

The next lemma says that in any AMM state, it is always possible to virtualize any set of assets.

LEMMA 12. Let A be an (m+n)-dimensional AMM in state (a,b), where $a \in \mathbb{R}^m_{>0}$, $b \in \mathbb{R}^n_{>0}$, and valuation $v \in int(\Delta^n)$. We claim that for any $a' \in \mathbb{R}^m_{>0}$, there is a unique $t \in \mathbb{R}$ such that (a',b+tv) is a state of A.

PROOF. We seek $t \in \mathbb{R}$ such that A(a', b + tv) = 0. There are several cases. If A(a', b) = 0, then t = 0 and we are done. Suppose A(a', b) < 0. If A(a', b + v) = 0, then t = 1 and we are done. If A(a', b + v) < 0, pick a vector $\mathbf{c} = (c_1, \dots, c_n) \in \mathbb{R}^n_{>0}$ such that A(a', b + c) = 0. Let $\epsilon > 0$,

$$s_i = \frac{c_i + \epsilon}{v_i}, 0 \le i \le n,$$

and $s = \max_{0 \le i \le n} s_i$. It follows that $sv \ge c + \epsilon 1$, and $b + sv \ge b + c + \epsilon 1$. Since $A_{a'}$ is strictly increasing, $A(a', b + sv) \ge A_{a'}(b + c) = 0$. Define $\alpha(t) : [0, 1] \to \mathbb{R}$ by $\alpha(t) = A_{a'}(b + v + t(s - 1)v)$. Because α is continuous, the intermediate value theorem guarantees a unique $t^* \in (0, 1)$ such that $\alpha(t^*) = 0$. Taking $t = (1 + t^*(s - 1))$ establishes the claim. If A(a', b + v) > 0, let

$$s_i = \frac{c_i - \epsilon}{v_i}, 0 \le i \le n,$$

and $s = \min_{0 \le i \le n} s_i$, and the claim follows from a symmetric argument.

Suppose A(a',b) > 0. Pick a vector $\mathbf{c} = (c_1, \dots, c_n) \in \mathbb{R}^n_{>0}$ such that $A(a', b - \mathbf{c}) = 0$. Let $\epsilon > 0$, $s_i = \frac{c_i + \epsilon}{v_i}$, $0 \le i \le n$. and $s = \max_{0 \le i \le n} s_i$. It follows that $s\mathbf{v} \ge \mathbf{c} + \epsilon 1$, and $\mathbf{b} - s\mathbf{v} \le \mathbf{b} - \mathbf{c} - \epsilon 1$, so $A(a', \mathbf{b} - s\mathbf{v}) \le A(a', \mathbf{b} - \mathbf{c}) = 0$. Let $\alpha(t) : [0, 1] \to \mathbb{R}$ be $\alpha(t) = A(a', \mathbf{b} + (t - 1)s\mathbf{v})$. As before, the intermediate value theorem guarantees a unique $t^* \in (0, 1)$ such that $\alpha(t^*) = 0$. Taking $t = (1 - t^*)s$ establishes the claim.

THEOREM 5. Given an (n+m)-dimensional AMM A(x, y) = 0, and a valuation $v \in int(\Delta^m)$, the virtualized (A|v)(x, z) is an (n+1)-dimensional AMM.

PROOF. It is enough to check that A|v is twice-differentiable, strictly increasing, and upper(A|v) is strictly convex. (A|v) is twice-differentiable because A is twice-differentiable.

To show that A|v is strictly increasing, let $x' \ge x$ and $z' \ge z$.

$$(A|v)(x_1,...,x_n,z)$$

$$= A(x_1,...,x_n,v_1z + r_1,...,v_mz + r_m)$$

$$< A(x'_1,...,x'_n,v_1z' + r_1,...,v_mz' + r_m)$$

$$= (A|v)(x'_1,...,x'_n,z').$$

To show that upper(A|v) is strictly convex, pick distinct (x, z) and (x', z') on the manifold:

$$(A|v)(x,z) = (A|v)(x',z') = 0.$$

For $t \in (0, 1)$,

$$(A|v)(tx + (1-t)x', tz + (1-t)z')$$

$$= A(tx + (1-t)x', v(tz + (1-t)z') + r)$$

$$= A(tx + (1-t)x', t(vz + r) + (1-t)(vz' + r))$$

$$= A(t(x, vz + r) + (1-t)(x', vz' + r)) > 0$$

by the strict convexity of upper(A).

Stable points are well-behaved under virtualization. Let A(x, y) = 0 be an (n + m)-dimensional AMM and $v \in int(\Delta^n)$ a valuation. If A is an (n + m)-dimensional AMM in state (a, b), and $v \in int(\Delta^n)$ a valuation, then (A|v)(a, c) = A(a, cv + r) = A(a, b) = 0. For any state in the virtualized AMM, (A|v)(x, t) = A(x, b + (t-c)v) = 0. This expression depends on b and v, where c is a constant determined by b and v. Since (A|v)(x, t) is an (n + 1)-dimensional AMM, we can write t = f(x) for some $f: \mathbb{R}^n_{>0} \to \mathbb{R}$. The virtualized AMM can be expressed as (x, f(x)) where A(x, b + (f(x) - c)v) = 0.

Lemma 13. If (a^*, b^*) is the stable point on AMM A(x, y) for valuation (v, w), then $(a^*, f(a^*))$ is the stable point on the virtualized AMM (A|w)(x, t) for the valuation $(v, ||w||_2^2)/||(v, ||w||_2^2)||_1$.

PROOF. Suppose $(a^*, f(a^*))$ is not a stable point for $(v, \|w\|_2^2)$: there is a distinct point $(a, f(a)) \in A|w$ where $v \cdot a + \|w\|_2^2 f(a) < v \cdot a^* + \|w\|_2^2 f(a^*)$. Now define $b = b^* + w(f(a) - f(a^*))$, which by the virtualization construction we have $(a, b) \in A$. Then

$$v \cdot a + w \cdot b = v \cdot a + w \cdot b^* + w \cdot w (f(a) - f(a^*))$$

$$= v \cdot a + ||w||_2^2 f(a) - ||w||_2^2 f(a^*) + w \cdot b^*$$

$$< v \cdot a^* + ||w||_2^2 f(a^*) - ||w||_2^2 f(a^*) + w \cdot b^*$$

$$= v \cdot a^* + w \cdot b^*$$

This is a contradiction since (a^*, b^*) is the stable point for (v, w).

6 SEQUENTIAL COMPOSITION

AMMs are intended to be composed. A Uniswap AMM typically converts between an ERC-20 token and ether cryptocurrency. To convert, say, florin tokens to guilder tokens, one would first convert florins to ether, then ether to guilders. Bancor uses a proprietary BNT token for the same purpose. Some form of composition seems to be essential to making AMMs useful, but we will see that while there are many ways in which AMMs might be composed, not all of them make sense. The most basic property one would demand is closure under composition: the result of composing two AMMs should itself be an AMM.

Being closed under composition should not be taken for granted. For example, consider two constant-product AMMs: A := (x, 1/x), initialized in state (a, 1/a), and B := (y, 1/y), initialized in state (b, 1/b). Their composition $A \otimes B = (x, h(x))$, where

$$h(x) = \frac{1}{b + \frac{1}{a} - \frac{1}{x}} = \frac{ax}{x - a + abx}.$$
 (2)

The set of constant-product AMMs is thus not closed under composition.

6.1 One-to-One Composition

We first consider the result of composing 2-dimensional AMMs, that is, AMMs that trade between two asset types.

Let A = (x, f(x)), initialized to (a, f(a)), and B = (y, g(y)), initialized to (b, g(b)). A trades between asset types X and Y, and B between Y and Z. Their composition, initialized to (a, g(b)), trades between X and Z.

Operationally, composition is defined as follows.

- Move *A* from state (a, f(a)) to state (x, f(x)), yielding profit-loss vector (a x, f(a) f(x)).
- Add f(a) f(x) to the Y balance of B, yielding new state (b + f(a) f(x), g(b + f(a) f(x))).

This trade takes the composition from (a, g(b)) to (x, g(b + f(a) - f(x))). Let h(x) = g(b + f(a) - f(x)). The composition $A \otimes B$ is given in the form (x, h(x)). Because f, g are twice-differentiable:

LEMMA 14. $(A \otimes B)(x, y)$ is twice-differentiable.

Lemma 15. $(A \otimes B)(x, y)$ is strictly increasing.

PROOF. We show that if $(x', y') \ge (x, y)$, meaning at least one coordinate is strictly greater, then $(A \otimes B)(x', y') > (A \otimes B)(x, y)$. Recall that f and g are strictly decreasing by hypothesis. There are two cases. First, suppose x' > x and

 $y' \geq y$.

$$x' > x$$

$$f(x') < f(x)$$

$$b + f(a) - f(x') > b + f(a) - f(x)$$

$$g(b + f(a) - f(x')) < g(b + f(a) - f(x))$$

$$y - g(b + f(a) - f(x')) > y - g(b + f(a) - f(x))$$

$$y' - g(b + f(a) - f(x')) > y - g(b + f(a) - f(x))$$

$$(A \otimes B)(x, y') > (A \otimes B)(x, y)$$

The second case, where $x' \ge x$ and y' > y is similar. \square

Lemma 16. The upper contour set $upper(A \otimes B)$ is strictly convex.

PROOF. Since $upper(A \otimes B) = epi(h)$, by Lemma 1, it is enough to check the strict convexity of h. Pick two different x and x'. Recall that for $t \in (0,1)$, f(tx+(1-t)x') < tf(x)+(1-t)f(x'), and similarly for g(y) by hypothesis. For $t \in (0,1)$,

$$f((1-t)x + tx') < (1-t)f(x) + tf(x')$$

$$b + f(a) - f((1 - t)x + tx')$$

$$> (1 - t)(b + f(a) - f(x))$$

$$+ t(b + f(a) - f(x'))$$

$$g(b+f(a) - f((1-t)x + tx'))$$

$$< (1-t)g(b+f(ag) - f(x))$$

$$+ tg(b+f(a) - f(x')))$$

$$h((1-t)x + tx) < (1-t)h(x) + th(x')$$

which proves the claim.

The next lemma relates stability and sequential composition for 2-dimensional AMMs.

LEMMA 17. Let (v, w, v') be a valuation for assets X, Y, Z. If (a, f(a)) is the stable point for AMM A := (x, f(x)) for valuation $(v, w)/\|(v, w)\|_1$ and (b, g(b)) the stable point for AMM (y, g(y)) for valuation $(w, v')/\|(w, v')\|_1$, then (a, h(a)) is the stable point for the valuation $(v, v')/\|(v, v')\|_1$.

PROOF. If (a, h(a)) is not the stable point for (v, v'), there is some $x \neq a$ such that (x, h(x)) is the stable point. By assumption, for $x \neq a$ and $y \neq b$,

$$va + wf(a) < vx + wf(x)$$

$$wb + v'q(b) < wy + v'q(y)$$

Also by assumption,

$$vx + v'h(x)$$

$$< va + v'h(a)$$

$$= va + v'g(b + f(a) - f(a))$$

$$= va + w'g(b)$$

$$= va + wf(a) - wf(a) + wb - wb + v'g(b)$$

$$< vx + wf(x) - wf(a) + wy + v'g(y) - wb$$

$$= vx + w(f(x) - f(a)) + w(y - b) + v'g(y)$$

$$= vx + w(f(x) - f(a)) + w(f(a) - f(x))$$

$$+ v'g(b + f(a) - f(x))$$

$$= vx + v'h(x),$$

a contradiction. (The last step follows by taking $y = b + f(a) - f(x) \neq b$, possible since f strictly decreasing). \Box

The converse is false, as the following example shows: Consider two constant product AMMs $A := (r, \frac{1}{2})$ as

Consider two constant product AMMs $A:=(x,\frac{1}{x})$ and $B:=(y,\frac{1}{y})$, both initially in state (1,1). The composed AMM is given by $A\otimes B$ $(x,\frac{x}{2x-1})$ with state (1,1). Now if $(v,w,v')=(\frac{1}{4},\frac{1}{2},\frac{1}{4})$, (1,1) is the stable point of $A\otimes B$ with respect to (v,w,v'). However, the stable point for (v,w) on A is $(\sqrt{2},\frac{\sqrt{2}}{2})$ and (w,v') on B is $(\frac{\sqrt{2}}{2},\sqrt{2})$.

6.2 Many-to-One Composition

AMM *A* trades asset types $X_1, ..., X_m, Z$, with initial state (a, f(a)), where $a = (a_1, ..., a_m)$, and f(a) is the explicit function defining the *Z* coordinate in terms of the others.

AMM *B* trades asset types $Z, Y_1, ..., Y_n$, with initial state (c, b, g(c, b)), where $b = (b_1, ..., b_{n-1})$, and g(c, b) is the function defining the Y_n coordinate in terms of the others.

The Z asset flows between A and B but is not directly accessible to traders. The composition $A \otimes B$ trades asset types $X_1, \ldots, X_m, Y_1, \ldots, Y_n$, with initial state (a, b, h(a, b)), for h to be defined.

Operationally, the composition works as follows. The trader changes each a_i to x_i , $0 \le i \le m$, and each b_i to y_i , $0 \le i \le n-1$. Let $\mathbf{x} = (x_1, \dots, x_m)$ and $\mathbf{y} = (y_1, \dots, y_{n-1})$. The new state of A is $(\mathbf{x}, f(\mathbf{x}))$. The amount of Z that flows from A to B is $f(\mathbf{a}) - f(\mathbf{x})$. The new state of B is $(c + f(\mathbf{a}) - f(\mathbf{x}), \mathbf{y}, g(c + f(\mathbf{a}) - f(\mathbf{x}))$. The new state of $A \otimes B$ is $(\mathbf{x}, \mathbf{y}, g(c + f(\mathbf{a}) - f(\mathbf{x})))$

Define h(x, y) = g(c + f(a) - f(x), y). $(A \otimes B)(x, y, z) = z - h(x, y)$.

LEMMA 18. $(A \otimes B)(x, y, z)$ is twice-differentiable.

PROOF. Immediate because f, g are twice-differentiable by hypothesis.

LEMMA 19. $(A \otimes B)(x, y, z)$ is strictly increasing.

PROOF. We show that if $(x', y', z') \not\ge (x, y, z)$, then $(A \otimes B)(x', y', z') > (A \otimes B)(x, y, z)$. There are two cases. First, suppose $(x', y') \not\ge (x, y)$ and $z' \ge z$.

$$(x',y') \geq (x,y)$$

$$f(x',y') < f(x,y)$$

$$b + f(a) - f(x',y') > b + f(a) - f(x,y)$$

$$g(b + f(a) - f(x',y')) < g(b + f(a) - f(x,y))$$

$$z - g(b + f(a) - f(x',y')) > z - g(b + f(a) - f(x,y))$$

$$z' - g(b + f(a) - f(x',y')) > z - g(b + f(a) - f(x,y))$$

$$(A \otimes B)(x',y',z') > (A \otimes B)(x,y,z)$$

The second case, where $x' \ge x$ and z' > z is similar. \Box

Lemma 20. $upper(A \otimes B)$ is strictly convex.

PROOF. Again it is enough to show h is strictly convex. Pick two different points (x, y) and (x', y'). Recall that f and g are strictly convex by hypothesis. For $t \in (0, 1)$,

$$f((1-t)(x,y) + t(x',y'))$$

< $(1-t)f(x,y) + tf(x',y')$

$$b + f(a) - f((1 - t)(\mathbf{x}, \mathbf{y}) + t(\mathbf{x}', \mathbf{y}'))$$

$$> (1 - t)(b + f(a) - f(\mathbf{x}, \mathbf{y}))$$

$$+ t(b + f(a) - f(\mathbf{x}', \mathbf{y}'))$$

$$\begin{split} g(b+f(a)-f((1-t)(\pmb{x},\pmb{y})+t(\pmb{x'},\pmb{y'}))) \\ &< (1-t)g((b+f(a)-f(\pmb{x},\pmb{y})) \\ &+ tg(b+f(a)-f(\pmb{x'},\pmb{y'})) \end{split}$$

$$h((1-t)(\boldsymbol{x},\boldsymbol{y})+t(\boldsymbol{x}',\boldsymbol{y}'))$$

$$<(1-t)h(\boldsymbol{x},\boldsymbol{y})+th(\boldsymbol{x}',\boldsymbol{y}')$$

Here is how stable points behave under sequential composition of multi-dimensional AMMs.

THEOREM 6. Let (v, w, v') be a valuation, $v \in \mathbb{R}^n$, $w \in \mathbb{R}$, and $v' \in \mathbb{R}^m$. If $(a^*, f(a^*))$ is the stable point on AMM (x, f(x)) for valuation $(v, w)/\|(v, w)\|_1$ and $(c, b^*, g(c, b^*))$ is the stable point on AMM (z, y, g(z, y)) for valuation $(w, v')/\|(w, v')\|_1$, then $(a^*, b^*, h(a^*, b^*))$ is the stable point for the valuation $(v, v')/\|(v, v')\|_1$.

PROOF. if $(a^*, b^*, h(a^*, b^*))$ is not a stable point for (v, v'), there is some $(x, y) \neq (a^*, b^*)$ such that (x, y, h(x, y)) is the stable point. We write $v' = (v'_{m-1}, v'_m)$ to separate the first

m-1 components from the m-th component. By assumptionm for $x \neq a^*$ and $(z, y) \neq (c, b^*)$,

$$bv \cdot \boldsymbol{a}^* + wf(\boldsymbol{a}^*) < v \cdot x + wf(x)wc + v'_{m-1} \cdot \boldsymbol{b} + v'_{m}g(c, \boldsymbol{b})$$
$$< wz + v'_{m-1} \cdot \boldsymbol{y} + v'_{m}g(z, \boldsymbol{y})$$

Also by assumption

$$\begin{split} v \cdot x + v'_{n-1} \cdot y + v'_m h(x,y) \\ &< v \cdot a^* + v'_{m-1} \cdot b^* + v'_m h(a^*,b^*) \\ &= v \cdot a^* + w f(a^*) - w f(a^*) + w c - w c + v'_{m-1} \cdot b^* \\ &+ v'_m g(c,b^*) \\ &< v \cdot x + w f(x) - w f(a^*) + w z + \\ &v'_{m-1} \cdot y + v'_m g(z,y) - w c \\ &= v \cdot x + v'_{m-1} \cdot y + v'_m g(z,y) \\ &= v \cdot x + v'_{m-1} \cdot y + v'_m g(c + f(a^*) - f(x),y) \\ &= v \cdot x + v'_{m-1} \cdot y + v'_m h(x,y) \end{split}$$

a contradiction.

6.3 Many-to-Many Composition

What does it mean to compose two AMMs that share multiple assets? We will see that this definition requires some care.

Here is the most obvious definition. Let W, X, Y, Z be asset types. Consider two 3-dimensional AMMs, A defined by wxy = 1 and B defined by xyz = 8. where A is in state (1,1,1) and B in state (2,2,2). We want to compose them into a 2-dimensional AMM $(A \oplus B)$ between W and Z treating X and Y as "hidden" intermediate assets.

A trader adds dx > 0 units of X to $(A \oplus B)$:

- (1) pick any $dy, dz \le 0$ satisfying (1+dx)(1+dy)(1+dz) = 1, yielding a profit-loss vector on Y, Z of (dy, dz).
- (2) Subtract this profit-loss vector from the first two components of B, then solve for $dw \le 0$ so that (2-dy)(2-dz)(2+dw) = 8.

More generally, let A be initialized in (a, b, c), B in (b', c', d). and $A \oplus B$ in (a, d). Then (w, z) is in $A \oplus B$ if there exist $dx, dy \le 0$ (the assets transferred) such that (w, b+dx, c+dy) is in A and (b'-dx, c'-dy, z) is in B.

Here is why this naïve definition of composition is flawed. As before, A(w, x, y) := wxy - 1 = 0 starts in state (1, 1, 1), B(x, y, z) := xyz - 8 = 0 starts in (2, 2, 2). Let dw = 3, $dx = -\frac{1}{2}$, $dy = -\frac{1}{2}$, so dz satisfies:

$$(2-dx)(2-dy)(2+dz) = \frac{25}{4}(2+dz) = 2$$

implying $dz = -\frac{42}{25}$. If instead dw = 3, $dx = -\frac{1}{4}$, $dy = -\frac{2}{3}$, then dz satisfies:

$$(2-dx)(2-du)(2+dz) = 6(2+dz) = 2$$

meaning dz = -5/3. So both $(4, \frac{8}{25})$ and $(4, \frac{1}{3})$ are valid states in the naïve composition, violating the requirement that each coordinate is a function of the others.

We have just seen that AMMs joined by more than one hidden asset type provide too many degrees of freedom to allow composition in a simple, well-defined way. In a practical sense, however, if two AMMs agree on a valuation for the hidden assets, then it makes sense to transfer them in proportion to their agreed-upon relative values. To define composition for AMMs with multiple hidden assets, we create a single virtual asset from a convex combination of the hidden assets (Section 5.2). Although any convex combination would produce a well-behaved composition, it makes sense to use the current market valuation, if one exists. By reducing the number of hidden assets to one, virtualization reduces many-to-many composition to many-to-one composition, as analyzed in Section 6.2.

Our next theorem concerns stable points under this most general type of sequential composition.

THEOREM 7. Let (v, w, v') be a valuation, $v \in \mathbb{R}^n, w \in \mathbb{R}^k, v' \in \mathbb{R}^m$. Let (a^*, b^*) be the stable state for valuation $(v, w)/\|(v, w)\|_1$ on the AMM A(x, y) and (c^*, d^*, e^*) the stable state for $(w, v')/\|(w, v')\|_1$ on the AMM B(z, r, q). If we write A|w as (x, f(x)) and B|w as (z, r, g(z, r)), then $(a^*, d^*, h(a^*, d^*))$, where $e^* = h(a^*, d^*)$ is the stable state on $(A|w) \oplus (B|w)$ for the valuation $(v, v')/\|(v, v')\|_1$.

PROOF. Applying Lemma 13 twice, $(\boldsymbol{a}^*, f(\boldsymbol{a}^*))$ is the stable point for valuation $(\boldsymbol{v}, \|\boldsymbol{w}\|_2^2)/\|(\boldsymbol{v}, \|\boldsymbol{w}\|_2^2)\|_1$ and $(t^*, \boldsymbol{d}^*, g(t^*, \boldsymbol{d}^*))$ is the stable point for valuation $(\|(\boldsymbol{w}\|_2^2, \boldsymbol{v}')/\|(\|(\boldsymbol{w}\|_2^2, \boldsymbol{v}'))\|_1$, where $e^* = g(t^*, \boldsymbol{d}^*)$. Applying Lemma 13, $(\boldsymbol{a}^*, \boldsymbol{d}^*, h(\boldsymbol{a}^*, \boldsymbol{d}^*))$ is stable on $(A|\boldsymbol{w}) \oplus (B|\boldsymbol{w})$ with respect to valuation $(\boldsymbol{v}, \boldsymbol{v}')/\|(\boldsymbol{v}, \boldsymbol{v}')\|_1$.

This result shows that the composition of AMMs at stable points remains stable.

7 PARALLEL COMPOSITION

Parallel composition arises when a trader is faced with multiple AMMs, but wants to treat them them as if they were a single AMM. In sequential composition, the composed AMMs exchange "hidden" assets . In parallel composition, the composed AMMs compete for overlapping assets.

Suppose Alice wants to trade asset X for asset Y. Bob and Carol both offer AMMs to convert from X to Y. Bob's AMM is $B(x,y):=x^2y=\frac{3}{4}$ in state $(1,\frac{3}{4})$, while Carol's AMM is C(x,y):=xy=1 in state (1,1). Alice would like to compose the two AMMs and treat them as one AMM. Bob provides a better initial rate of exchange for small trades, but Carol provides less slippage for large trades. One can check that if Alice converts 1 unit of X, she gets more Y assets from

Bob than from Carol, while if she converts 3 units, she gets more from Carol. (This process is not the same as order-book clearing, because order-book offers are typically expressed in terms of a fixed amount and a fixed price, while parallel AMM offers are expressed in terms of price curves.) This type of composition occurs, for example, when a trader is faced with multiple pools as in Uniswap v3 [3].

A rational Alice will split her assets between B and C to maximize her return. Suppose A(x, y) is in state (a, f(a)) and B(x, y) in (b, g(b)). Alice splits her d assets, transferring td to Bob's AMM and (1 - t)d to Carol's, returning

$$f(a) - f(a+tx) + g(b) - g(b+(1-t)x)$$

units of *Y*. Let h(x) = f(a+tx) + g(b+(1-t)x). Define the *parallel composition* of *B* and *C* with respect to v = (t, 1-t) to be.

$$(B||C)(x,y) := y - h(x) = 0.$$

LEMMA 21. (B||C)(x,y) is a 2-dimensional AMM.

PROOF. (B||C)(x,y) is twice-differentiable because f and g are twice-differentiable. To check that (B||C)(x,y) is strictly increasing, let $x' \ge x$ and $y' \ge y$ where at least one inequality is strict. For the first case, suppose x' > x and $y' \ge y$.

$$f(a+tx') < f(a+tx)$$

$$g(b+(1-t)x') < g(b+(1-t)x)$$

$$f(a+tx') + g(b+(1-t)x') < f(a+tx) + g(b+(1-t)x)$$

$$h(x') < h(x)$$

$$y - h(x') > y - h(x)$$

$$y' - h(x') > y - h(x)$$

The case where $x' \ge x$ and y' > y is similar.

To check that upper((B||C)(x,y)) is strictly convex, we can verify h is strictly convex. Pick distinct x, x'. For $s \in (0,1)$,

$$\begin{split} sf(a+tx) + (1-s)f(a+tx') \\ &> f(s(a+tx) + (1-s)(a+tx')) \\ sg(b+(1-t)x) + (1-s)g(b+(1-t)x') \\ &> g(s(b+(1-t)x) + (1-s)(b+(1-t)x')) \\ s(f(a+tx) + g(b+(1-t)x)) \\ &+ (1-s)(f(a+tx') + g(b+(1-t)x')) \\ &> f(s(a+tx) + (1-s)(a+tx')) \\ &+ g(s(b+(1-t)x) + (1-s)(b+(1-t)x')) \end{split}$$

sh(x) + (1-s)h(x') > h(sx + (1-s)x')

П

which establishes the claim.

LEMMA 22. Let A := (x, f(x)), B := (y, g(y)) be two AMMs trading assets X and Y, such that (a, f(a)) and (b, g(b)) are their respective stable points for the valuation (v, 1 - v). If $h_t(x) = f(a + tx) + g(b + (1 - t)x)$, then $(0, h_t(0))$ is stable point on $A \parallel B$ with respect to (v, 1 - v) for all $t \in \mathbb{R}$.

PROOF. By assumption we have

$$va + (1 - v)f(a) < v(a + tx) + (1 - v)f(a + tx)$$
$$(1 - v)f(a) < tvx + (1 - v)f(a + tx)$$

and

$$vb + (1 - v)g(b) < v(b + (1 - t)x) + (1 - v)g(b + (1 - t)x)$$
$$(1 - v)g(b) < (1 - t)vx + (1 - v)g(b + (1 - t)x),$$

yielding

$$v0 + (1 - v)h_t(0) = (1 - v)h_t(0)$$

$$= (1 - v)f(a) + (1 - v)g(b)$$

$$< tvx + (1 - v)f(a + tx) + (1 - t)vx$$

$$+ (1 - v)g(b + (1 - t)x)$$

$$= vx + (1 - v)(f(a + tx) + g(b + (1 - t)x))$$

$$= vx + (1 - v)h(x)$$

so
$$(0, h_t(0))$$
 is a stable point for $(v, (1-v))$.

Parallel composition is well-defined for any valuation, but what valuation should a rational Alice pick? Differentiating with respect to t yields

$$0 = -xf'(a+tx) + xg'(b+(1-t)x)$$

$$xf'(a+tx) = xg'(b+(1-t)x)$$

$$f'(a+tx) = g'(b+(1-t)x).$$
(3)

Alice maximizes her return when she splits her assets so that Bob and Carol end up offering the same rate. Informally, if Bob had ended up providing a better rate, then Alice should have given him a larger share. If there is no $t \in (0,1)$ that satisfies Equation 3, Alice should give all her assets to the AMM with the better rate.

How should parallel composition be defined for AMMs with multiple asset types? Suppose Alice has some combination d of assets in X_1, \ldots, X_p that she wants to convert into some combination of assets in Y_1, \ldots, Y_q . Alice has a choice of two alternative AMMs: $B(x_1, \ldots, x_p, y_1, \ldots, y_q)$ and $C(x_1, \ldots, x_p, y_1, \ldots, y_q)$. Perhaps the most sensible way to define parallel composition is through asset virtualization. Alice's input asset vector d induces a valuation $d/\|d\|_1$ which can be used to define a virtual asset X from X_1, \ldots, X_p . Alice chooses a valuation v for her Y_1, \ldots, Y_q outputs (perhaps the market valuation) which can be used to define a virtual asset X from X_1, \ldots, X_q . After asset virtualization, the alternative AMMs have the form: $\tilde{B}(x,t)$ and $\tilde{C}(x,t)$, and the definition of parallel composition proceeds as before. The next lemma

describes properties of stable points for two 2-dimensional AMMs composed in parallel.

If (x, f(x)) and (y, g(y)) are two n-dimensional AMMs in states (a, f(a)) and (b, g(b)), we can define parallel composition as follows. For $t \in [0, 1]^{n-1}$, let $h_t(x) = f(a + t * x) + g(b + (1 - t) * x)$, where * is component-wise multiplication.

LEMMA 23. Let $(\mathbf{v}, 1 - \|\mathbf{v}\|_1)$ be a valuation, and let $A := (\mathbf{x}, f(\mathbf{x}))$ and $B := (\mathbf{y}, g(\mathbf{y}))$. If $(\mathbf{a}, f(\mathbf{a}))$ and $(\mathbf{b}, g(\mathbf{a}))$ are both stable points with respect to $(\mathbf{v}, 1 - \|\mathbf{v}\|_1)$). then $(0, h_t(0))$ is the stable point on $A \|B$ for $(\mathbf{v}, 1 - \|\mathbf{v}\|_1)$ for all $t \in \mathbb{R}^{n-1}$.

PROOF. For
$$t \in \mathbb{R}^{n-1}$$
,

$$v \cdot a + (1 - ||v||_1) f(a)$$

 $< v \cdot (a + t * x) + (1 - ||v||_1) f(a + t * x)$

$$(1 - \|\mathbf{v}\|_1)f(\mathbf{a}) < (\mathbf{t} * \mathbf{x}) \cdot \mathbf{v} + (1 - \|\mathbf{v}\|_1)f(\mathbf{a} + \mathbf{t} * \mathbf{x})$$

and

$$v \cdot b + (1 - ||v||_1)g(b)$$

 $< v \cdot (b + (1 - t) * y) + (1 - ||v||_1)g(b + (1 - t) * y)$

$$(1 - ||v||_1)g(b) < ((1 - t) * y) \cdot v + (1 - ||v||_1)g(b + (1 - t) * y),$$

implying

$$\begin{aligned} v \cdot 0 + & (1 - \|v\|_1)h_t(0) \\ &= (1 - \|v\|_1)h_t(0) \\ &= (1 - \|v\|_1)f(a) + (1 - \|v\|_1)g(b) \\ &< (t * x) \cdot v + (1 - \|v\|_1)f(a + t * x) \\ &+ (1 - t) * x \cdot v + \\ & (1 - \|v\|_1)g(b + (1 - t) * x) \\ &= v \cdot x + (1 - \|v\|_1)(f(a + t \cdot x) \\ &+ g(b + (1 - t) \cdot x)) \\ &= v \cdot x + (1 - \|v\|_1)h(x) \end{aligned}$$

so
$$(0, h_t(0))$$
 is a stable point for $(\boldsymbol{v}, 1 - \|\boldsymbol{v}\|_1)$.

Most generally, if we have two AMMs A(x, z) and B(y, z'), and valuation w, we can write A|w as (x, f(x)) and B|w as (y, g(y)). We then can define parallel composition as before.

THEOREM 8. Let (v, v') be a valuation, and let A(x, z) and B(y, z') two AMMs. If (a^*, b^*) is the stable point for A, and (c^*, d^*) the stable point for B, both with respect to (v, v'), then $(v, ||v'||_2^2)/||(v, ||v'||_2^2)||_1$ is the stable point for (A|v')||(B|v').

PROOF. Lemma 13 implies that $(\boldsymbol{a}^*, f(\boldsymbol{a}^*))$ and $(\boldsymbol{c}^*, g(\boldsymbol{c}^*))$ are stable on $A|\boldsymbol{v}'$ and $B|\boldsymbol{v}'$, both with respect to $(\boldsymbol{v}, \|\boldsymbol{v}'\|_2^2)$. Lemma 23 implies that $(0, h_t(0))$ is the stable point for $(A|\boldsymbol{v}')\|(B|\boldsymbol{v}')$ with respect to valuation $(\boldsymbol{v}, \|\boldsymbol{v}'\|_2^2)$.

8 RELATED WORK

Angeris and Chitra [5] introduce a *constant function market maker* model and consider conditions that ensure that agents who interact with AMMs correctly report asset prices. Our work, based on a similar but not identical AMM model, focuses on properties such as defining AMM composition, AMM topology, and the role of stable points.

Uniswap [6, 22] is a family of constant-product AMMs that originally traded between ERC-20 tokens [13] and ether cryptocurrency. Trading between ERC-20 assets requires sequential composition of the kind analyzed in Section 6. Uniswap v2 [2] added direct trading between selected pairs of ERC-20 tokens, and Uniswap v3 [2] allows liquidity providers to restrict the range of prices in which their asset participate, giving rise to a form of parallel composition of the kind analyzed in Section 7.

Bancor [16] AMMs permit more flexible pricing schemes. The state space manifold is parameterized by a weight, where different weights yield different curves. Bancor AMMs trade between ERC-20 assets and Bancor-issued BNT tokens. Prices are a function of assets held and BNT tokens in circulation. Later versions [7] include integration with external "price oracles" to keep prices in line with market conditions.

Balancer [18] AMMs trade across more than two assets. Instead of constant product, their state space is given by constant mean formula $c = \prod_{i=1}^{n} x_i^{w_i}$, where c is constant, x_i is amount of asset X_i held by the contract, and the w_i are adjustable weights that form a valuation.

Curve [12] uses a custom curve specialized for trading across multiple *stablecoins*, digital assets whose values are tied to fiat currencies such as the dollar, and thus unlikely to be volatile.

There are many are more examples of AMMs: see Pourpouneh *et al.* [20] for a survey.

Before there were AMMs for decentralized finance, there were AMMs for *event prediction markets*, where parties trade securities that pay a premium if and only if some event occurs within a specified time. A community of researchers has focused on prediction-market AMMs [1, 10, 11, 14, 15]. Despite superficial similarities, event prediction AMMs and security

AMMs differ from DEFI AMMs is important ways: pricing models are different because prediction outcome spaces are discrete rather than continuous, prediction securities have finite lifetimes, and composition of AMMs is not a concern.

We also note that the mathematical structure of AMMs resembles that of a consumer utility curve from classical economics [19], where assets are replaced with goods. The optimal arbitrage problem is not new. A consumer choosing an optimal good bundle for a fixed set of prices is the same as an arbitrageur choosing an optimal point on an AMM with respect to a market valuation. This is known as the *expenditure minimization problem* [19]. While AMMs and consumer indifference surfaces are mathematically similar, they are fundamentally different in application. In particular, traders interact with AMMs via composition, an issue that does not arise in the consumer model.

9 CONCLUSIONS

Modern AMMs provide increasingly complex rules for trades. For example, Uniswap v3 [3] allows liquidity providers to choose to take trades only over finite ranges. In the future, we would like to understand how to define composition operators for varieties of range-restricted AMMs, but we hope the reader is convinced that understanding composition of simple AMMs is already challenging, and an important step to understand more general cases.

AMMs are increasingly being integrated with external price information oracles. For example, Krishnamachari *et al.* [17] describe a family of AMMs capable of adjusting their curves in response to reported price changes. In future work, we plan to investigate composition for AMMs that make use of oracle services..

The work presented here is a first step toward analyzing decentralized finance from a distributed computing perspective. In future work, we hope to consider more complex networks of AMMs, the challenges of cross-chain AMMs, synchronization problems such as front-running, as well as more adaptive AMM structures.

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