

**NFT GAMING COLLECTIBLES VALUATION METHODS, CHALLENGES,
AND FUTURE DIRECTIONS**

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Abstract

The explosive popularity of blockchain-based gaming has brought non-fungible tokens (NFTs) into the fray, as tradable collectibles that connote not only monetary worth but an experience as well. In contrast to traditional in-game items, gaming NFTs exist in two domains of use, on-chain marketplaces and off-chain ecosystems, providing new twists to their valuation. This paper analyzes the approach, issues, and possible developments of the NFT gaming collectibles valuation by consolidating the knowledge in economics, data science, and digital games research. We then present a taxonomy of value drivers, gameplay utility, rarity, provenance, community effects, and tokenomics, to develop a conceptual background. We then compare the relative advantages and (*) disadvantages of popular valuation methods, including hedonic pricing models, repeat-sales indices, machine learning, and liquidity-adjusted approaches. We also identify structural issues of this nascent market, including wash trading, fragmentation of liquidity, evolving product utility, mutability of metadata, and regulatory uncertainty. Critical synthesis helps us see how valuation practices are susceptible to abuse on both sides: how we can abuse the market, but also how we can design the technical aspects of markets to cheat. Lastly, we discuss research directions in the form of standardized benchmarks, dynamic valuation frameworks, interoperability-adjusted models, and integrity checks. The paper will contribute to both the academic and practice worlds as it will provide a reproducible course of action by scholars, game development companies, investors, and marketplaces interested in designing open and sustainable valuation methods for the gaming NFTs.

Keywords: NFT valuation; gaming collectibles; on-chain analytics; hedonic pricing models; market microstructure; wash trading detection; tokenomics.

INTRODUCTION

1.1 Motivation

The ascendancy of blockchain-based technology has transformed digital ownership, and non-fungible tokens (NFTs) have become the primary medium of representing digital, scarce, and tradable assets. One of the fastest-growing and dynamic segments is gaming NFTs that include all forms, such as weapons, skins, avatars, digital land parcels, and in-game passes. Unlike traditional game assets that are stored in centralized servers, NFTs can be verified and traded on the blockchain. Their proliferation has made it possible to have secondary markets in which digital goods can be appreciated, depreciated, and circulated without any control of the issuing studio.

Such a fast growth has, however, brought a lot of confusion. Investors, players, and studios have not been able to clearly answer a very fundamental question: what is the fair price of a gaming NFT collectible? Conventional valuation practices based on art, subordinated financial assets, or imaginary economies are inadequate since gaming NFTs have a dual aspect: (i) a financial value determined by market forces and (ii) a utility value determined by gameplay utility, scarcity, and community value. Consequently, valuations tend to be speculative, volatile, and manipulative. This valuation confusion issue can now be seen as the bane of both serious investors and intermediaries looking to build out tokenized economies, and regulators trying to regulate such environments.

The ante is high. The in-game economies have attained multi-billion-dollar revenues, such that some NFT-based games can approach the revenues of traditional forms of entertainment. Investors see gaming NFTs as another asset, and an ever-increasing number of studios monetize by selling NFTs and royalties on them. In the meantime, mispricing, speculative bubbles, and market manipulation undermine trust and sustainability. A sound system of valuation is therefore compulsory to clarify the legitimacy and long-term sustainability of gaming NFTs.

1.2 Scope

NFT games collectibles are a particular area where collectibles represent; there is:

- Items (weapons, tools, accouterments with gameplay purpose).
- Skinning (the straightening or otherwise changing of an aesthetic article/aesthetic alteration without functional effect),
- Avatars/Fixed Characters (game characters with changeable traits),
- Land plots (metaverse-style platforms),
- Laps/Badges (Passes/Badges granting access to the events, modes, or rights).

NFTs that are purely generative art and defi-NFT hybrids are not in the ambit of the scope unless they overlap with the direct areas of gaming environments. This interest allows exploring valuation systems in a more controlled land context that focuses on what is peculiar in interactive and gameplay-driven ecosystems. Although some similarities can be found with art or DeFi collectibles, gaming NFTs are unique in being treated as an integrated part of a live game that operates on balance writings and where the user community continues to evolve.

1.3 Contributions of this Article

The present article will make contributions in the following ways:

- Among the main contributions of the article to both academic and practice-oriented discourse on NFT gaming valuation, it is possible to single out the following ones:
- A taxonomy valuation according to which the most significant contributive factors to value in gaming collectibles can be classified (utility, rarity, network effects, tokenomics, etc.).
- A comparative evaluation of valuation techniques, both traditional, such as hedonic pricing and time sales standards, as well as more recent technology-based techniques, such as machine learning and liquidity-adjusted techniques.
- A risk and bias map that provides distortions of valuation due to market manipulation (e.g., wash trading), liquidity droughts, and changing gameplay rules.
- A research agenda, with the perspective of methodological and technological prospects of more reliable and dynamic valuation.

Informed recommendations to the relevant stakeholders in the game industry--development and distribution studios, marketplaces, analysts, and outsiders--relating to what best practices should be followed regarding valuation reporting, data reliability, and sustainable ecosystem design.

1.4 Roadmap of the Paper

The rest of the paper is organized in the following way:

- Section 2 provides a conceptual grammar of gaming NFT value, presenting the main factors that may influence utility and speculative value.
- Section 3 briefly outlines the metrics and data needed to perform the valuation and what on-chain and off-chain sources could be used.
- Section 4 compares the different valuation techniques, noting their strength, weaknesses, and merits in the various contexts.
- In Section 5, system disadvantages and trappings that jeopardize proper valuation are discussed, such as wash trading and liquidity fragmentation.
- Section 6 gives a set of case briefs exemplifying the valuation mechanisms in action of various forms of gaming NFTs.
- Section 7 summarizes guidelines to assist studios, marketplaces, and analysts in the task of valuation.
- Section 8 examines what might come, discussing advancements that include dynamic NFTs, into put-puts, and regulation.
- Section 9 ends by summing up the lessons and reaffirming the programme of vigorous and open valuation structures.

This article aims at bridging that gap by integrating analytical clarity, methodological assessment, and futuristic perspectives, trying to adequately understand NFT gaming collectible valuation, an aspect that currently lacks proper academic-grade and professional interpretation.

2. CONCEPTUAL FOUNDATIONS OF GAMING NFT VALUE

2.1 Dimensions of Value in Gaming NFTs

The valuation of the gaming NFT cannot be measured in one dimension. Rather, it is a multi-factor model of value creation, which is a combination of both the intrinsic market-related game features and extrinsic market indicators. The dimensions that matter the most are as follows:

- **Utility Value:** Certain NFTs may have utility in a game, like superior damage weapons, faster mounts, or unique abilities. This value is directly convertible to player demand, hence price premiums.
- **Aesthetic/Status Value** Although an NFT may not be directly related to gameplay, its cosmetic aspect (skins, avatars) also represents possible symbolism. These NFTs serve the same functions as luxury items in the physical world do: They are markers of status in the online community.
- **Scarcity and Provenance:** Rarity is still one of the foundations of value. Uniquely scarce on-chain items, or those attached to slices of conquests (first-edition drops, tournament rewards), tend to be more expensive.

- Network Effects & Community: The better and the harder players play with a game and make use of the NFTs, the higher the demand for collectibles. Communities enhance value through building reputation and cultural value.
- Creator Reputation & Intellectual Property (IP) Applicable to NFTs connected with a known asset or well-established intellectual property (i.e., Marvel, Pokémon), high rankings may be more frequently assigned to the item.
- Interoperability: The cross-game and cross-chain compatibility increases long-term demand as users view value in portability.
- Royalties & Tokenomics Royalties or utility built into staking contracts can also be enforced through smart contracts. These design characteristics impact the short-term liquidity and future expected value.

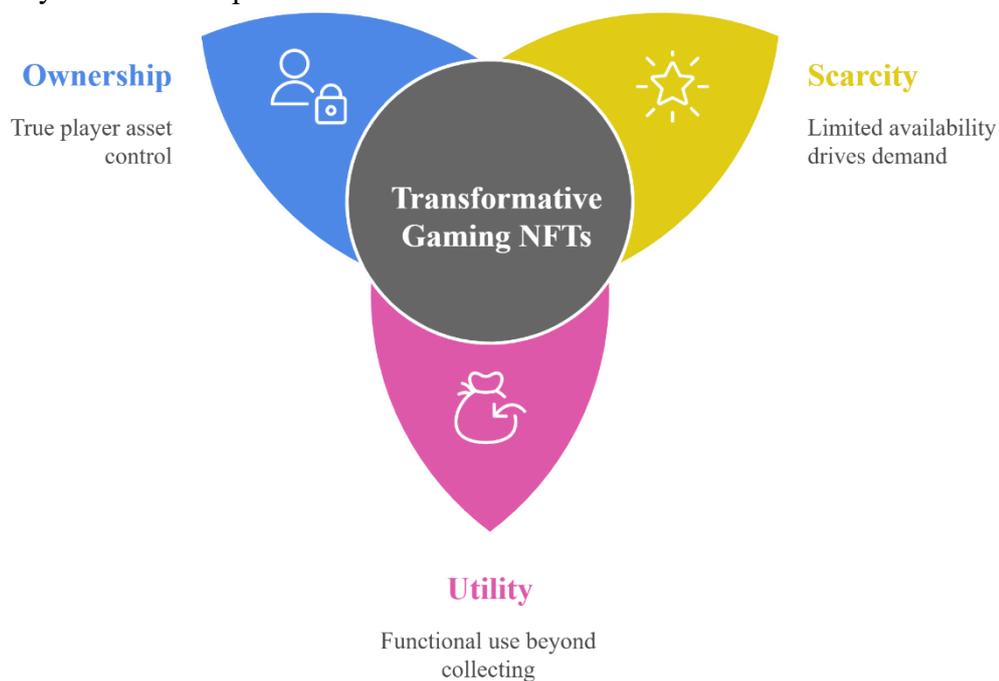


Figure 1: Dimensions of Value in Gaming NFTs

2.2 Speculative price vs. fundamental value

It is necessary to note the difference between fundamental/game-utility value and speculative price.

- Fundamental value is based on in-game functionality, design quality, and community usage. It is more or less stable and related to the practicality of the NFT in the gaming experience.
- Speculative value is the product of bullishness, fads, and outside financial markets. Streams of speculation will increase the prices temporarily, but they create instability and problematize the predictability.

NFT prices are a combination of the two in practice. Even an exclusively functional aspect (such as a performance-boosting weapon) can experience over-valuation at some times (such as when a popular game is subject to hype), whereas a purely cosmetic skin can increase in value due to celebrity endorsements or viral marketing efforts.

2.3 On-chain vs. Off-chain determinants

Gaming NFT valuation involves accountability of the combined picture of on-chain and off-chain expenses:

- On-chain factors: ownership trades, rarity scores, trading volumes, wallet concentration, and royalty settings.
- Off-chain factors: game server mechanics, patch updates, tournament events, and balance changes.

Such bi-dependence necessitates valuation models integrating blockchain analytics with game telemetry data, which introduces a methodological gap because it is not the approach that has been taken by other digital collectibles.

Table 1: Value Drivers Taxonomy for Gaming NFT Collectibles

Driver	Example in Games	Measurable Proxy (On/Off-chain)	Expected Effect on Price	Caveats
Utility Value	Weapon with higher damage stats	Win-rate correlation; gameplay telemetry (off-chain)	Positive premium for performance	Balance patches can reduce or nullify value
Aesthetic/Status Value	Limited-edition avatar skin	Rarity rank (on-chain), event exclusivity	High if community values aesthetics	May fade if design trends shift
Scarcity & Provenance	First-edition tournament trophy NFT	Mint date, provenance trail (on-chain)	Strong premium for historical rarity	Replicas or derivatives may dilute uniqueness
Network Effects	Popular game with large active base	DAU/MAU metrics, holder network centrality	Higher adoption increases demand	Value falls if community shrinks
Creator Reputation & IP	Branded character from major studio	Studio reputation, IP brand recognition	Premium from trust and recognition	Dependent on ongoing brand strength
Interoperability	NFT usable across multiple games	Cross-chain utility flags, integration count	Higher due to broader functionality	Technical risks in maintaining compatibility
Royalties & Tokenomics	NFT with embedded staking rewards	Smart contract royalty %, staking yield data	Premium if rewards are sustainable	Tokenomics can collapse if poorly designed

3. DATA & METRICS FOR VALUATION

The valuation of gaming NFTs needs accurate, multi-faceted data feeds that reflect not only blockchain activity but also in-game specifics. Unlike traditional assets that have standard

reports, e.g., equities, NFT data is spread across the ledgers on-chain, centralised game servers, and marketplaces. The section described the essential categories of the metrics, which will be required to build meaningful valuation models.

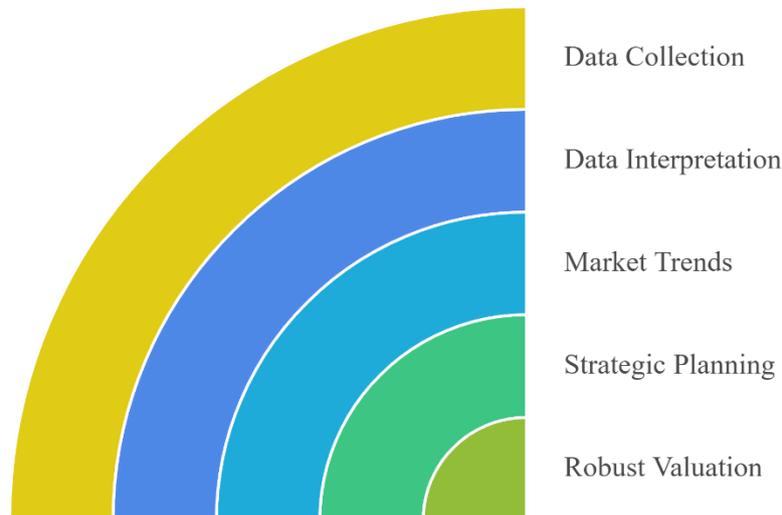


Figure 2 : Data & Metrics for Valuation

3.1 On-chain data

Information on the chain is the core of NFT valuation, since it represents transactions and ownership confirmed by the chain. Major indicators are:

- Mints: The first minting of NFTs gives initial supply statistics. Small-scale production mints can tend towards the increased scarcity premiums.
- Transfers: The liquidity and the user activity are traced. The high transfer velocity may be an indication of speculative trading.
- Top Wallet Concentration: The distribution risk of NFTs is revealed by the percentage of the top wallets. A concentrated holder base can contribute to price volatility.
- Time-to-Resale: The time between resale and purchase puts into evidence turnover in the market. Short periods imply speculation; long periods can imply either attachment or ill liquidity.
- Active listings: Available inventory at various price points allows an insight into the liquidity on the supply side.
- Rarity Scores: Beats that rank high with respect to other beats in the same series receive the maximum rarity rank.
- Chain of custody: Ownership chain tends to boost the impressions of authenticity and value, particularly of event-based NFTs.

3.2 Market microstructure

NFTs use market spaces that have a similar microstructure to financial exchanges with reduced transparency. Metrics include:

- Bid-Ask Gaps: The distinction between the uppermost bids and the lowest asks is an indicator of liquidity. Wider gaps are indicators of illiquidity and price uncertainty.
- Order-Book Depth: Depth data can help investors withstand large trades since they are not offered on all platforms.
- Floor price vs. Median Price: the lowest listing price is widely used, but this metric alone is misleading when it is stale. The comparison of median prices is a more solid estimate.
- Stale Listings: A stale listing is a listing that has failed to sell after some amount of time and indicates that the floors may be inflated.
- Cross-Market Fragmentation: Major price differences exist among marketplaces (e.g., OpenSea and Blur). The cross-market tracking minimizes arbitrage bias.

3.3 Off-chain / game telemetry

Gaming NFTs rely partly on gameplay to establish value, and so off-chain telemetry is critical:

- Drop Rates: Scarcity is often related to the drop mechanics of the game.
- Patch / Nerf History: The balance is made by the game developers; a nerf can render previously overpowered things useless.
- Win-Rate Impact: Functional NFTs can be rated by their statistical value in increasing the probability of a win.
- Playtime Correlations: An NFT increases in embedded utility the more it is used by the player in-game.
- DAU/MAU: Demand is directly influenced by the activity in the community.
- Seasonal schedules: Seasonal events or tournaments can really boost a spock in demand.

3.4 Risk & integrity signals

NFT markets become a threat to valuation integrity. Measurements of the detection and mitigation of these risks include:

- Wash Trading Flags:- Unusually numerous trades between the same wallets at excessive prices.
- Sybil / Airdrop Farming: Seemingly multiple wallets being controlled by the same actor make the participation and the holder information biased.
- Transaction Costs: Marketplace fees and royalties are direct costs to net returns, which may also affect the frequency of trading.
- Oracle or Metadata Risks: The on-chain assets are backed by off-chain metadata that is stored through centralized or mutable means, thus creating reliability risks that can depreciate the assets instantaneously.

3.5 Outcome variables

Lastly comes the valuation, which involves the determination of measurable variables in outcome terms:

- Sale Price: Raw prices, whilst rather noisy, turn out to be the most common benchmark out there.
- Log-Price: Log-transformations stabilize the variance and enhance regression.
- Liquidity-Adjusted Price: takes into consideration the expected discount to make the liquidation faster.

- Risk-Adjusted Returns: Adjusts returns to volatility so that they are more aligned across NFTs.

Table 2: Core Metrics Map

Metric	Type	Definition	How to Compute	Pros	Limitations
Mints	On-chain	Number of NFTs initially created	Blockchain mint event logs	Establishes baseline supply	Limited without context of demand
Holder Concentration	On-chain	% controlled by top wallets	Herfindahl index / Gini coefficient	Reveals ownership risk	Doesn't capture off-chain affiliations
Bid-Ask Gap	Market Structure	Difference between highest bid & lowest ask	Order book analysis (if available)	Proxy for liquidity	Not always transparent
Floor vs Median Price	Market Structure	Comparison of lowest listing vs central tendency	Median of recent sales vs floor price	Highlights inflated or stale floors	Sensitive to outliers
Drop Rates	Off-chain	Frequency of item drops in gameplay	Game telemetry / developer disclosures	Links scarcity to design	Often proprietary / unavailable
Patch/Nerf History	Off-chain	Record of game balance changes	Patch notes & server logs	Captures evolving utility	Retrospective; hard to quantify immediately
Wash Trading Flags	Risk & Integrity	Repeated circular trades between same wallets	Wallet transaction graph analysis	Detects manipulation	False positives possible
Liquidity-Adjusted Price	Outcome Variable	Price adjusted for time-to-sale	Discounting models based on hazard rate	Realistic liquidation estimate	Requires rich transaction data

4. VALUATION METHODS (COMPARATIVE)

Gaming NFT valuation is different in its own way since such assets are at the crossroads between financial markets, gaming ecosystems, and social communities. In contrast to other secured assets with consistent cash flows or intrinsic value indicators, the gaming NFTs use a murky combination of in-game value, visual aesthetics, network plays, and popular speculation to establish value. This section explores and compares the key approaches that today are or may

apply to valuing NFT gaming collectibles. Each of the methods is described through theory, gaming NFTs application, benefits, and drawbacks.

4.1 Hedonic pricing models

Concept:

Hedonic pricing estimates the value contribution to an NFT market price of its attributes. This is similar to the valuation of real estate, where the value of a house will be determined by the size, location, and amenities to be considered. In Non Fungible Tokens, rarity rank, collection identity, visual traits, in-game performance stats, and provenance indicators are attributes.

Gaming NFTs Use-Case:

As an example, a situs slot baru rare weapon skin in a blockchain-based shooter is likely to sell at a higher price, not because it exists, but because:

- It is part of an exclusive drop,
- It ranks high in rarity in the collection.
- It also used to belong to a famous esports gamer.

The contribution of each can be measured using regression-based hedonic models that separate how much value each factor contributes at the expected price.

Strengths:

- Good interpretability so that results can be made accessible in the hands of the analysts, regulators, and studios.
- It works well when giving price advice in sets of items with clear characteristics.

Limitations:

- NFTs in gaming develop dynamically (patches/ nerfs). Where a weapon that has provided great benefits becomes useless overnight, it can lead to the loss of model stability.
- Assumes linear and additive effects, although in reality many valuations are nonlinear (e.g., the top 1 percent becomes a different price with a greater premium).

Enhancements:

- They include splines, interaction terms, and collection fixed-effects, which allow them to capture non-linearities and the game contexts.
- Control of patch or season effects can be accomplished by combining hedonic models with time dummies.

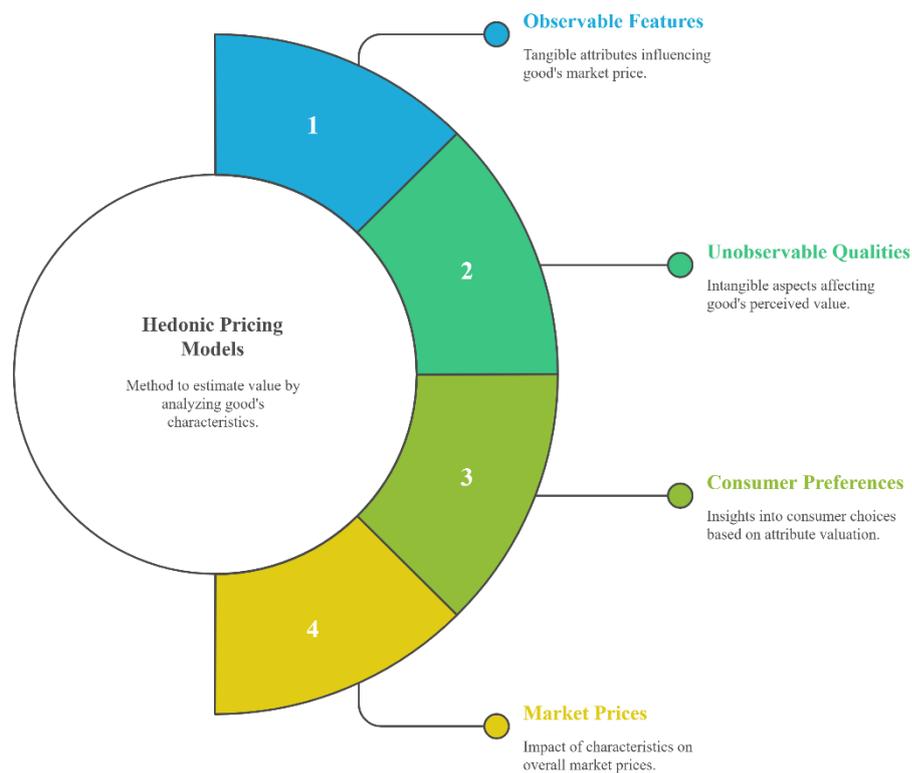


Figure 3: Hedonic Pricing Models

4.2 Repeat-sales & index construction

Concept:

Repeat-sales models time the changes in price of the same NFT in multiple transactions and, in this way, avoid modeling the traits. This is similar to a real estate index (e.g., Case-Shiller index).

Use Case in Gaming NFT:

Assume that the same virtual sword is sold 3 times within a year, then repeat-sales regression would be able to estimate its appreciation, independent of its rarity or other undifferentiated traits. Assembling these prices together by item, it is possible to create price indices of collections or whole marketplaces, which are comparable to stock indices.

Strengths:

- Unobservable item-specific features that can arise as the same item is resold.
- Enables the creation of market-wide indices, which are handy for comparing asset classes or sets.

Limitations:

- NFT markets are liquidity-poor. Most collectibles can only be sold once and generate small data.
- Prone to survivorship bias: Survivorship bias is likely to occur because items that resell may not reflect the wider market (e.g., only items that are hot sellers again).

Enhancements:

- Combine the repeat-sales indices with hedonic models (hybrid indices).
- Do bootstrap-based and Bayesian shrinkage to overcome the small sample biases.

4.3 Machine Learning Methods

Concept:

Machine learning (ML) models find complex associations using large datasets: they include gradient boosting (XGBoost, LightGBM), random forests, and elastic-net regression. Some of these features focus on on-chain indicators crowding (rarity rank, wallet concentration, transaction velocity), whereas other features look at off-chain telemetry: (patch history, DAU/MAU).

Applications to Gaming NFTs:

A short-term NFT forecasting model based on LSTM may be applied to such markets as Axie Infinity or the Sandbox to predict their sales prices. They are capable of introducing network graph-based features (e.g., whether a product is possessed by a high-profile wallet) and temporal variables (market momentum, social media buzz).

Strengths:

- Can deal with high-dimensional relationships, nonlinear relationships, and interaction.
- The possibility of higher predictive power as compared to the missed models

Limitations:

- Black-box problem: hard to understand by the stakeholders.
- Data sparseness compromises the performance of training; overfitting is a phenomenon.
- The cross-validation assumptions do not work in the case of NFT markets with non-IID distributions (clustered trading, hype cycles).

Enhancements:

Interpretability tools (SHAP values, LIME) with explainability can be used to increase transparency.

4.4 Liquidity-adjusted valuation

Concept:

NFTs do not work with liquidity. The methods to introduce liquidity include bid-ask spreads, order book depth, and hazard models of time-to-sale. The price of an NFT is less than what has to be discounted to redeem it within a reasonable period.

NFTs to Gaming:

A collectible land parcel in Decentraland could be listed at 10 ETH, but due to market data indicating an average sale of 60 days, it may need a 20% liquidity discount, thus changing its liquidity-adjusted value to 8 ETH.

Strengths:

- Makes more realistic valuations, particularly for those investors interested in exit plans.
- Explains directly as to thin order books and fragmented liquidity.

Limitations:

- Jump data needs granular data (bid-ask spreads, order-book snapshots) that is not always accessible across the NFT ecosystem.
- Susceptible to volatility in the activity of trading.

Enhancements:

- Combine survival analysis models with the calculation of liquidation probabilities.
- Develop liquidity-adjusted indices that could be used as more spot-on market benchmarks.

4.5 Option-style & real-options intuition

Concept:

NFTs in Gaming can often represent embedded optionality: new value can be unlocked in the future (via patches, tournaments, metaverse integrations, etc.). In the real-options theory, this is modelled in terms of the right--but not the duty--to invoke future opportunities.

Use Case for Gaming NFTs:

An NFT that gives access to future esports gaming contests will be effective as an option: the higher the likelihood and the attractiveness of the event, the more valuable an NFT will become. Just as well, land NFTs in The Sandbox can have their own doomsday clock-like option value insofar as potential partnerships or brand integrations are concerned.

Strengths:

- Records future, trending elements of gaming NFTs.
- Appreciates uncertainty and optionality: not just present utility.

Limitations:

- It is hard to parametrize the probability of game updates or event-driven demand.
- Depends in large measure on speculative inferences regarding developer roadmaps.

Enhancements:

- Calibrate option premiums (e.g., measure price changes surrounding patch releases using event-study techniques).
- He modeled the future utility with binomial-lattice constructs.

4.6 Network & social signals

Concept:

Network analytics and social measures can also be applied in NFT valuation. Gaming NFTs flourish in community ecosystems, and they are frequently subject to network effects.

Applications to Gaming NFTs:

- Holders' graph centrality: The NFT that is owned by such wallets as streamers, esports players, etc., may gain value quickly.
- Community energy: The volatility in Discord/Reddit/Twitter takes place simultaneously with increased volatility in the prices.
- Influencer endorsements: First or second-party endorsement by a celebrity gamer will make a premium valuation.

Strengths:

- Tracks the real-time demand indicators.
- Reflects the societal and cultural anchoring associated with NFTs.

Limitations:

- Low resistance to bot-manipulation (bot activity, shill campaigns).
- It is hard to separate meaningful activity and hype.

Enhancements:

- Apply graph anomaly detection to determine organically grown data and artificially generated content.
- Combine with sentiment analysis in order to get a better sense.

4.7 Price integrity adjustments

Concept:

The manipulations, such as wash trading, are common; hence, valuations will need filters and robustness testing. Techniques include:

- Graph-based AO: detection of circular trading between wallets.
- The robust statistic estimators: Huber regression or M-estimators that depress the outliers.

NFTs Live Games:

As an example, a wallet that repeatedly sells/buys a cosmetic skin at an over-market price may be flagged as an integrity issue, and the transactions filtered out before value is estimated.

Strengths:

- They are vital in coming up with believable valuations.

- Spreads these distorted measures across investors and regulators.

Limitations:

- False positives, i.e., valid trades are identified or marked.
- It can only be monitored continuously because there is an evolution in the strategies of manipulation.

Enhancements:

- Uniformity in integrity screens at marketplaces
- Apply consensus-driven validation systems to separate false and honest activity.

Table 3: Comparison of Valuation Methods for Gaming NFTs (Expanded)

Method	Data Requirements	Captures Utility?	Handles Liquidity?	Robust to Manipulation?	Interpretability	Best Use Cases	Key Limitations
Hedonic Pricing Models	NFT attributes, rarity scores, metadata	Yes (static)	No	Low	High	Collections with rich attribute data	Sensitive to game patches; assumes linearity
Repeat-Sales / Indices	Historical resale data	Indirect	No	Medium	High	Market indices, long-term tracking	Sparse data; survivorship bias
Machine Learning Models	On-chain + off-chain features	Strong (dynamic)	Indirect	Medium	Low–Medium	High-volume markets; predictive analytics	Black-box problem; overfitting
Liquidity-Adjusted Models	Bid-ask gaps, time-to-sale	Weak	Strong	Medium	Medium	Illiquid assets; investor exit strategies	Data intensive; fragile to shocks
Option-Style Models	Event forecasts, roadmap data	Medium	Weak	Low	Low	Anticipating patch/event-driven value	Highly speculative
Network/Social Models	Social metrics, holder graph data	Weak	Indirect	Low	Medium	Early-stage adoption, hype-sensitive assets	Easily manipulated; noisy
Price Integrity Filters	Transaction graphs, anomaly detection	N/A	N/A	Strong	Medium	Baseline filter for all models	Risk of excluding genuine trades

5. CHALLENGES AND PITFALLS

It is not easy to value NFT game collectibles. The tools and metrics highlighted in the previous sections present significant opportunities that are subject to limitations that are both due to the

relative immaturity of the market and its volatility. The major risks, distortions, and structural pitfalls that complicate valuation are outlined in the following subsections with an emphasis on how they materialize in practice.

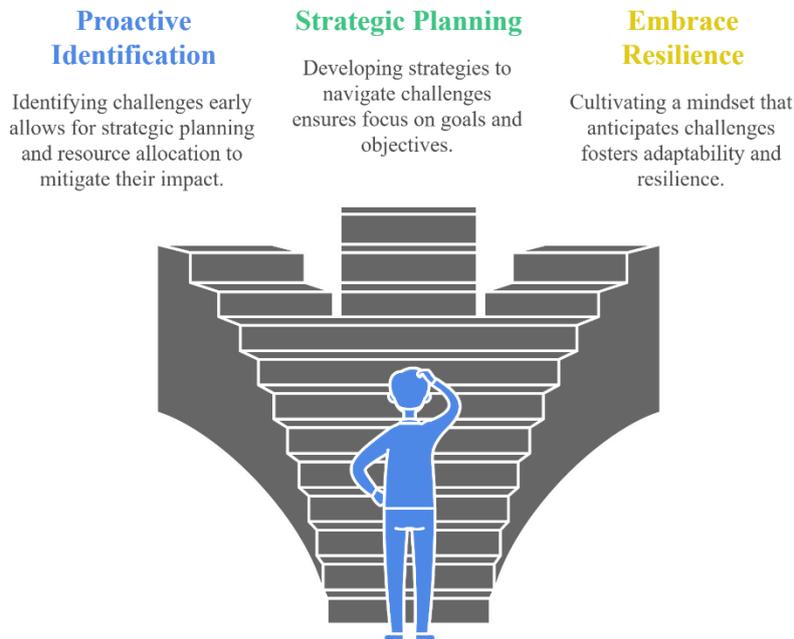


Figure 4: Challenges and Pitfalls

5.1 Market integrity: Wash trading, spoof listings, coordinated floors

One of the greatest threats to integrity in the market is market integrity. Wash trading creates unrealistic demand signals by having a user buy an asset and then sell it to them on a different wallet. Minute observations are that more than 40 percent of NFT turnover in certain markets is manipulative, which contributes to the overestimated indices and wrongful grounding of valuation. In a similar vein, spoof listings (listings at unrealistic floor prices with no interest in selling) may distort the impression of demand. Worse still, there are coordinated floor campaigns, in which camps of sellers commit not to undercut each other, and thus visibly keep the bottom price high. Devoid of powerful detection, valuation models can magnify manipulated data.

5.2 Liquidity & fragmentation: Thin books, multi-market price dispersion, stale floors

Unlike equity markets, NFT markets frequently have thin liquidity, making it so that only a few units may trade weekly. The relatively low prices of thin books mean that they are vulnerable to outlier transactions and thus highly volatile with respect to their intrinsic value. On top of that, NFTs tend to appear across multiple marketplaces, which results in fragmented order books. As a floor price on one platform does not necessarily reflect concurrent sales on other platforms, there can be high values of price dispersion. Valuation models based on single-platform data might therefore give a one-sided/misleading image. The factor of stale floors--when prices are not adjusted because of no trading activity--adds to the problem, particularly in the older collections.

5.3 Evolving utility: Patches, nerfs, buffs, metadata mutability, off-chain dependencies

The utility of Gaming NFTs is different in that its purpose can be changed overnight. At some point, a weapon NFT can get nerfed in a balance patch, and drop in demand dramatically. On

the other hand, buffs or being implemented into new game modes can raise the value quite drastically. Many NFTs are also based on the off-chain metadata that is stored on centralized servers. In case the hosting provider modifies, corrupts, or deletes data, the perceived properties of the NFT can thus alter. The use of off-chain game servers exposes it to dependency risks, e.g., in case of cessation of operations by the game studio, the NFT may become intrinsically worthless even in the case of persistence on the blockchain.

5.4 Royalties & fees: Protocol and marketplace shifts

Unlike traditional financial assets, NFTs tend to come with creator royalties, whereby each sale of the asset automatically transfers a percentage to the studio or creator. Though this is beneficial by aligning developer and player incentives, it puts a valuation uncertainty in place. Certain marketplaces have moved to optional royalty enforcement, which is harmful to the projected cash flows to creators. In the same breath, marketplace fees (e.g., OpenSea vs Blur marketplace fee structures) have been observed to have large impacts in terms of net seller proceeds and buyer willingness to pay. This translates to: on valuation, gross/net price discrepancy would need to be explicitly modeled.

5.5 Standards & interoperability: Cross-chain bridges, token standards, dynamic NFTs

The NFT space is decentralised across various blockchains (Ethereum, Polygon, Solana, ImmutableX), with different NFT standards (ERC-721, ERC-1155, home-grown implementations). The absence of interoperability implies that different liquidity pools and valuations may exist in one and the other trading where the same collectible is offered. Dynamic NFTs, in which the metadata changes (such as advancing a character or enhancing a weapon), add further complications to valuation: histories of transactions cannot always be compared. The valuation models should be dynamic as well as static.

5.6 Legal/IP & compliance: Ownership, EULAs, resale restrictions

Ownership conveyed by NFTs is on-chain, and their legal status is unclear. Most game studios claim that the purchasers do not actually own the intellectual rights (IP) of the product, but grant the purchaser a license to use it under defined terms. EULAs may apply even when blockchain records are applied, and this creates a contentious issue of whether or not such an EULA is enforceable. Such matters are made even more complex by jurisdictional variation: what is a legal asset sale in one jurisdiction may be a service contract in another. This creates risks of compliance and enforceability, which can not be disregarded as far as valuation is concerned.

5.7 Data quality: Incomplete telemetry, survivorship bias, nonstationarity

Consistent valuation necessitates quality data, yet in NFT gaming, data is not equally available. On-chain information will be publicly available, but off-chain telemetry (sink rates, patch rates, player performance) can be patchy and inaccessible. Analysts exposed to survivorship bias commonly discuss only successful projects that have been active long enough to record some activity. Discussed projects tend to be omitted because they were either unsuccessful or were delisted. Market dynamics are also very nonstationary- player preferences, patch effects, and event-driven price spikes combine to mean that past performance no longer forebodes the future.

Table 4: Risk & Bias Map with Mitigations

Risk/Bias	How It Distorts Valuation	Detection Signal	Mitigation Technique	Residual Risk
Wash trading	Inflates volume and prices	Circular wallet activity, repeated addresses	Graph analysis, filtering heuristics	False positives/negatives
Thin liquidity	Volatile price swings	Long intervals between trades	Liquidity-adjusted models	Cannot fully eliminate
Evolving utility	Sudden re-pricing from game patches	Patch notes, win-rate shifts	Dynamic valuation models, event-adjusted indices	Lag in response
Royalties & fees	Distort net vs gross proceeds	Fee structure changes across platforms	Explicit modeling of net-of-fees pricing	Future policy unpredictability
Standards fragmentation	Inconsistent liquidity pools, cross-chain differences	Same collection across chains	Chain-specific indices, normalization layers	Limited interoperability progress
Legal/IP ambiguity	Ownership disputes, resale restrictions	Divergent EULAs, jurisdictional rulings	Legal disclaimers, risk premiums	Uncertain global regulation
Data quality gaps	Misleading or incomplete valuation signals	Missing telemetry, patch notes	Hybrid on/off-chain datasets, proxy variables	Incomplete coverage persists

6. CASE SNAPSHOTS

Ownership conveyed by NFTs is on-chain, and their legal status is unclear. Most game studios claim that the purchasers do not actually own the intellectual rights (IP) of the product, but grant the purchaser a license to use it under defined terms. EULAs may apply even when blockchain records are applied, and this creates a contentious issue of whether or not such an EULA is enforceable. Such matters are made even more complex by jurisdictional variation: what is a legal asset sale in one jurisdiction may be a service contract in another. This creates risks of compliance and enforceability, which can not be disregarded as far as valuation is concerned.

6.1 Skin/Weapon collection case

Weapon skins are commonly the most liquid NFTs in competitive shooters. Their valuation relies heavily on aesthetic/status value and rarity scores rather than functional gameplay advantages. An example of a sniper rifle skin is a case study that illustrates that premium pricing is fuelled by the prestige and scarcity levels of the community. Here, hedonic pricing models can be used well since the attributes (rarity, collection, and finish) can be measured. Nevertheless, spoofed floors are standard market manipulations, and liquidity-adjusted models are required to prevent overvaluation.

6.2 Character/Companion assets case

In role-playing or MOBA-style games, characters, NFTs, and companions are functional utility; the traits of the game can directly influence the win rates. As an example, one companion NFT with a healing effect soared in price after a tournament in an esports event that highlighted its functionality. This shows how off-chain telemetry (win-rate, match stats) needs to be valuable. Such dynamics are better represented through machine learning methodologies using the gameplay data as opposed to static hedonic modeling. The con: patch nerfs are extremely fast value-diluting, which underlines the necessity of dynamic event-knowledge indices.

6.3 Land/Estate items case

Consistent valuation necessitates quality data, yet in NFT gaming, data is not equally available. On-chain information will be publicly available, but off-chain telemetry (sink rates, patch rates, player performance) can be patchy and inaccessible. Analysts exposed to survivorship bias commonly discuss only successful projects that have been active long enough to record some activity. Discussed projects tend to be omitted because they were either unsuccessful or were delisted. Market dynamics are also very nonstationary- player preferences, patch effects, and event-driven price spikes combine to mean that past performance no longer forebodes the future.

Mini-Summary

These situations make it apparent that there is no universal approach for all types of assets. Land: repeat-sales and network-based indicators. Skins: liquidity-adjusted hedonic models. Characters: telemetry-integrated gameplay. Dynamic modeling is essential, along with cross-data integration in valuation, in all instances.

7. PRACTITIONER GUIDELINES FOR STUDIOS, MARKETPLACES, AND ANALYSTS

Basing the valuation of gaming NFTs must be not only a technical but also a practical challenge. Game studios, NFT marketplaces, and independent analysts who are stakeholders must have actionable models to allow valuation practices to be consistent, transparent, and aligned with the dynamic nature of the gaming ecosystems. In this section, the conceptual and methodological lessons of the article are translated into practical rules, namely data hygiene, model choice, liquidity adjustment, and governance.

7.1 Data hygiene and pre-processing

The basis of NFTs' valuation analysis is clean, trusted data. On-chain and off-chain indicators are prone to noise, manipulation, and structural biases. Marketplaces and studios are supposed to incorporate:

- Transaction filtering: Omit wash trading, circular transfer, or self-dealing flagged transactions.
- Normalization of timestamps: Adjust to time-zone difference and align to in-game events (patch releases, tournaments).
- Metadata fixing: Inspect that off-chain data (e.g., files, game statistics) associated with the NFT is cryptographically bound with the NFT token. This avoids any “rug pulls” through editable metadata.

Analysts must record the source of their data, as well as clearly flag their data gaps due to a lack of coverage in a telemetry or the marketplace.

7.2 Model selection by collection maturity

Depending on the maturity of a gaming NFT collection, the valuation will be determined by the most suitable method:

- In early-stage collections (low trade volume), Hedonic models with strong priors or reasonably well-developed scoring by experts can be potentially useful since market transactions adjustments (repeat-sales and liquidity adjustments) are not possible with rare data.
- Mid-stage portfolios (moderate liquidity, some historical sales): Improved repeat-sales indices and hedonic hybrids can enable an appreciation tracking and an attribute-based pricing.
- High volume, deep liquidity collections: Gradient boosting, neural embeddings, and liquidity-adjusted hazard models are possible, but the risk is high.

7.3 Liquidity haircuts and transaction costs

NFT valuation lies at the center of liquidity risk. Investors and analysts are required to quantify a liquidity haircut- the assumed decline in value in case a specific asset has to be sold within a limited period.

Practical techniques include:

- Modeling the floor prices compared to attained median prices measures the depth of the market.
- That requires modeling of time-to-sale distributions; a way of downward adjustment of expected values in case of urgent liquidations.
- Including market place fee regimes and royalty policies in the net proceeds.

Lack of modeling such adjustments can trigger excessive valuations among investors and studios and lead to a liquidity crunch.

7.4 Reporting templates

To enhance transparency, the memos on valuation must contain:

- Assumptions made: That patch stability will be used, event calendars, or rarity definitions.
- Prevention measures: The way the wash trades or spoof listings have been filtered out.
- Sensitivity analysis: The implications of different royalties, depth of liquidity, or the number of users.
- Benchmark comparisons: The asset is compared with collection indices or other non-fungible token gaming benchmarks.

These common reports would enable investors, studios, and regulators to compare valuations under one parameter.

7.5 Governance and tokenomics alignment

Lastly, the decisions made concerning royalties, increasing the supplies, and communication involving balance changes are essential factors of valuation sustainability. Studios should:

- Align royalties percentages with long-term asset health instead of short-term profit maximization.

- Report balance changes openly, with information on how a nerf, a buff, or a new addition could impact the marketplace.
- Promoting governance methods available to individuals in the community who can be represented by the token holders, and further building on trust and legitimacy.

In summary, practitioner guidelines highlight a realistic gap between theory and practice between academic valuation theories and the practical requirements of stakeholders (who have to consider the trust of investors and participation of players, and at the same time, the long-term sustainability).

8. FUTURE DIRECTIONS

The NFT gaming industry is an emerging sector, and valuation techniques will have to develop with technological and regulatory developments. This section presents the main future research and practice directions.

8.1 Dynamic and programmable NFTs

The features of the NFTs are fixed (e.g., color of skin, scarcity); however, the trends are rising towards dynamic NFTs with changing characteristics in real-time. Just to give an example, a sword NFT could gain more power once the player achieves certain in-game accomplishments, or a land parcel could appreciate when players build something into it. This programmability makes the valuation process complicated because attributes of the asset change over time.

Future models would need to have real-time tele-metry and option-style pricing to reflect embedded future potential.

8.2 Cross-game and cross-chain interoperability

The future of the metaverse is anchored to NFTs, which can be useful in more than a single game or blockchain. A weapon that can work across the chains, or a skin that can be applied to two AAA games, may cost much more.

This poses a demand for interoperability-modulated valuation models that estimate the measure of portability as a multiplier of base utility.

8.3 Risk-adjusted benchmarks

There is a great need to develop standardized indices that would reflect risk-adjusted returns of gaming NFTs. Just like in the case of the equity CAPM models, one could see future work to include liquidity-adjusted factor benchmarks that measure systematic risk across collections to allow more appropriate comparisons and, ultimately, portfolio optimization.

8.4 Trusted data and open standards

With mistrusted data, the valuation will be incomplete and obscure. Standards need to be developed:

- Cryptographic metadata promises to stop silent changes.
- Elegant anti-wash strategies that can be replicated on all markets.

- Open telemetry APIs that will enable researchers to incorporate gameplay information into their valuation models.

8.5 Regulatory pathways

Legal constitutions will have a great influence on valuation techniques. Such issues are related to the key ones, as follows:

- These would be recognition of the NFTs as a form of digital property that has crystal clear rights of reselling them.
- The tax treatment of game NFTs is either collectibles or securities.
- The ability to centralize compliance mechanisms of international marketplaces involving more than one jurisdiction.

These routes will decide whether valuation practices can be legitimate and enjoy wide currency or not.

9. CONCLUSION

NFT gaming collectibles are a bit of a different intersection between entertainment, money/finance, and blockchain technology. Gaming NFTs have value based on utility, scarcity, aesthetics, community, and interoperability as opposed to the pure speculation of purely non-gaming NFTs. However, it is exactly this complexity that makes it extremely difficult to come up with fair and reproducible valuation frameworks.

This paper presented a comprehensive taxonomy of value drivers, catalogued fundamental valuation measures, and contrasted major valuation techniques--such as hedonic models, machine learning, and liquidity-adjusted models. It also recognized these issues, like wash trading, fragmentation of liquidity, mutability of metadata, and uncertainties in compliance. By offering case snapshots and practitioner guidelines, the discussion has not only been given theoretical clarity but also practical directions that could be used by the studios, marketplaces, and analysts who are required to make valuation decisions.

Quickly peering into the future, the next step in NFT valuation will be the need to support dynamic attributes, cross-chain interoperability, and risk-adjusted benchmarking. Reputable data channels/anti-pipeline manipulation schemes and regulatory transparency will play a key role in winning the trust of investors and the players.

In the end, the sustainable practices in valuation will need to produce an equitable balance between financial integrity and player engagement so that the gaming NFTs would not become a source of scandal in the framework of entertainment as well as the digital economy. Their path forward requires a cross-disciplinary approach, so they work across game development, blockchain analytics, finances, and law. NFT gaming collectibles have the opportunity to move beyond speculative curiosities to stand the test of the market with rigorous, clear, and adaptive approaches to methods.

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