

# The GP Selection Governance Framework

The 2025 Autumn Governance Series



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## Why This Matters

General partner (GP) selection is one of the most critical fiduciary responsibilities of investment committees. Historically, this process relied heavily on human judgment—relationships, reputation, and subjective diligence. While human evaluation remains indispensable, it is inconsistent, prone to bias, and often weak in integrating tax and systemic risk considerations.

AI tools are now entering this process. They can screen managers faster, aggregate broader data, and provide more consistent comparisons. Yet they also introduce new risks: over-reliance on historical patterns, portfolio convergence, and blind spots around taxation and legal nuance. Governance must therefore evolve. This framework offers committees a structured approach to integrating AI into GP selection while safeguarding fiduciary rigour and transparency with limited partners (LPs).

## The Traditional GP Selection Process

In the pre-AI model, GP selection typically began with screening based on strategy, size, and track record. Quantitative analysis focused on IRR, multiples, and loss ratios, while qualitative diligence emphasized team cohesion, governance practices, and cultural alignment with LPs. Relationships and references were heavily weighted.

This model's strengths were adaptability and contextual awareness—humans can detect red flags beyond the reach of models. Weaknesses included inconsistency, and limited integration of tax implications. Documentation was often uneven, and overrides were justified through consensus rather than

systematic protocols.

## AI's Role in GP Selection

AI augments this process by automating screening, aggregating data from Prequin, PitchBook, and internal records, and highlighting historical performance patterns. It can identify pacing anomalies, simulate outcomes under different macro assumptions, and deliver more consistent quantitative benchmarking.

But AI has limits. Models are only as good as their training data and struggle to adapt to regime shifts such as the post-2022 interest rate environment. They lack the interpretive ability to address tax-legal structures and cannot evaluate soft factors like team culture. Moreover, reliance on the same datasets across LPs and GPs risks convergence and correlation, eroding diversification.

## Authority Allocation and Override Protocols

A hybrid governance model should explicitly distinguish between what is automated, what requires AI-assisted validation, and what must remain human-led.

- **Automated:** routine filters (fund size, domicile, sector classification)
- **AI-assisted:** performance attribution, pacing models, and risk-adjusted projections—subject to human validation.
- **Human-led:** qualitative judgments on governance quality, cultural alignment, and fiduciary-sensitive tax/legal issues.

## Override Authority:

- Resting authority lies with the committee chair, supported by the chair or lead independent director.
- If the chair is unavailable or conflicted, authority transfers to the v chair.
- The committee collectively may override a chair's decision not to c with a two-thirds supermajority.
- All overrides must be documented in minutes with a rationale tied fiduciary considerations.

## AI Vendor Governance

Where committees employ external AI tools, vendor oversight is critical. Reliance on opaque or conflicted vendors introduces risks that cannot be ignored.

Committees should require vendors to:

- Disclose methodology, including training data sources, model architecture, and validation approach.
- Demonstrate independence—vendors advising GPs in the same str should be prohibited.
- Submit to annual reviews of data governance, privacy, and cybersecurity protocols.
- Provide sufficient transparency to enable committees to assess the reliability.

Boards must retain authority to discontinue vendor relationships if quality independence deteriorates.

## Tax Integration

AI models often optimize for gross IRR, yet fiduciary responsibility requires focus on net, after-tax returns. OECD and IMF research shows that cross-border structures can materially affect outcomes—but the impact varies widely.

- **Timeframe:** Drag may appear annually or across a fund's life.
- **Asset class:** More acute in private equity and infrastructure; less relevant in tax-sheltered real estate vehicles.
- **LP domicile:** US tax-exempt endowments may face negligible exposure; taxable European LPs investing in US PE funds may face withholding on carried interest treatment.
- **Fund structure:** Delaware LPs, Cayman exempted funds, and Luxembourg SCSps each create different exposure pathways.

**Board Protocol:** All references to “material” tax drag must be contextualized. Committees should state clearly that “cross-border structures can materially reduce investor outcomes, with impacts varying widely based on LP tax status, fund domicile, and treaty protections. External tax counsel should assess exposure case by case.”

To operationalize, committees should adopt one of three Tax Optimization Layer models:

- **Minimal:** annual external tax review of AI-recommended allocation
- **Standard:** quarterly external counsel review with formal sign-off.
- **Comprehensive:** in-house tax specialist with veto authority.

## Manager Evaluation

Nearly all GPs today use some form of data analytics. The governance challenge is to evaluate where a manager sits along the AI adoption spectrum.

- **Full-stack adopters:** efficient but at risk of clustering.
- **Partial adopters:** balance efficiency with judgment.
- **Traditional-leaning:** potentially resource-constrained but may deliver uncorrelated exposure.

Committees should diversify across this spectrum rather than default to a binary view of “AI vs. non-AI.”

## Correlation Risk

The Bank for International Settlements has cautioned that reliance on machine learning models may create systemic convergence. In practice, we face data limitations that make direct measurement difficult. Most portfolio level transparency is annual, and vendor disclosure is often considered proprietary.

Boards should acknowledge these constraints and apply correlation monitoring primarily to **new commitments forward**. For now, practical proxies include sector concentration, geography, and fund size cluster

Over time, as transparency improves, committees may integrate statistics overlap metrics.

## Scenario Planning and Liability

Committees should prepare for scenarios where AI introduces governance stress. Model failure under new regimes, vendor herding, or data poisoning must be considered. Regulatory conflicts may emerge, as OECD BEPS prioritizes tax collection while SEC fiduciary standards prioritize investor protection.

Liability is a critical consideration.

- **When overriding AI:** Defensible rationale must be documented, ideally supported by independent validation.
- **When following AI,** Documentation should include awareness of AI limitations, vendor methodology inquiries, and reliance on established SEC guidance (IA Release No. 5248) and DOL fiduciary commentary.

Boards must recognize that fiduciary duty cannot be delegated to algorithms. Documentation and inquiry are the safe harbour.

## Limited Partner Transparency

Governance does not end with internal decision-making. LPs increasingly expect transparency on the role of AI in manager selection. Committees should adopt a communication strategy that includes:

- Annual disclosure of AI's role in GP selection during LP meetings.

- Attribution analysis distinguishing AI-recommended allocations from committee-modified ones.
- Incident reporting if an AI model failure materially affects outcome
- Availability of the governance framework upon LP request.

This transparency protects committees from reputational risk and manages LP expectations.

## Framework Review and Continuous Improvement

To avoid becoming a static compliance exercise, this framework must be subject to continuous review. Committees should:

- Compare AI-driven and human-overridden decisions annually, with attribution analysis.
- Commission a biennial external review by a consultant or peer institution
- Trigger reviews automatically following material underperformance or market regime shifts.
- Apply a three-year sunset clause, requiring revalidation or revision of framework.

## Conclusion

The GP selection process is shifting from relationship-driven discretion to hybrid human-AI governance. AI offers speed, scale, and consistency, but also brings convergence risk, tax blind spots, and liability challenges. By clarifying override authority, embedding tax optimization protocols, scrutinizing vendors, communicating transparently with LPs, and embedding review



mechanisms, boards can capture AI's benefits without abdicating fiduciary responsibility. This framework balances innovation with accountability ensuring that committees remain credible stewards of LP capital in an increasingly algorithmic financial landscape.

*This article explores ideas from “[Shaping the Decade: Governance, Sustainability, and AI 2026–2036](#),” a comprehensive guide for boards navigating the intersection of governance, technology, and stakeholder capitalism. For those ready to learn more about AI, this book offers practical strategies and insights.*

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