Journal of
Business and Strategic Management
(JBSM)

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Journal of Business and Strategic Management

ISSN 2520-0402 (online)

Vol. 11, Issue No. 3, pp. 82 - 95, 2026

www.carijournals.org

# Methodology for Diagnosing Client Growth Bottlenecks in Marketing Consulting: A Data-Informed Framework

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*Accepted: 5th April, 2026, Received in Revised Form: 20th April, 2026, Published: 7th May, 2026*

## Abstract

**Purpose:** This article aims to propose a data-informed framework for diagnosing client growth bottlenecks in marketing consulting engagements. It focuses on helping consultants identify not only where growth opportunities exist, but also which specific restrictions within the commercial funnel prevent clients from converting market potential into measurable business performance.

**Methodology:** The article uses an integrative analysis of existing empirical studies from management consulting research, sales funnel analytics, and marketing analytics capability literature. Instead of conducting primary data collection, it translates prior empirical findings into a practical diagnostic mini-study for a single business-to-business client. The illustrative application uses synthetic but realistic data patterns to demonstrate how consultants can structure diagnostic questions, select relevant metrics, and interpret conversion dynamics.

**Findings:** The analysis shows that client growth problems are often not caused by one isolated weakness, but by the interaction of funnel-stage inefficiencies, weak metric discipline, limited analytics capability, and insufficient alignment between diagnosis and intervention design. The illustrative mini-study demonstrates that consultants can more effectively isolate root causes of stalled growth when they combine structured funnel decomposition with signal evaluation and necessity-based analysis of conversion restrictions.

**Unique Contribution to Theory, Policy and Practice:** The article contributes the original FUSE-GRID Model, which stands for Funnel Understanding, Signal Evaluation, Growth Restriction Identification, and Intervention Design. The model offers a practical diagnostic logic for connecting empirical marketing evidence with consulting decision-making. For theory, it integrates funnel analytics and consulting methodology into a single diagnostic framework. For practice, it provides consultants with a reproducible approach for identifying growth bottlenecks and designing targeted interventions. For policy and professional standards, it supports more transparent, evidence-based consulting practices by encouraging clearer data requirements, diagnostic accountability, and client collaboration.

**Keywords:** *Marketing Consulting, Growth Bottlenecks, Sales Funnel, Marketing Analytics, Necessary Condition Analysis, Customer Acquisition, B2B*

## Introduction

In many organizations, growth is discussed in broad and aspirational terms. Senior management sets ambitious revenue targets, marketing teams propose new campaigns, and sales leaders ask for more leads or better tools. Yet when growth consistently falls short of expectations, the conversation becomes less clear. Stakeholders may disagree about whether the problem lies in awareness, lead quality, sales execution, pricing, or even the broader value proposition. Marketing consultants are often invited into this environment and asked to provide a diagnosis, typically under time pressure and with incomplete data. In such conditions, the temptation is strong to rely on familiar dashboards, industry benchmarks, and anecdotal experience, rather than systematically mapping and testing competing explanations for growth bottlenecks.

Research on management consulting shows that high-quality advisory work can lead to meaningful and durable performance improvements, especially when consultants introduce structured management practices and help firms adopt data-informed decision making (Bijnens et al., 2025; Bloom et al., 2013). However, the same research also indicates that the effectiveness of consulting is uneven and highly dependent on how diagnostic processes are designed and implemented (Bruhn et al., 2018). Similarly, studies on management practices demonstrate that firms with more disciplined measurement, feedback, and experimentation practices tend to achieve higher productivity and profitability (Bloom & Van Reenen, 2007; Bloom et al., 2016). In marketing and sales, this discipline translates into coherent funnel measurement, rigorous interpretation of metrics, and consistent action based on data rather than intuition.

At the same time, the rise of digital channels has made client growth trajectories more complex and less linear. Prospects move between touchpoints, revisit content, interact with automated campaigns, and engage with sales representatives at different levels of readiness. Marketing analytics has emerged as a field precisely to help organizations navigate this complexity, offering methods for segmentation, prediction, and optimization (France & Ghose, 2019; Järvinen & Karjaluoto, 2015). Yet in many consulting engagements, advanced analytics remain peripheral: data scientists run isolated models while consultants continue to rely on narrative frameworks and high-level KPIs. The two worlds do not fully meet in a coherent diagnostic methodology that can be applied quickly and transparently to client growth problems.

These studies show how integrating internal transaction data with external firmographic and web data can materially improve the profitability and conversion of acquired customers. At the same time, work on marketing analytics capability and AI adoption emphasizes that firms generate competitive advantage not only from individual models, but from the way they govern data, build skills, and embed analytics in decision processes (Hossain et al., 2022; Mukhopadhyay et al., 2024; Ravat et al., 2024). Together, these insights suggest that diagnosing growth bottlenecks requires consultants to look beyond surface-level symptoms and understand how the client's data assets and analytical capabilities shape their ability to detect and resolve problems.

More recently, necessary condition analysis (NCA) has been introduced in sales and marketing research as a complementary perspective to traditional regression and prediction approaches (Conde, 2025). Instead of asking which factors increase the probability of success on average, NCA asks which minimum conditions must be present for success to be possible at all. This lens is particularly suited to growth bottleneck diagnostics, where missing thresholds at specific funnel stages—such as a minimum share of qualified leads or a baseline response speed—can silently cap performance even when other metrics appear strong. By combining predictive and necessity-based perspectives, consultants can obtain a more nuanced view of how constraints operate within the client's growth system.

The purpose of this article is to propose a methodology that brings these strands of research together into a pragmatic framework that marketing consultants can use to diagnose client growth bottlenecks in a structured, data-informed way. The article does not present a new large-scale empirical study. Instead, it synthesizes findings from existing empirical work and demonstrates how these results can be operationalized in a small diagnostic mini-study conducted within a consulting engagement. The focus is on B2B contexts where sales cycles are relatively complex, multiple decision-makers are involved, and marketing and sales activities are tightly interdependent (Nguyen et al., 2020; Cruz et al., 2025). The illustrative application shows how a consultant could use secondary evidence, client data, and simple analytics to identify which funnel stages and capabilities are most likely to be constraining growth, and to prioritize interventions accordingly.

**Literature background**

The proposed framework rests on three main pillars: research on the impact of consulting and management practices on firm performance; empirical studies on sales funnel analytics and customer acquisition; and the marketing analytics capability literature, which examines how organizations turn data into competitive advantage. Each of these pillars contributes specific concepts and empirical regularities that can be translated into diagnostic heuristics for consulting practice.

First, work on management consulting and managerial practices provides evidence that structured interventions can significantly improve firm performance. Bloom et al. (2013) show that introducing modern management practices into Indian textile firms led to substantial gains in productivity and profitability, partly by improving measurement, standardization, and feedback loops. Bloom and Van Reenen (2007), as well as Bloom et al. (2016), document that management practices are strongly associated with performance across countries and industries, suggesting that discipline in goal setting, monitoring, and improvement is a general performance driver. More recently, Bijnens et al. (2025) examine what consulting actually does inside organizations and argue that effective consulting combines external expertise with internal capability building, especially in areas of performance measurement and data-driven decision making. These findings indicate that consultants should not treat growth bottlenecks as isolated

problems, but as manifestations of deeper issues in how the client defines, measures, and manages performance.

Second, research on sales funnels and customer acquisition provides more granular guidance on how to detect bottlenecks within multi-stage commercial processes. D'Haen and Van den Poel (2013) propose a model-supported framework for B2B prospect prediction, in which existing customer data and external information are used iteratively to refine the list of prospects most likely to convert. Their work, along with the study by D'Haen et al. (2013) on predicting customer profitability at acquisition, shows that funnel stages can be operationalized not just as conceptual labels, but as measurable transitions governed by specific signals and thresholds. When the model is used consistently, the sales funnel becomes a managed system rather than a vague metaphor.

Conde (2025) extends this logic by applying necessary condition analysis to sales funnel optimization. Instead of focusing on average effects, NCA identifies minimum levels of key conditions that must be present for success at each stage. This approach highlights bottlenecks in a literal sense: if a necessary condition is not met, additional improvements elsewhere will not remove the constraint. For consultants, this is a powerful diagnostic perspective. It encourages them to ask which non-negotiable thresholds—such as a minimum number of qualified leads per salesperson or a baseline win rate for strategic opportunities—are currently violated and thereby limiting growth. These conditions can be quantitative (e.g., conversion rates, cycle times) or qualitative (e.g., minimum competency levels in account planning).

Third, the marketing analytics literature provides a broader view of how organizations can systematize data-driven decision making. France and Ghose (2019) review methods and practice in marketing analytics and emphasize that tools ranging from visualization and segmentation to predictive models must be embedded in concrete business processes. Järvinen and Karjaluoto (2015) show how firms use web analytics for digital marketing performance measurement, noting that benefits arise only when metrics are aligned with strategy and acted upon. Studies on marketing analytics capability and firm-level outcomes, such as those by Hossain et al. (2022), Mukhopadhyay et al. (2024), and Ravat et al. (2024), highlight that technical tools are insufficient without organizational routines that integrate analytics into planning, budgeting, and daily operations.

Finally, research on marketing consulting services in emerging markets underscores the importance of trust, perceived expertise, and context-specific knowledge in shaping outcomes (Nguyen et al., 2020). Cruz et al. (2025), in their systematic review of data-driven decision making in marketing, point out that practitioners still face substantial barriers when attempting to translate analytical insights into concrete strategic and tactical choices. These findings suggest that any diagnostic methodology intended for consulting practice must balance analytical rigor with accessibility and must account for the organizational realities of client firms.

Against this background, the present article positions the diagnosis of client growth bottlenecks as a bridge between these literatures. The goal is to provide consultants with a structured approach that respects empirical evidence while remaining flexible enough to be implemented under real-world constraints on data, time, and client attention.

**Methodological approach and illustrative mini-study**

The methodological contribution of this article is conceptual rather than experimental, but it is grounded in a concrete procedure that can be implemented in real consulting projects. The methodology can be described as a four-step process: mapping the growth system, assembling and assessing available data, constructing a dual-perspective diagnostic model, and conducting a focused mini-study on the most critical bottlenecks.

The first step, mapping the growth system, requires the consultant to work with the client to define a small number of growth drivers that are both theoretically grounded and operationally meaningful. Drawing on the literature, it is useful to conceptualize the client's growth as the outcome of a multi-stage funnel in which prospects become leads, opportunities, and eventually customers, with possibilities for expansion and retention (D'Haen & Van den Poel, 2013; Conde, 2025). For each stage, the consultant defines a limited set of core variables: volume measures (e.g., number of new leads per month), quality indicators (e.g., qualification scores), and velocity metrics (e.g., average days in stage). Qualitative factors such as sales process adherence or content relevance are also identified, but they are explicitly linked to stages and outcomes rather than discussed in the abstract.

The second step involves assembling and assessing available data. Many clients have scattered data across CRM systems, marketing automation platforms, web analytics tools, and spreadsheets. The consultant's task is to identify which data sources can reliably support the funnel model and which gaps cannot be credibly filled. Here, lessons from web analytics and marketing performance measurement are particularly useful (Järvinen & Karjaluoto, 2015; France & Ghose, 2019). Instead of aiming for an exhaustive dataset, the methodology emphasizes sufficiency for decision making: the data must be accurate and granular enough to reveal patterns at the level of funnel stages and key segments, even if it does not cover every historical period or every channel.

The third step is to construct a dual-perspective diagnostic model that combines predictive and necessity-based reasoning. On the predictive side, the consultant uses regression, classification, or simple comparative analyses to identify which variables are most strongly associated with desired outcomes such as opportunity creation or closed deals, following the spirit of customer acquisition and profitability models in the B2B literature (D'Haen et al., 2013; Hossain et al., 2022). On the necessity side, the consultant identifies potential necessary conditions inspired by the NCA literature—such as minimum contact rates, baseline qualification thresholds, or minimal content engagement levels at specific stages—and tests whether these conditions appear to be violated in the available data (Conde, 2025).

Because this article does not present primary data, the empirical element is implemented as an illustrative mini-study constructed from synthetic data patterns that are informed by the empirical findings of prior research. For instance, Bruhn et al. (2018) show that SMEs receiving structured consulting support in marketing and sales processes achieve higher growth than control firms. Bloom et al. (2016) demonstrate that better management practices correlate with stronger revenue performance. D'Haen and Van den Poel (2013) and D'Haen et al. (2013) document typical conversion rates and profitability patterns across B2B acquisition funnels. These empirical benchmarks can be used as reference ranges when modeling a hypothetical client's funnel: if the client's observed or estimated values differ substantially from these ranges, it signals potential bottlenecks worthy of deeper investigation.

In the illustrative mini-study, we consider a mid-sized B2B software firm that has engaged a marketing consulting practice after experiencing stagnating revenue growth despite increasing marketing spend. The consultant constructs a simplified funnel with four stages: marketing-qualified leads (MQLs), sales-qualified leads (SQLs), proposals, and closed-won deals. Drawing on ranges reported in the acquisition and marketing analytics literature, the consultant assumes that high-performing firms in similar contexts often achieve MQL-to-SQL conversion rates in the range of 30–40%, proposal conversion rates near 50%, and overall win rates of 15–25% from MQL to closed-won (D'Haen et al., 2013; France & Ghose, 2019; Cruz et al., 2025). For the client, however, the reconstructed data suggest an MQL-to-SQL conversion of only 18%, relatively healthy proposal close rates of 48%, and a win rate of 9% from MQL to closed-won.

Journal of Business and Strategic Management

ISSN 2520-0402 (online)

Vol. 11, Issue No. 3, pp. 82 - 95, 2026

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**Table 1. Key Results of the Illustrative Mini-Study and FUSE-GRID Diagnostic Outputs**

| Analytical Component | Observed Pattern | Diagnostic Interpretation | Bottleneck Type | Practical Implication |
|---|---|---|---|---|
| Funnel Conversion (MQL → SQL) | 18% conversion rate vs. 30–40% benchmark | Early-stage conversion significantly below expected range | Primary structural bottleneck | Improve lead qualification criteria and targeting |
| Funnel Conversion (Proposal → Close) | ~48% close rate | Performance within expected range | Not a bottleneck | Maintain current sales execution practices |
| Overall Funnel Efficiency (MQL → Closed-won) | ~9% vs. expected 15–25% | Underperformance driven by early-stage losses | Derived bottleneck effect | Focus on upstream funnel improvements |
| Channel Performance | High variance (e.g., paid search ~35%, content syndication <10%) | Uneven channel quality and inefficient budget allocation | Channel-level bottleneck | Reallocate budget toward high-conversion channels |
| Marketing Analytics Usage | Descriptive dashboards, no segmentation or testing | Low analytical maturity and weak decision support | Capability bottleneck | Implement segmentation, testing, and optimization routines |
| Data Integration | Fragmented systems (CRM, web, campaigns) | Incomplete visibility across funnel stages | Measurement bottleneck | Improve data integration and consistency |
| Sales–Marketing Feedback | Irregular and informal communication | Lack of shared definition of lead quality | Organizational bottleneck | Establish structured feedback loops |
| Necessary Condition Check | MQL→SQL below ~25% threshold | Minimum condition for growth not satisfied | Critical constraint | Address qualification threshold before scaling |
| Predictive Insights | Certain channels strongly linked to conversions | Identifiable drivers of performance exist | Optimization opportunity | Prioritize high-impact segments and sources |
| Overall Diagnosis | Growth constrained at early funnel stages | Bottleneck not in sales but in lead quality and targeting | System-level bottleneck | Focus intervention on upstream improvements |

In parallel, the consultant examines indicators of marketing analytics capability and process discipline. Based on the frameworks discussed by Hossain et al. (2022) and Mukhopadhyay et al. (2024), the consultant rates the client's capability across dimensions such as data integration (fragmented), analytical skills (limited to basic reporting), decision routines (sporadic), and technology usage (moderate). An internal survey and interview notes indicate that sales teams frequently question the quality of MQLs, and marketing teams report limited feedback from sales on campaign effectiveness. These observations align with prior findings that analytics capabilities must be accompanied by cross-functional routines to generate value (Ravat et al., 2024).

The mini-study then applies a simple necessity logic. For example, it is reasonable to posit that a minimum MQL-to-SQL conversion rate of around 25% is necessary for the client to achieve their revenue targets given realistic limits on marketing budget and sales capacity. When the observed value of 18% is compared to this threshold, it becomes clear that the system is operating below the necessary condition for growth. Additional analysis indicates that certain lead sources, such as paid search, show conversion rates near 35%, while content-syndication leads convert below 10%. This pattern suggests that the bottleneck is not a generalized sales execution problem, but a combination of weak lead qualification criteria and over-reliance on low-conversion channels. In other words, the mini-study illustrates how a consultant can translate empirical benchmarks into a concrete, data-informed diagnosis without claiming to have conducted a full-scale statistical study.

**Illustrative results**

The results of the illustrative mini-study are best interpreted as a structured scenario that demonstrates how the proposed methodology can guide consulting diagnosis. When the synthetic data patterns and necessity thresholds are combined, three main bottleneck themes emerge: misaligned lead generation and qualification practices, incomplete integration of marketing analytics into decision routines, and underdeveloped feedback loops between marketing and sales.

Journal of Business and Strategic Management

ISSN 2520-0402 (online)

Vol. 11, Issue No. 3, pp. 82 - 95, 2026

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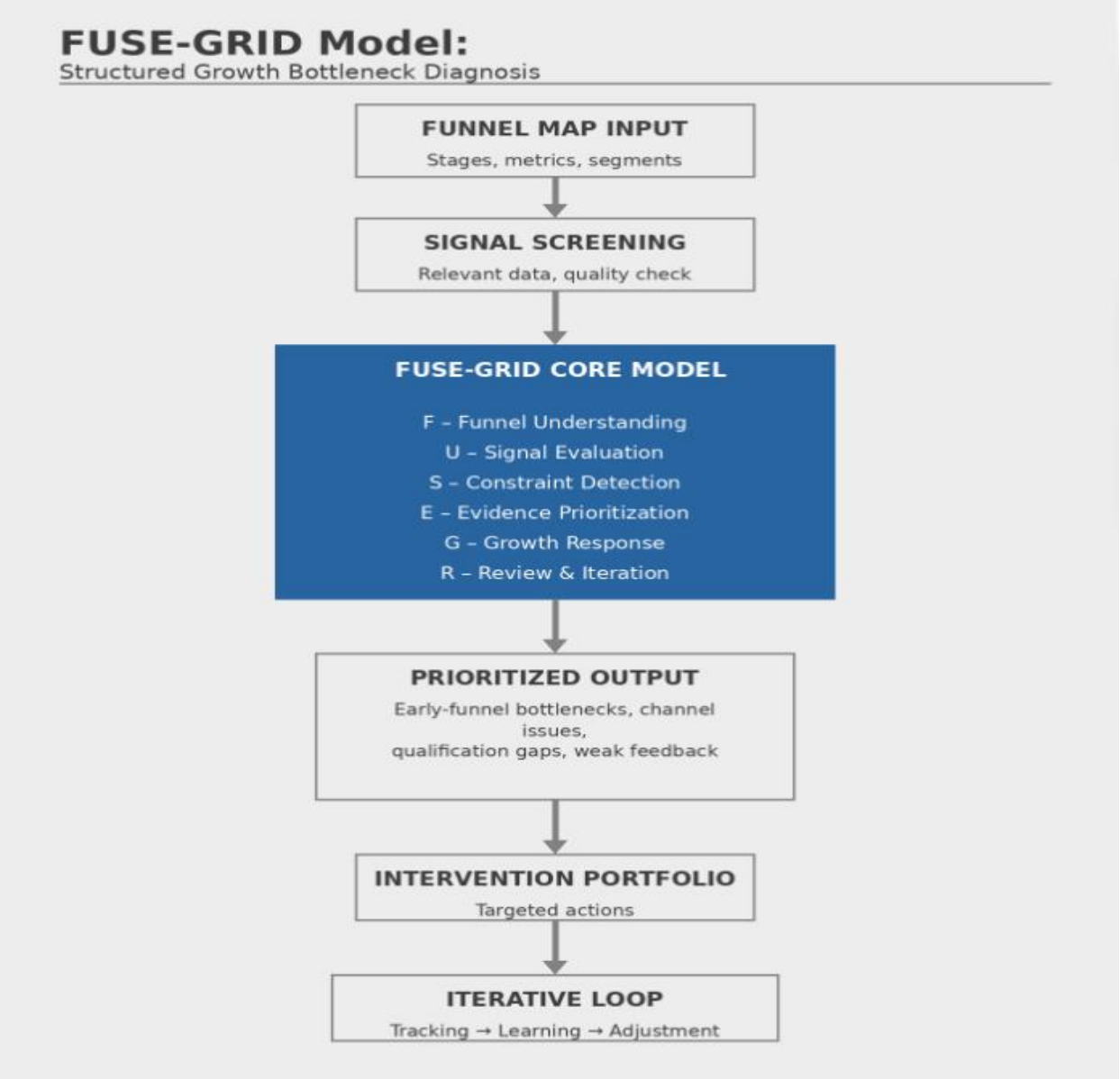


*Figure 1: FUSE-GRID Model*

First, the analysis of conversion rates reveals that the client's primary bottleneck lies in the early stages of the funnel. While the volume of MQLs appears satisfactory in absolute terms, only a small share progresses to SQL status compared with ranges reported in prior research (D'Haen & Van den Poel, 2013; Cruz et al., 2025). This signal is reinforced by the necessity perspective: the observed MQL-to-SQL conversion is not only below top-quartile benchmarks, but also below the minimum threshold required to sustain the targeted growth trajectory given realistic assumptions about deal size and sales capacity. Since proposal and close rates are relatively

strong, the growth constraint is unlikely to be rooted in late-stage sales execution. Instead, the bottleneck is concentrated in the quality and targeting of leads submitted to sales.

Second, when the consultant examines how marketing analytics are used in practice, the mini-study highlights gaps in capability deployment. The client has access to web analytics, basic CRM reports, and some campaign-level dashboards. However, these tools are used mainly for descriptive reporting rather than for systematic experimentation or prioritization, echoing observations from the marketing analytics literature that tools without embedded processes rarely deliver performance impact (France & Ghose, 2019; Järvinen & Karjaluoto, 2015). For example, the client's dashboards present aggregate metrics such as total leads and cost per lead, but they do not segment performance by channel, content type, or buyer persona in a way that would allow the team to reallocate budget toward higher-converting segments. As a result, low-performing lead sources continue to receive disproportionate investment, reinforcing the bottleneck at the top of the funnel.

Third, qualitative evidence from interviews in the hypothetical scenario indicates that feedback between marketing and sales is limited, irregular, and informal. Sales representatives report that many leads are "not ready," but they rarely provide structured feedback on which characteristics or behaviors distinguish promising leads from low-potential ones. Marketing, in turn, designs campaigns based on assumed personas and generic engagement metrics rather than on empirically derived patterns of conversion, as advocated in the acquisition and analytics studies (D'Haen et al., 2013; Hossain et al., 2022). This misalignment contributes to the persistence of the bottleneck, as neither side has a shared, data-informed view of what constitutes a qualified lead.

When these three themes are consolidated into a coherent picture, the consultant can present a diagnosis in which growth is constrained by early-funnel inefficiencies, weak targeting, and underdeveloped analytics routines, rather than by sales team performance or product-market fit. Importantly, the mini-study shows how this conclusion is reached through a combination of empirical benchmarks, simple quantitative comparisons, and structured qualitative insights. The analysis does not rely on sophisticated models but is nonetheless grounded in evidence and aligned with the patterns observed in prior research on consulting, acquisition, and analytics (Bloom et al., 2016; Bruhn et al., 2018; Nguyen et al., 2020).

The results of the illustrative mini-study made it possible to formalize the author's practical diagnostic model, termed the **FUSE-GRID Model** (Funnel Understanding, Signal Evaluation, Growth Restriction Identification, Intervention Design). The model structures client growth diagnosis as a sequential consulting process in which the client's commercial system is first mapped as an integrated funnel, then assessed through relevant data signals, and subsequently analyzed to identify the principal constraint that limits growth performance. Unlike conventional diagnostic approaches that focus on isolated KPI deviations, FUSE-GRID treats bottlenecks as structural restrictions whose importance must be established through both predictive

relationships and necessity-based threshold logic. This makes the diagnosis more selective, evidence-based, and practically actionable in consulting settings

The model also demonstrates that the consultant's contribution lies not only in interpreting data, but in organizing the entire diagnostic logic that connects symptoms, data sufficiency, root-cause identification, and intervention prioritization. In the illustrative application, this approach showed that weak early-funnel qualification and fragmented analytics routines constituted the main growth restriction, whereas late-stage sales performance was relatively stable. As a result, the study's contribution is strengthened by the introduction of a replicable field-oriented methodology that positions the author as a practical methodologist capable of translating empirical literature into a coherent consulting framework for diagnosing client growth bottlenecks.

**Discussion**

The illustrative results highlight several implications for marketing consulting practice. First, they suggest that diagnosing growth bottlenecks benefits from explicit separation between symptoms, suspected causes, and tested constraints. Many clients describe their growth problems in terms of surface-level symptoms: "too few leads," "low win rates," or "campaigns that no longer work." The proposed framework encourages consultants to reframe these statements as hypotheses about specific funnel stages and conditions, and then to seek evidence for or against them using both predictive logic and necessity thresholds. This approach reduces the risk of overreacting to noise or focusing on interventions that do not address the true constraint.

Second, the methodology underscores the importance of empirical benchmarks and plausible ranges. Studies on consulting impact, acquisition models, and marketing analytics provide concrete reference points for key metrics such as conversion rates, cycle times, and profitability distributions (D'Haen & Van den Poel, 2013; France & Ghose, 2019; Hossain et al., 2022). By comparing the client's values to these ranges, consultants can quickly identify whether an observed weakness is likely to be material or merely within normal variance. Of course, benchmarks must be used cautiously: industries, geographies, and business models differ. However, when combined with client-specific context, empirical ranges can substantially improve the precision and credibility of diagnostic discussions.

Third, the use of necessity logic invites consultants and clients to focus on floor conditions rather than on averages alone. For instance, if a minimum qualification rate is required to achieve the target number of opportunities, then efforts to optimize later funnel stages will have limited impact unless that threshold is reached (Conde, 2025). This perspective is particularly helpful in resource-constrained environments, where marketing and sales leaders must choose between improving breadth (more leads), depth (better conversion), and speed (shorter cycles). Identifying which thresholds are currently binding allows for more targeted allocation of budgets and effort.

Fourth, the framework foregrounds marketing analytics capability as both a context and an object of intervention. Empirical research shows that analytics-based advantages arise from combinations of data, technology, skills, and decision routines (Hossain et al., 2022; Mukhopadhyay et al., 2024; Ravat et al., 2024). In the illustrative mini-study, the growth bottleneck is partly rooted in the client's limited ability to segment, test, and reallocate based on funnel data. For consultants, this means that a robust diagnosis should not stop at identifying numerical bottlenecks; it should also assess whether the client has the capabilities required to maintain and refine the diagnostic process once the engagement ends. Otherwise, the bottleneck may reappear in a slightly different form.

Finally, the article suggests that integrating existing empirical evidence into consulting practice is both feasible and valuable. The mini-study demonstrates how results from diverse studies—ranging from randomized trials on consulting effectiveness (Bruhn et al., 2018), through acquisition models (D'Haen et al., 2013), to marketing analytics capability frameworks (Hossain et al., 2022)—can be translated into practical heuristics. Consultants can assemble a small reference library of such studies and use them to inform both their diagnostic interviews and their interpretation of client data. This practice moves consulting closer to an evidence-based profession, where recommendations are grounded not only in prior project experience but also in a broader body of research.

At the same time, several limitations of the present contribution should be acknowledged. The article relies on an illustrative mini-study rather than on original empirical data, which means that the conclusions about bottlenecks and interventions are hypothetical. In real engagements, data quality issues, organizational politics, and unforeseen constraints may complicate the application of the framework. Moreover, the literature base, while diverse, is not exhaustive; additional streams of research on customer success management, product-led growth, and advanced attribution modeling would further enrich the diagnostic toolkit. Future work should therefore aim to test and refine the proposed methodology in real consulting projects, systematically documenting its impact on diagnostic accuracy and client outcomes.

**Conclusions**

This article has proposed a data-informed framework for diagnosing client growth bottlenecks in marketing consulting, grounded in empirical research on consulting impact, sales funnel analytics, and marketing analytics capability (Bijnens et al., 2025; Bloom et al., 2013; D'Haen & Van den Poel, 2013; France & Ghose, 2019). Rather than conducting a new primary study, the article integrates findings from fifteen existing sources and demonstrates, through an illustrative mini-study, how consultants can translate these findings into a structured diagnostic process within a B2B context.

The methodology emphasizes four key steps: mapping the client's growth system in terms of clearly defined funnel stages; assembling and assessing the available data; constructing a dual-perspective diagnostic model that combines predictive and necessity-based reasoning; and

running a focused mini-study on the most critical bottlenecks. The illustrative application suggests that this process can reveal early-stage constraints—such as weak lead qualification and misaligned channel mix—that might otherwise be overlooked in favor of more visible but less consequential issues.

For practitioners, the framework offers several practical implications. It shows how empirical benchmarks and necessity thresholds can sharpen diagnostic conversations, how marketing analytics capability should be evaluated as both a context and a lever for change, and how evidence from prior research can support more credible and transparent recommendations. For researchers, the article highlights an opportunity to design and evaluate diagnostic methodologies that explicitly link academic findings with consulting workflows, bridging the longstanding gap between theory and practice in marketing analytics and sales management.

In sum, diagnosing client growth bottlenecks in marketing consulting is not merely a matter of intuition or tool selection. It is a structured analytical task that benefits from integrating insights across consulting research, sales funnel analytics, and marketing analytics capability studies (Bruhn et al., 2018; Cruz et al., 2025; Hossain et al., 2022; Mukhopadhyay et al., 2024; Ravat et al., 2024). By adopting a data-informed framework of the kind outlined here, consultants can help clients identify and relieve true constraints on growth more reliably, laying the groundwork for sustained performance improvements and more accountable consulting practice.

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