

Why do prediction models for prolonged air leak keep failing?

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Prolonged air leak (PAL) remains a common complication following pulmonary resection, extending hospital stays and increasing morbidity. Clinicians need reliable tools to identify patients at high risk for PAL to guide perioperative management and potentially improve outcomes. But the current market of predictive modeling is, frankly, a mess.

KEY TAKEAWAYS

- **The Pivot:** Despite numerous attempts, existing prediction models for prolonged air leak after pulmonary surgery do not offer sufficient clinical utility or translation potential.
- **The Data:** A systematic review of prediction models found consistent issues with quality, performance, and external validation, rendering them largely unusable in practice.
- **The Action:** Clinicians should remain skeptical of current PAL prediction models and continue to rely on established clinical judgment and intraoperative assessments.

Pulmonary surgery, while often life-saving, carries inherent risks. One of the more vexing complications is prolonged air leak, defined variably but generally as an air leak persisting beyond five to seven days post-surgery. This condition not only prolongs hospitalisation but also increases healthcare costs and patient discomfort. The drive to predict PAL has led to a proliferation of models, each claiming to offer insight into patient risk.¹

A systematic review published in *Frontiers in Oncology* by Du, Zhang, and Wang aimed to map, critically appraise, and synthesize the quality and performance of these existing prediction models. The authors focused on their clinical utility and potential for translation into practice. They sought to provide a comprehensive overview of the current state of PAL prediction, a task that, by its very nature, highlights the field's shortcomings.¹

The Problem with Prediction

The review identified a fundamental issue: while many models exist, few demonstrate the rigor required for clinical application. The authors found a consistent pattern of poor methodological quality and insufficient external validation across the studies. This means that while a model might perform well in the dataset it was built on, its ability to predict PAL in a new, independent patient population is often questionable.¹

Many models suffered from small sample sizes, leading to overfitting and an exaggerated sense of predictive power. The lack of standardized definitions for PAL itself also complicated comparisons between studies. One might define PAL as an air leak lasting 5 days, another 7, and a third 10, making it difficult to synthesize results or apply a single model across different clinical settings.¹

The review highlighted that the majority of models did not undergo robust external validation, a critical step for assessing generalizability. Without this, any model remains largely theoretical. Clinicians cannot trust a model that has not proven its worth in diverse patient cohorts. This is not merely an academic point; it directly impacts patient care.¹

The Numbers, or Lack Thereof

The systematic review did not present new numerical data from a trial, but rather critically assessed the reported performance of existing models. The consistent finding was that reported performance metrics, such as AUC values, often appeared inflated due to internal validation only. When external validation was attempted, performance frequently dropped significantly. This discrepancy shows the challenge in translating statistical models into practical tools.¹

The authors noted that many studies failed to report essential details about patient characteristics, surgical techniques, or postoperative management protocols. This omission makes it nearly impossible to replicate the models or understand their applicability to different surgical contexts. A model developed in a high-volume academic center, for example, may not perform similarly in a community hospital with different patient demographics or surgical practices.¹

But the problem extends beyond just PAL. The broader field of predictive modeling in medicine, particularly when dealing with complex, multifactorial outcomes, faces similar hurdles. The allure of a simple algorithm to guide complex decisions is strong, but the reality often falls short. For a deeper examination into how gut dysbiosis can connect systemic diseases like rheumatoid arthritis and periodontitis, which also rely on complex biological interactions, one might consider the oral-systemic link. This illustrates the persistent challenge of isolating predictive factors in intricate biological systems.¹

Where Current Models Fall Short

The primary limitation of the existing models, as identified by Du, Zhang, and Wang, is their poor clinical utility. A model might be statistically sound on paper, but if it does not offer actionable insights or significantly improve upon current clinical judgment, its value is negligible. Many models were too complex, requiring data points not routinely collected, or too simplistic, overlooking critical confounders.¹

The review also touched upon the issue of model interpretability. For a clinician to trust and use a prediction model, they must understand how it arrives at its conclusions. Black-box algorithms, while potentially accurate, often fail to gain traction in clinical practice because they do not provide a clear rationale for their predictions. This lack of transparency is a significant barrier to adoption.¹

The authors did not find any models that consistently demonstrated high predictive accuracy, generalizability, and clinical feasibility. This means that for now, the decision to retreat, resect, or extract in cases of persistent apical periodontitis, or indeed to manage any complex postoperative complication, still relies heavily on the clinician's experience and judgment, informed by patient-specific factors rather than an algorithm. The Oxford Handbook of Clinical Medicine (11th ed) remains a more reliable guide for general clinical decision-making than many of these predictive models.¹

Still, the systematic review itself, while critical, provided a valuable service by consolidating the evidence and clearly articulating the deficiencies. It serves as a call to action for future research to adopt more rigorous methodologies, including prospective validation studies and clear reporting standards. Without these improvements, the promise of predictive analytics in pulmonary surgery will remain largely unfulfilled. The challenges in developing robust predictive models are not unique to this field; similar issues arise in areas like predicting persistent asthma risk in children with

Alternaria allergy, where multifactorial influences complicate straightforward prediction.¹

The review did not examine specific interventions based on model predictions because the models themselves were deemed insufficient. This highlights a critical gap: without reliable prediction, targeted interventions are difficult to implement effectively. The focus must shift from simply creating models to creating models that are truly useful at the bedside.¹

The Path Forward for Predictive Tools

The authors concluded that despite the volume of research, no existing prediction model for prolonged air leak is ready for widespread clinical use. This is a stark assessment, but an honest one. Future efforts must prioritize external validation, transparent reporting, and a clear demonstration of clinical benefit over mere statistical significance.¹

The emphasis should be on developing models that are not only accurate but also practical and interpretable for the busy clinician. This might involve simpler models incorporating readily available clinical data, or more complex models that offer clear explanations for their predictions. The goal is to augment clinical decision-making, not replace it with an unproven black box.¹

The systematic review did not offer a clear timeline for when such robust models might emerge. The field needs a renewed commitment to methodological rigor and a willingness to accept that a model, no matter how elegant, is only as good as its real-world performance. Until then, clinicians should approach any new prediction model with a healthy dose of skepticism. The challenges of validating complex models are also evident in areas like understanding disparities in diabetic retinopathy, where socioeconomic factors add layers of complexity to prediction.¹

CLINICAL IMPLICATIONS

The systematic review by Du, Zhang, and Wang delivers a blunt message: the current crop of prediction models for prolonged air leak after pulmonary surgery is not ready for prime time. Clinicians should not waste time attempting to integrate these unvalidated tools into their practice. Relying on them risks misclassifying patients and potentially misguiding perioperative care, which is a disservice to both patients and the healthcare system.

For industry, this review signals a need for a fundamental shift in how predictive models are developed and validated. The emphasis must move beyond internal statistical performance to rigorous external validation across diverse populations. A model that performs well in one institution but fails elsewhere is not a product; it is a research curiosity. Investment in robust, multi-center prospective studies is essential.

Patients, meanwhile, continue to face the uncertainty of postoperative complications like prolonged air leak. While the promise of personalized risk prediction is appealing, the current evidence suggests that this promise remains largely unfulfilled. Until genuinely validated and clinically useful models emerge, patient counseling and management will continue to depend on established clinical expertise and shared decision-making, rather than algorithmic predictions.

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REFERENCES

1. Du N, Zhang M, Wang X. Prediction models for prolonged air leak after pulmonary surgery: a systematic review and critical appraisal. *Front Oncol.* 2026;42656249.

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