

Opportunity Validation Through Community-Based Experimentation

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Abstract

Entrepreneurial opportunities are frequently assessed before sufficient evidence exists to determine whether a proposed product, service, or intervention addresses a meaningful problem under realistic conditions. Conventional opportunity assessment can become overly dependent on founder assumptions, secondary market information, generalized surveys, or forecasts produced before prospective users have interacted meaningfully with an emerging solution. This study conceptualizes community-based experimentation as an opportunity-validation capability through which entrepreneurs and innovation teams test assumptions collaboratively with prospective users, local stakeholders, lead users, community organizations, and other actors situated within the intended context of use.

The framework integrates entrepreneurial opportunity development, effectuation, systematic experimentation, lead-user innovation, living-lab research, and stakeholder co-creation. Because no primary community or entrepreneurial dataset was supplied, a transparent simulation-based explanatory design was adopted. A synthetic dataset comprising 330 hypothetical opportunity-validation initiatives was generated using community-based experimentation maturity, evidence quality, stakeholder learning, opportunity–community fit, and opportunity-validation confidence as seven-point constructs. The simulated analysis showed positive associations between community-based experimentation and validation confidence ($r = 0.715$), opportunity–community fit and validation confidence ($r = 0.658$), stakeholder learning and validation confidence ($r = 0.641$), and evidence quality and validation confidence ($r = 0.639$).

The four explanatory constructs jointly accounted for approximately 67.3% of simulated variation in validation confidence. Mean validation confidence increased from 3.46 among very-low-experimentation profiles to 5.83 among very-high-experimentation profiles. These results are synthetic and do not constitute evidence from completed community experiments. The study argues that opportunity validation becomes more credible when communities function not merely as respondents from whom opinions are collected, but as situated participants whose behavior, knowledge, constraints, and adaptations help reveal whether an opportunity is desirable, workable, and contextually legitimate.

Keywords: opportunity validation, community-based experimentation, entrepreneurship, stakeholder learning, user innovation, living labs, opportunity development, entrepreneurial experimentation



1. Introduction

Entrepreneurial opportunities are rarely fully specified at the moment an idea first appears. Research by Dimov challenges the interpretation of opportunity recognition as a single insight attributable entirely to one entrepreneur, instead emphasizing that entrepreneurial ideas develop through continuing action, interpretation, and interaction with social contexts. Opportunity development therefore involves converting an initially plausible proposition into an increasingly evidence-supported understanding of a problem, a potential solution, intended beneficiaries or customers, implementation requirements, and conditions under which value can be created. This distinction is important because enthusiasm for an idea can precede credible evidence that the underlying opportunity exists in the form initially imagined. Experimentation provides one mechanism for reducing this uncertainty.

Thomke characterized experimentation as a fundamental innovation activity through which organizations generate information concerning technical and customer-related problems, while subsequent work on sequential testing demonstrated that experiments can be designed to learn progressively from earlier tests. Entrepreneurship research provides compatible reasoning. Sarasvathy's effectuation perspective emphasizes action under uncertainty using available means, affordable commitments, stakeholder interaction, and adaptation rather than assuming that entrepreneurs can predict an unknown future accurately in advance. Fisher's behavioral comparison of effectuation, causation, and bricolage further demonstrates that entrepreneurial action can involve multiple logics during venture development rather than relying on a single planning model.

The value of experimentation increases when those tests engage people who possess context-specific knowledge. Von Hippel's lead-user research established that users experiencing emerging needs ahead of the general market can contribute valuable information about both needs and potential solutions, and Lilien and colleagues' study at 3M demonstrated the organizational usefulness of systematically incorporating lead-user knowledge into innovation processes. Living-lab research extends this logic by embedding innovation activities within real-life or realistic settings involving users and other stakeholders. Studies of living labs describe environments in which users participate in exploration, co-creation, and evaluation rather than functioning exclusively as subjects of conventional market research.

Community-based experimentation develops this participatory logic further by treating local communities as sources of contextual evidence about actual needs, usage practices, institutional constraints, cultural expectations, accessibility conditions, implementation risks, and unintended consequences. This does not imply that a geographically or socially bounded community automatically represents a commercially scalable market. Rather, a community can function as an evidence-rich testing environment in which broad entrepreneurial assumptions are converted into observable hypotheses. Opportunities can then be refined, narrowed, redirected, or discontinued according to evidence rather than founder commitment alone. Camuffo and colleagues' randomized controlled study of 116 Italian startups provides particularly relevant evidence that adopting a more scientific approach to entrepreneurial decision making can alter the way entrepreneurs evaluate and discontinue ideas.

The present study consequently conceptualizes community-based experimentation as an opportunity-validation capability. The proposed framework links experimentation maturity with evidence quality, stakeholder learning, opportunity–community fit, and validation confidence. The



framework does not equate positive community reactions with validated business opportunities. Instead, validation requires converging evidence concerning problem relevance, behavioral engagement, practical feasibility, contextual legitimacy, and willingness to commit resources or continue participation. Because no primary field data were supplied, the empirical illustration uses synthetic data and is explicitly separated from real-world evidence.

2. Objectives of the Study

2.1 To Conceptualize Community-Based Experimentation as an Opportunity-Validation Capability

The first objective is to establish community-based experimentation as more than customer consultation. Conventional interviews and surveys can reveal preferences, perceptions, and stated intentions, but experiments allow entrepreneurial assumptions to encounter observable behavior. Community-based experimentation therefore involves deliberately designed, limited-scale interventions through which entrepreneurs introduce prototypes, service concepts, process changes, trial offers, temporary delivery mechanisms, or other testable manifestations of an emerging opportunity within a relevant community context.

This approach treats experimentation as structured learning rather than informal trial and error. Each experiment should begin with an explicit assumption, define observable evidence capable of supporting or challenging that assumption, establish reasonable boundaries concerning resources and participant risk, and conclude with a learning decision. The outcome may be continuation, modification, segmentation, further testing, or abandonment. Failure to confirm an assumption is therefore not automatically experimental failure; it can represent valuable evidence preventing larger misallocation of resources.

2.2 To Examine Evidence Quality and Opportunity–Community Fit

The second objective is to evaluate the proposition that stronger community experimentation improves the quality of evidence available for opportunity validation. Evidence quality refers to the relevance, behavioral grounding, diversity, consistency, and contextual credibility of information generated during experimentation. A positive reaction obtained from a highly supportive group should not receive the same evidentiary weight as repeated behavioral engagement across heterogeneous prospective users under realistic constraints.

Opportunity–community fit refers to the extent to which the proposed opportunity corresponds with actual problems, local practices, willingness to participate, accessible resources, institutional conditions, and stakeholder priorities. This construct intentionally extends beyond conventional product–market fit because socially embedded opportunities can fail even where functional demand exists if delivery assumptions conflict with local capabilities or legitimate community expectations.

2.3 To Assess Stakeholder Learning and Validation Confidence

The third objective is to examine stakeholder learning as a mechanism through which experimentation changes opportunity understanding. Community experiments can generate learning in both directions. Entrepreneurs learn about needs, behavior, implementation barriers, and alternative uses, while



participating stakeholders learn what the proposed solution can and cannot deliver. This reciprocal learning can transform the opportunity itself rather than merely evaluate a fixed proposition.

Validation confidence is therefore conceptualized as evidence-calibrated confidence, not founder optimism. It represents the degree to which available experimental evidence justifies continued investment in the opportunity in its present or revised form. High validation confidence should require multiple converging signals rather than repeated confirmation of the same assumption by the same participants.

3. Review of Literature

3.1 Entrepreneurial Opportunity Development and Experimentation

Entrepreneurial-opportunity research has progressively moved beyond interpretations of opportunities as complete objects waiting to be discovered. Dimov argues that opportunities develop as ideas are acted upon and shaped through social interaction, thereby emphasizing temporal development rather than isolated recognition. His related research also links opportunity intentions with learning processes, reinforcing the importance of interaction between individuals and situations during opportunity development. Alvarez and Barney's distinction between discovery and creation perspectives similarly demonstrates that some entrepreneurial settings may involve opportunities that emerge through entrepreneurial action rather than being fully specified before that action begins.

Experimentation provides a particularly relevant mechanism within this developmental view. Thomke's research defines experimentation as an information-generating component of innovation and shows that the economics and sequencing of experiments influence learning during product development. The entrepreneurial implication is that large irreversible commitments can sometimes be decomposed into smaller tests that reveal whether critical assumptions deserve continued investment. Camuffo, Cordova, Gambardella, and Spina provide unusually strong empirical evidence for this reasoning. Their randomized controlled trial examined entrepreneurs exposed to a scientific approach involving explicit theories, hypotheses, and tests, showing that structured experimentation influenced entrepreneurial decision processes. This evidence is especially relevant to opportunity validation because it suggests that disciplined experiments can alter not only what entrepreneurs learn but also how decisively they respond to unfavorable evidence.

3.2 User Knowledge, Communities, and Co-Created Innovation

Von Hippel's lead-user perspective demonstrates that important innovation knowledge can reside outside producer organizations. Lead users experience needs earlier or more intensely than the broader market and can therefore provide insights into emerging problems and potential solutions. Urban and von Hippel subsequently incorporated market research into lead-user analysis, while Lilien and colleagues evaluated the lead-user approach within 3M and found that it generated commercially promising concepts relative to conventional idea-generation approaches.

User innovation research also shows that communities can become active sources of experimentation and solution development. Morrison, Roberts, and von Hippel examined user innovation and sharing within a local community, while later work has continued to demonstrate the strategic relevance of user knowledge. Flowers and Meyer, for example, examined how entrepreneurs



can benefit from user knowledge in digital-service innovation. Such findings challenge approaches in which communities are treated purely as passive markets awaiting professionally designed solutions. Living labs provide a related methodological foundation. Dell'Era and Landoni position living labs between user-centered and participatory design, while Leminen's research examines user innovation and value co-creation in real-life living-lab environments. Ballon's research likewise characterizes living labs through stakeholder involvement in exploration, co-creation, and evaluation. These approaches provide an important conceptual bridge from laboratory testing to community-based opportunity validation because they demonstrate how realistic settings can expose needs, behaviors, and constraints that controlled evaluations may overlook.

3.3 From Community Feedback to Credible Opportunity Validation

The literature nevertheless suggests that participation and validation should not be treated as interchangeable concepts. Community involvement can improve understanding while still producing misleading conclusions when participants are unrepresentative, tests are weakly designed, questions encourage confirmation, or founders interpret enthusiasm as evidence of future adoption. The central methodological problem is therefore determining when community interaction produces information sufficiently credible to support entrepreneurial commitment.

Effectuation provides one response by emphasizing stakeholder commitments under uncertainty. Sarasvathy's framework treats entrepreneurial environments as contexts in which outcomes can be shaped through interaction with committed stakeholders rather than predicted entirely in advance. This perspective complements community experimentation because actual commitments—such as participation, repeat use, contribution of resources, referrals, co-development effort, or willingness to pay—can constitute stronger validation signals than expressions of general approval.

The research gap lies in integrating opportunity-development theory, experimentation, and community participation into one validation framework. Existing entrepreneurship research explains opportunity development and entrepreneurial experimentation, while user-innovation and living-lab research demonstrate the value of engaging users in real contexts. The present study connects these literatures by proposing that community-based experimentation improves opportunity validation when it generates stronger evidence, reciprocal learning, and context-specific understanding rather than merely greater stakeholder involvement.

4. Research Methodology

4.1 Research Design

A quantitative simulation-based explanatory design was adopted because no completed community experiments, entrepreneurial survey data, transaction records, participant observations, or independently assessed venture outcomes were supplied. The numerical analysis therefore serves as a methodological demonstration rather than an empirical claim concerning real entrepreneurs or communities.

The synthetic analytical dataset contained 330 hypothetical opportunity-validation initiatives. Each observation represents a conceptual experimentation process involving an entrepreneurial or innovation team and a relevant community. No real participant, venture, locality, or commercial transaction is represented.

Table 1: Operationalization of the Simulation Framework

Construct	Scale	Conceptual Operationalization	Analytical Role
Community-Based Experimentation	1–7	Maturity of hypothesis-driven, low-scale tests involving relevant community actors under realistic conditions	Primary predictor
Evidence Quality	1–7	Relevance, behavioral grounding, diversity, consistency, and contextual credibility of experimental evidence	Explanatory mechanism
Stakeholder Learning	1–7	Degree of reciprocal learning between innovators and community participants across experimentation cycles	Explanatory mechanism
Opportunity–Community Fit	1–7	Compatibility of the emerging opportunity with real community needs, practices, resources, constraints, and willingness to engage	Explanatory mechanism
Opportunity-Validation Confidence	1–7	Degree to which accumulated evidence justifies continued commitment to the opportunity or a revised version of it	Dependent variable

4.2 Synthetic Data Generation

Community-based experimentation maturity was generated as the principal explanatory variable. Evidence quality was modeled as positively associated with experimentation maturity because structured, repeated tests should produce more behaviorally grounded information than unsupported assumptions. Residual variation was retained because poor sampling or badly chosen measures can still generate weak evidence even within active experimentation programs.

Stakeholder learning was generated as a related but distinct construct. Opportunity–community fit was subsequently modeled from experimentation maturity, evidence quality, stakeholder learning, and random variation. This sequence reflects the theoretical assumption that opportunity fit is not fully known before interaction but becomes progressively clearer through situated testing and interpretation. Opportunity-validation confidence was generated from all four explanatory constructs plus stochastic error.

4.3 Statistical Procedures and Evaluation Criteria

The analytical procedure included descriptive statistics, Pearson correlations, maturity-group comparison, and ordinary least squares multiple regression. Opportunity-validation confidence was specified as the dependent variable.

The 330 synthetic initiatives were divided into five equal categories according to community-based experimentation maturity: very low, low, moderate, high, and very high, with 66 profiles in each

category. The analysis evaluated whether validation confidence increased progressively across these levels and whether evidence quality, stakeholder learning, and opportunity–community fit remained independently associated with validation confidence.

5. Analysis

5.1 Relationships Among Community Experimentation Constructs

The simulated dataset produced a mean community-based experimentation score of 4.16, evidence-quality score of 3.83, stakeholder-learning score of 3.70, opportunity–community-fit score of 4.24, and opportunity-validation-confidence score of 4.66.

These relationships illustrate the theoretical distinction between testing activity and validation itself. Conducting more experiments does not automatically validate an opportunity. Experimentation is useful when it improves the quality of available evidence, produces substantive learning, and reveals stronger correspondence between the emerging proposition and the context in which it is intended to operate.

Table 2: Simulated Outcomes by Community-Based Experimentation Maturity

Experimentation Maturity	Synthetic Initiatives (n)	Mean Experimentation	Mean Evidence Quality	Mean Stakeholder Learning	Mean Opportunity – Community Fit	Mean Validation Confidence
Very Low	66	2.40	2.83	2.82	3.26	3.46
Low	66	3.53	3.43	3.30	3.85	4.17
Moderate	66	4.25	3.90	3.72	4.23	4.59
High	66	4.84	4.17	4.12	4.65	5.25
Very High	66	5.76	4.82	4.54	5.22	5.83

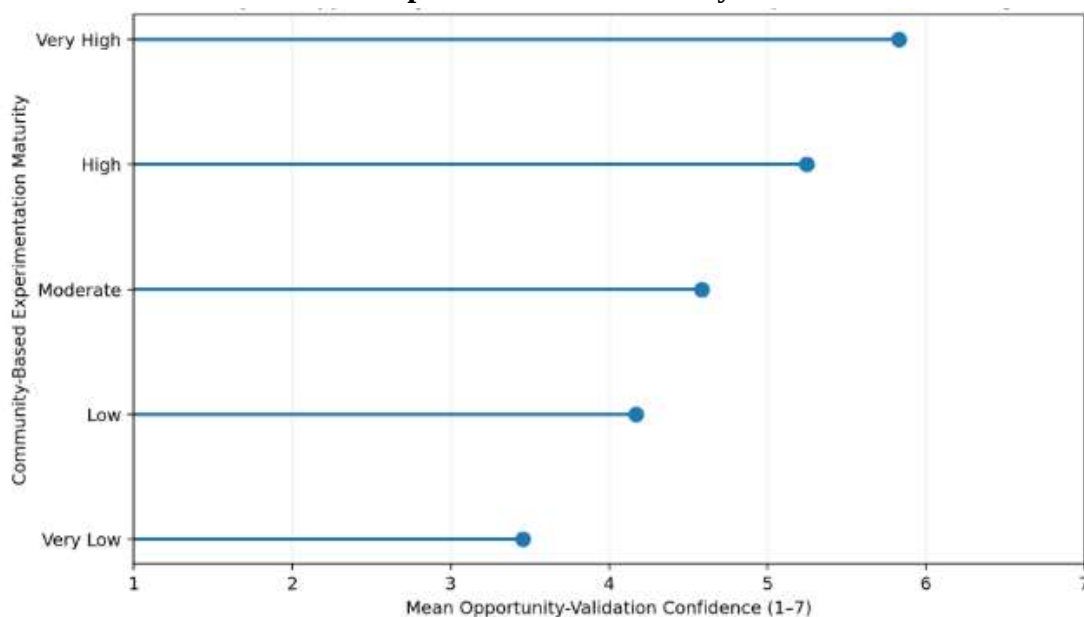
5.2 Experimentation Maturity and Validation Confidence

Mean validation confidence increased from 3.46 among very-low-maturity initiatives to 4.17 at low maturity, 4.59 at moderate maturity, 5.25 at high maturity, and 5.83 at very high maturity. The illustrative difference between the extreme categories was therefore approximately 2.37 points on the seven-point scale.

The progressive pattern suggests that opportunity confidence becomes more defensible as experimentation moves from occasional informal feedback toward structured cycles of hypothesis formulation, situated testing, behavioral observation, stakeholder interpretation, and revision. This

interpretation is consistent with Camuffo and colleagues' finding that scientific entrepreneurial decision processes can alter how founders evaluate entrepreneurial projects. It also aligns with Thomke's treatment of experiments as mechanisms for generating information rather than rituals designed merely to justify preexisting choices.

Graph 1: Simulated Opportunity-Validation Confidence Across Community-Based Experimentation Maturity



Interpretation: The lollipop graph shows a consistent upward progression in mean validation confidence as experimentation maturity increases. Higher experimentation maturity corresponds with progressively stronger evidence-supported confidence in the opportunity.

Key Finding: Within the simulated framework, mature community-based experimentation is associated with substantially stronger opportunity-validation confidence than low or informal experimentation.

5.3 Multivariate Validation Model

Multiple regression was used to determine whether the four theoretical constructs retained independent relationships with opportunity-validation confidence. The model produced an R^2 of 0.673 and an adjusted R^2 of 0.669, indicating that approximately 67.3% of the simulated variance in validation confidence was jointly explained.

Stakeholder learning produced an unstandardized coefficient of approximately 0.289, evidence quality approximately 0.275, community-based experimentation approximately 0.243, and opportunity–community fit approximately 0.232. All coefficients remained positive within the synthetic model.

The multivariate pattern reinforces an important conceptual conclusion: the greatest value of community experimentation may lie not in experimentation frequency itself but in what experiments allow participants to learn. An organization can conduct numerous pilots without improving validation if



experiments repeatedly expose the same population to the same proposition while leaving critical assumptions untouched.

6. Discussion

6.1 Community Participation Should Produce Evidence, Not Merely Endorsement

Community involvement is sometimes treated as inherently validating. This creates a methodological risk because participation can easily generate affirmative narratives without revealing whether an opportunity will survive realistic use conditions. Lead-user and user-innovation research is valuable precisely because it treats users as holders and creators of need-related knowledge rather than as sources of generic approval. Von Hippel's lead-user framework and Lilien and colleagues' subsequent empirical work demonstrate why carefully selected users can reveal emerging needs and potential solutions that conventional market research may not identify.

Community experimentation should therefore seek disconfirming as well as confirming evidence. An effective pilot asks which assumption could make the opportunity unworkable and then creates observable conditions under which that assumption can fail. If participants repeatedly abandon a prototype, require unsustainable assistance, refuse to commit resources, or adapt the solution in unexpected ways, those behaviors should be treated as substantive evidence rather than inconveniences to be explained away.

The methodological implication is that community enthusiasm and opportunity validity must remain separate constructs. A socially attractive idea may still lack economic feasibility, technological reliability, scalable delivery, or sufficient recurring demand. Conversely, apparently negative feedback can reveal a more valuable opportunity when experiments identify a different user group, problem definition, or application than originally expected.

6.2 Reciprocal Learning Converts Experiments Into Opportunity Development

A second contribution concerns stakeholder learning. Opportunity experimentation is often represented as a process in which entrepreneurs possess hypotheses and users provide answers. Such a model underestimates how participation can alter both parties' understanding. Living-lab research emphasizes exploration, co-creation, and evaluation with users in real-life environments, demonstrating that innovation can develop through interaction rather than one-directional information extraction.

Communities can reveal tacit information concerning routines, affordability, access, trust, institutional expectations, informal workarounds, seasonal variation, language, infrastructure, and social relationships. Some of these factors are difficult to identify through conventional market surveys because participants themselves may not articulate them until they encounter an actual prototype or service process. Experimentation externalizes assumptions and makes mismatches observable.

Entrepreneurs should therefore record not only experimental outcomes but also how the opportunity definition changed. A mature validation process maintains an evidence trail showing the original assumption, experiment design, observed behavior, stakeholder interpretation, decision, and revised hypothesis. Such documentation reduces hindsight bias and makes later investment decisions more accountable.

6.3 Validation Confidence Must Be Calibrated to the Strength of Evidence

The third contribution is the concept of evidence-calibrated validation confidence. Entrepreneurs operating under uncertainty must eventually act without complete information. Effectuation acknowledges this limitation by emphasizing control through available means, affordable commitments, and stakeholder participation rather than requiring accurate prediction of distant outcomes. However, uncertainty should not justify treating every entrepreneurial belief as equally credible.

Community-based experimentation can create a graduated evidence architecture. Early interviews may establish problem plausibility. Prototype engagement can test usability or behavioral interest. Repeated use can reveal continuing value. Willingness to contribute time, data, referrals, payments, space, or organizational support can provide increasingly costly signals of commitment. Each stage justifies a different degree of confidence.

This interpretation also creates a disciplined basis for discontinuation. Camuffo and colleagues' randomized controlled research is particularly important because scientific entrepreneurial decision making is valuable not only when it identifies promising opportunities but also when it helps entrepreneurs terminate weak ideas earlier. Community experimentation should therefore be considered successful when it prevents unjustified scaling just as much as when it confirms an opportunity.

7. Conclusion

This study conceptualized community-based experimentation as a capability for transforming entrepreneurial opportunities from assumption-rich ideas into progressively evidence-supported propositions. The central argument is that communities can contribute more to opportunity validation than generalized preference information. When experimentation is appropriately designed, prospective users and local stakeholders provide situated evidence concerning actual problems, behavioral engagement, usability, legitimacy, delivery barriers, adaptations, and willingness to make meaningful commitments.

The simulation-based analysis provided an internally consistent illustration of the framework. Community-based experimentation maturity was positively associated with opportunity-validation confidence, while evidence quality, stakeholder learning, and opportunity–community fit also contributed positively within the constructed model. Validation confidence increased considerably across experimentation-maturity categories. These numerical results are entirely synthetic and must not be presented as outcomes from completed community or entrepreneurial experiments.

The paper contributes to entrepreneurship and innovation research by integrating opportunity development, effectuation, systematic experimentation, user innovation, and living-lab thinking within a single community-based validation framework. The central implication is that opportunity validation should not ask only whether prospective users *like* an idea. It should examine what they actually do when confronted with an emerging solution, what contextual knowledge becomes visible through that interaction, which assumptions survive repeated tests, and whether stakeholders become willing to make increasingly consequential commitments.

Future empirical research should test the framework through longitudinal field experiments involving entrepreneurs, community enterprises, social innovators, public-service innovations, and locally embedded commercial ventures. Researchers should compare stated preferences with observable



behavior, examine how experimentation changes opportunity definitions over time, and assess whether stronger community validation predicts subsequent adoption, revenue, continuity, social legitimacy, or venture survival. Particular attention should be given to sampling bias, unequal participation, research ethics, community benefit, privacy, and the possibility that highly engaged participants may not represent broader markets. Community-based experimentation is therefore most valuable not when it confirms an entrepreneur's original idea, but when it creates a disciplined process through which reality is allowed to change that idea before costly scaling occurs.

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