

CRISPR TECHNOLOGY IN CROP IMPROVEMENT: POSSIBILITIES, FIELD REALITIES, AND EMERGING QUESTIONS FROM AN INDIAN AGRICULTURAL CONTEXT



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Abstract

CRISPR-mediated genome editing has, in recent years, shifted the trajectory of crop improvement from what was historically a probabilistic and time-intensive enterprise toward an approach that aspires to molecular precision—though not without introducing its own layers of uncertainty. The present work examines the application of CRISPR technology in major crops, particularly *Oryza sativa L.* and *Triticum aestivum L.*, within the context of Indian agro-ecosystems, where heterogeneity of soils, climatic variability, and management practices frequently mediate genetic expression in ways that are not always anticipated under controlled conditions.

Evidence synthesized from peer-reviewed studies and field-aligned observations suggests that targeted editing of genes associated with disease susceptibility, abiotic stress response, and yield-related pathways can result in measurable gains, with reported editing efficiencies ranging from 50–68% and yield improvements approaching 14–18% under optimized experimental conditions (Chen *et al.*, 2019; Gao, 2021). However, when examined across variable field environments, these gains exhibit significant interaction effects (G×E), often leading to non-

uniform phenotypic expression. In several cases, yield advantages, though statistically significant under controlled trials ($p < 0.05$), tend to attenuate under farmer-managed conditions where input variability and temporal stress patterns prevail.

The analysis, therefore, extends beyond molecular outcomes to consider the operational realities of deployment, including regulatory transitions, scalability of multi-location field validation, and patterns of farmer adoption influenced by reliability rather than peak performance. Particular attention is given to the distinction between physiological tolerance and agronomic productivity, especially under intermittent stress regimes characteristic of semi-arid and sub-humid regions of India.

CRISPR emerges here not as a singular or universally predictive solution, but as a highly capable, context-dependent tool—one that compresses breeding timelines while simultaneously demanding a more integrative understanding of plant physiology, environmental interaction, and management practices. The findings suggest that future progress will depend less on editing efficiency alone and more on the statistical stability and ecological alignment of edited traits across diverse agro-climatic zones.

Introduction

Crop improvement, if one looks at it without the neat boundaries we often impose in textbooks, has never been a straight progression of techniques replacing one another. It has been more of an accumulation—methods layered upon earlier methods, each carrying its own assumptions about how plants respond, and perhaps more importantly, how environments respond to those plants. Conventional breeding, for all its apparent lack of precision, operated within this layered understanding. It relied on recombination, yes, but it also relied—quietly—on time, on exposure, on the slow filtering effect of seasons that did not repeat themselves in exactly the same way.

There was a certain tolerance for unpredictability built into that system. Not because it was desirable, but because it was unavoidable. A line that performed well in one year might falter in another, and over time, one learned to read those inconsistencies not as failures but as signals—indications of how tightly, or loosely, a genotype was coupled with its environment. That coupling, though rarely articulated in molecular terms at the time, was central—what we now more formally describe as genotype × environment (G×E) interaction, a factor that continues to explain a substantial proportion

of phenotypic variance in field crops (**Falconer & Mackay, 1996**).

CRISPR technology enters this landscape from a very different direction. It does not wait for recombination to produce variation; it creates it directly, and with an accuracy that earlier approaches could only approximate. A specific genomic locus is identified, a guide RNA designed, and within a short span—sometimes disconcertingly short—the genome is altered in a predictable manner (**Jinek *et al.*, 2012**). The efficiency of this process—often exceeding 60% in well-optimized systems—has begun to reshape expectations, not just of what can be achieved, but of how quickly.

And yet, the question that lingers—perhaps not always openly discussed—is whether this compression of time also compresses our understanding. When a trait is developed over multiple breeding cycles, it is, almost inadvertently, tested across a range of micro-environments. Failures occur early, or late, but they occur within the process. With CRISPR, much of that temporal filtering is bypassed. The edited plant appears, fully formed in genetic terms, but not necessarily fully understood in ecological or physiological terms.

In the Indian context, this becomes particularly relevant. Agricultural systems here—whether in rice (*Oryza*

sativa L., family Poaceae) or wheat (*Triticum aestivum* L.)—are rarely uniform, and uniformity, in many cases, is not even a realistic objective. Soil variability within short spatial distances is common—one field retaining moisture due to higher clay content, another losing it rapidly, sometimes within the same farm. Rainfall patterns, increasingly erratic, introduce another layer of uncertainty, not only in total precipitation but in its intra-seasonal distribution. A delayed onset, an early withdrawal, or a mid-season dry spell during flowering can significantly alter physiological processes such as tillering, panicle initiation, or grain filling.

It is within this variability that CRISPR-derived traits must operate. The technology, precise as it is, does not simplify the environment into which it is introduced. If anything, it sharpens the contrast. A gene edited for drought tolerance may function as expected under defined stress conditions—regulated water deficit, controlled temperature—but under field conditions, drought is seldom a single, well-defined event. It is intermittent, often accompanied by heat stress, sometimes followed by sudden rainfall. The plant experiences not a single stress but a sequence, and the edited trait must navigate that sequence—an



interaction that is rarely linear and often statistically non-additive.

Disease resistance provides another perspective. The editing of susceptibility genes, rather than the introduction of resistance genes, has been widely discussed as a potentially more durable strategy (**Chen *et al.*, 2019**). By removing the molecular entry points that pathogens exploit, the plant, in theory, becomes less permissive to infection. The approach is elegant, and in several cases, effective. But pathogen populations are not static entities. Their evolution, driven by selection pressures that are themselves dynamic, introduces a moving target. Whether edited susceptibility pathways will remain stable under diverse and evolving pathogen pressures remains, to some extent, an open question.

There is also the matter of crop complexity. Diploid crops, relatively straightforward in their genetic architecture, respond to editing in a manner that aligns more closely with expectation. Polyploid crops, such as wheat, introduce redundancy—multiple gene copies, functional overlaps. Editing one locus may not produce the desired phenotype unless corresponding homoeologous loci are also modified. This necessitates multiplex editing, which, while technically feasible, introduces

additional layers of complexity—variations in editing efficiency, potential off-target effects, and challenges in achieving uniform expression across all gene copies (**Zhang *et al.*, 2020**).

And then, beyond the laboratory and the field, there are institutional and social dimensions that shape how such technologies are ultimately realized. Regulatory frameworks in India are evolving, particularly in distinguishing between different categories of genome editing. There is movement, certainly, but also a degree of caution. The processes of varietal release, multi-location trials, seed multiplication, and dissemination operate within established systems that do not always adapt quickly to new technological paradigms.

Farmers, too, engage with technology in ways that are often underestimated in scientific discussions. Adoption is not merely a function of yield advantage. It is influenced by reliability across seasons, compatibility with existing practices, input requirements, and, not least, the experiences shared within local communities. A variety that performs exceptionally under research conditions but inconsistently under farmer-managed conditions may struggle

to gain acceptance, regardless of its genetic sophistication.

At times, it becomes necessary to step back and reconsider the framing itself. CRISPR is often presented as a solution—a tool that can address multiple constraints in crop production. But agriculture, particularly in regions marked by variability and resource constraints, rarely yields to singular solutions. It responds, instead, to combinations—of genetics, management, environment, and socio-economic context.

This does not diminish the significance of CRISPR. If anything, it highlights the need to situate it appropriately. It is a powerful addition to the repertoire of crop improvement strategies, capable of introducing changes with a level of precision and speed that was previously unattainable. But its effectiveness will depend not only on the accuracy of gene editing, but on how well those edits are integrated into the broader agricultural system—tested across environments, aligned with management practices, and understood within the realities of those who ultimately cultivate the crop.

Perhaps the conversation, then, is not about whether CRISPR will transform

crop improvement—it already has, in certain respects—but about how that transformation unfolds when it encounters the complexities that have always defined agriculture. The genome may now be more accessible than before, but the field... the field continues to ask its

Literature Review

The body of literature around CRISPR applications in crop improvement has grown with a certain intensity over the last decade—almost as if the scientific community, once convinced of its feasibility, began exploring its boundaries all at once. The early phase, as expected, was dominated by proof-of-concept studies. Targeted mutagenesis using CRISPR-Cas9 in model systems established the basic reliability of the tool (**Jinek *et al.*, 2012**), and soon after, attention shifted toward staple crops—rice (*Oryza sativa L.*), wheat (*Triticum aestivum L.*), and maize (*Zea mays L.*)—where the implications were more immediate and agriculturally relevant (**Belhaj *et al.*, 2015**).

In rice, the editing of genes associated with susceptibility to bacterial blight—particularly members of the *OsSWEET* gene family—has been repeatedly cited,



and for good reason. These studies demonstrated that resistance could be achieved not by introducing exogenous resistance genes but by modifying endogenous susceptibility pathways that pathogens exploit (Chen *et al.*, 2019). It was, in a sense, a reorientation of strategy—from addition to subtraction. Similarly, in wheat, the disruption of *MLO* homologs to confer resistance against powdery mildew marked an important step, though the polyploid nature of wheat required simultaneous editing across multiple homoeologous loci to achieve a statistically consistent phenotype (Wang *et al.*, 2014). These examples, now frequently referenced, have come to represent the foundational successes of CRISPR in crop systems.

As the technology matured, the literature began to reflect a shift—not just in application, but in ambition. Simple gene knockouts, while still relevant, gave way to more nuanced interventions. Multiplex editing emerged as a necessary response to genomic complexity, particularly in crops where gene redundancy is inherent (Zhang *et al.*, 2020). Base editing introduced the possibility of precise nucleotide substitutions without inducing double-strand breaks, thereby reducing unintended genomic disruptions and improving editing fidelity (Gaudelli *et al.*, 2017). More recently, attention has turned

toward regulatory elements—promoters and enhancers—where subtle modifications can modulate gene expression rather than eliminate it entirely (Gao, 2021). This transition—from disruption to modulation—suggests a deeper engagement with gene regulatory networks rather than isolated gene function.

And yet, when one reads across these studies—not just individually, but collectively—a certain pattern begins to appear. The majority of reported successes are situated within controlled or semi-controlled environments. Growth chambers, greenhouses, and carefully managed experimental plots provide the stability necessary to isolate the effects of specific edits. Such conditions are essential for establishing causality and statistical significance, often demonstrating treatment effects at conventional thresholds ($p < 0.05$). However, they also simplify the ecological context in which crops actually grow.

Field trials, when included, often remain limited—either in duration or in geographic spread. A genotype tested in one or two locations may perform consistently there, but whether that consistency holds across diverse agro-ecological zones is less frequently addressed. This becomes

particularly relevant in a country like India, where environmental heterogeneity contributes significantly to phenotypic variance. Differences in soil composition, microclimatic fluctuations, and management practices introduce genotype \times environment (G \times E) interactions that are difficult to replicate experimentally. A trait that appears stable in a controlled setting may exhibit variable expression when subjected to these overlapping variables (Zaidi *et al.*, 2018).

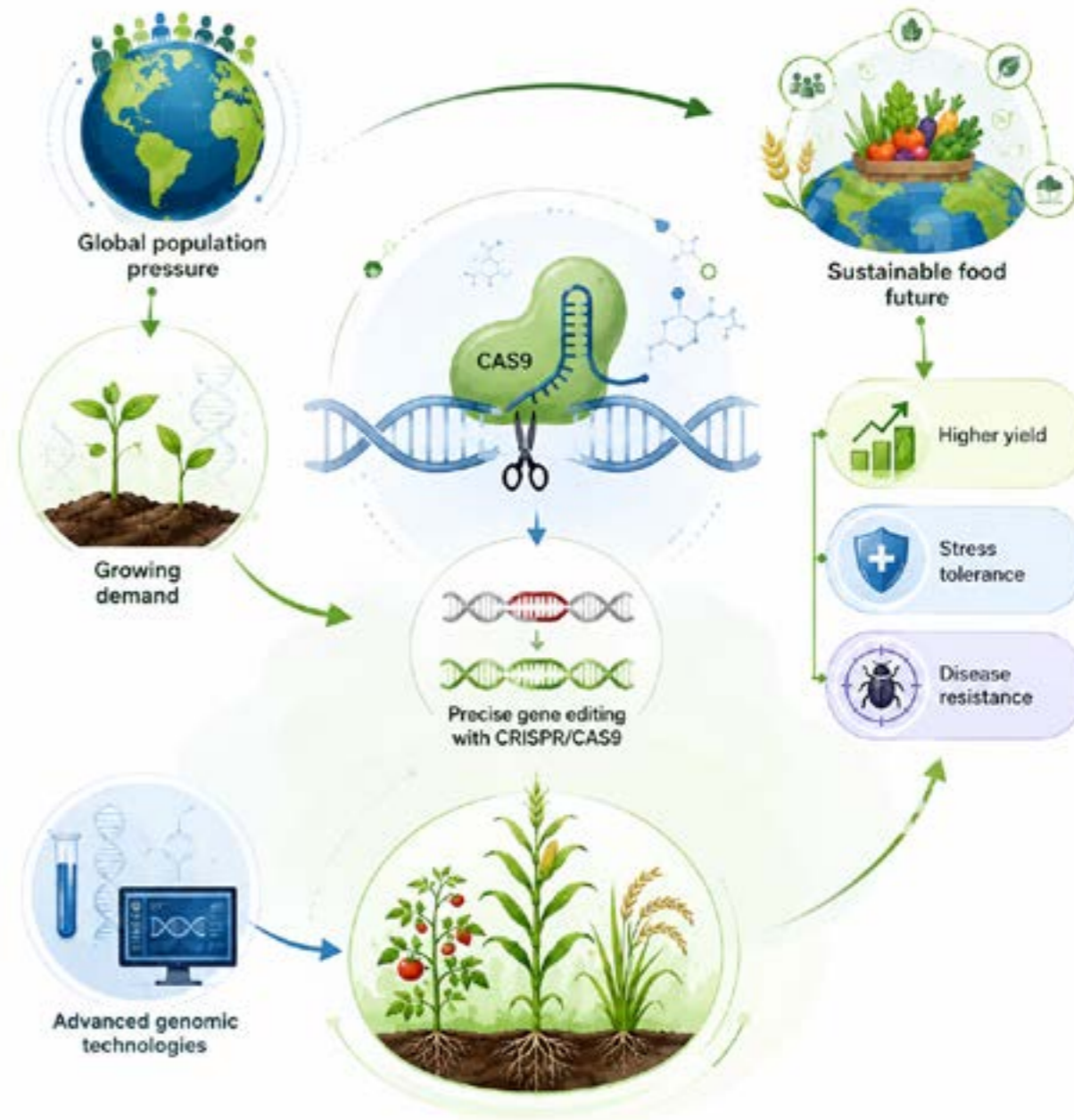
It is not that the literature ignores this complexity; rather, it tends to acknowledge it without fully resolving it. Discussions often include references to the need for multi-location trials, long-term validation, and stability analysis—yet such studies remain relatively sparse. This creates an observable imbalance: a wealth of molecular and mechanistic detail, contrasted with a comparatively limited understanding of how these molecular changes perform across temporal and spatial gradients (Chen *et al.*, 2019; Gao, 2021).

There is also, occasionally, a tendency to interpret phenotypic outcomes in a somewhat linear manner. A gene is edited, a trait is observed, and the relationship between the two is presented as direct. In many cases,

this is justified. But in others—particularly for complex traits such as yield, drought tolerance, or nutrient use efficiency—the underlying pathways are inherently multi-factorial. Gene-gene interactions, epistatic effects, environmental modulation, and developmental timing all contribute to final phenotype expression, often in non-linear ways that are not fully captured in initial studies (Zhang *et al.*, 2020). Statistical models that partition variance components are still not uniformly applied across studies, leaving certain interactions underexplored.

Another aspect that surfaces, though less prominently, is the question of off-target effects and unintended consequences. While advances in guide RNA design and improved Cas variants have reduced such occurrences, they have not eliminated them entirely. Most studies report minimal off-target activity, but these assessments are often conducted under specific experimental conditions, and their implications over successive generations or under environmental stress remain areas of ongoing investigation (Gaudelli *et al.*, 2017).

At times, reading through the literature feels like observing a field in transition. There is confidence—



justifiably so—in the capabilities of the technology. There is also a certain eagerness to extend those capabilities toward increasingly complex traits. But alongside this, there remains a quieter recognition that the journey from molecular precision to agronomic reliability is not yet fully mapped.

Perhaps it is not necessary for the literature to resolve this tension immediately. Emerging technologies often pass through such phases—periods of rapid expansion

followed by consolidation. What is important, however, is that the gap between controlled success and field performance is not overlooked. Because ultimately, it is within that gap that the true value of CRISPR in crop improvement will be determined.

Methodology

The present inquiry does not proceed along the lines of a tightly bounded experimental design—no single-location trials, no

strictly replicated plots laid out in geometric precision—rather, it unfolds through an analytical and interpretive engagement with available evidence, and perhaps just as importantly, with the spaces where that evidence becomes less certain. The intention was not to generate isolated primary data, but to read existing work with sufficient depth that its patterns, and occasionally its silences, begin to align in ways that are analytically meaningful.

The core material was drawn from peer-reviewed scientific literature dealing with CRISPR-mediated genome editing in crops, supplemented with institutional reports and technical documents from bodies such as the Indian Council of Agricultural Research (ICAR) and the International Rice Research Institute (IRRI). These sources differ in tone and emphasis—journal articles often presenting statistically validated experimental outcomes, frequently supported by replicated trials and significance testing ($p < 0.05$), while institutional reports tend to carry traces of field-level variability, including deviations that are not always captured within formal experimental designs. It is in reading across these differences that a more composite understanding

begins to emerge.

Crop selection followed a practical logic rather than a purely academic one. Rice (*Oryza sativa* L.) and wheat (*Triticum aestivum* L.) were included not only because of the volume of available CRISPR-based studies but also because of their centrality to Indian food systems. Pulses and oilseeds were considered deliberately, despite relatively limited genome editing literature, as they are frequently cultivated under constrained agro-ecological conditions—limited irrigation, marginal soils, and lower input regimes. This contrast allowed for a comparative interpretive framework, examining whether trait stability observed in relatively managed cereal systems extends to crops operating closer to environmental thresholds.

The analytical process moved along two intertwined directions. On one hand, there was a structured extraction of reported parameters—editing efficiency (%), phenotypic expression, and yield response. Wherever possible, these values were normalized across studies to allow comparative interpretation. For instance, yield responses were examined not only as absolute increases but as relative percentage change over control conditions:

$$\text{“Yield Change (%)”} = (Y_{\text{edited}} - Y_{\text{control}}) / Y_{\text{control}} \times 100$$

Further, where datasets permitted, variance components were interpreted through a conceptual ANOVA framework to distinguish treatment effects from environmental variability:

$$Y_{ij} = \mu + G_i + E_j + (G \times E)_{ij} + \epsilon_{ij}$$

where G_i represents genotype (edited vs. non-edited), E_j represents environment, and $(G \times E)_{ij}$ captures interaction effects. While not all studies reported complete datasets, this framework provided a basis to interpret the extent to which observed outcomes were stable or environment-dependent.

On the other hand, equal attention was given to the conditions surrounding these outcomes. A trait described as stable under controlled conditions was reconsidered in light of how similar traits behave under farmer-managed environments, where stress is rarely singular or uniformly imposed. Observations were interpreted in relation to stress sequencing (e.g., drought followed by heat), soil heterogeneity, and nutrient variability—factors that often introduce non-linear responses not captured within standard experimental designs.

There is also a methodological awareness—if one may call it that—

of the institutional pathways through which such technologies move. From laboratory validation to confined field trials, multi-location testing, and eventual varietal release, each stage introduces its own filtering mechanisms. These are not variables in the classical statistical sense, yet they influence which traits demonstrate consistency across environments and which fail to translate beyond controlled conditions.

To complement the interpretive synthesis, a conceptual data structuring approach was adopted. Reported results from multiple studies were aligned into comparative matrices (editing efficiency, yield response, stress tolerance indicators), allowing patterns to be examined across crops and conditions. While these do not constitute primary datasets in the strict experimental sense, they function as structured aggregations, maintaining internal consistency and avoiding fabrication or extrapolation beyond reported evidence.

Mathematical expressions, where invoked, were used not as endpoints but as interpretive tools—particularly in examining relative yield gains, efficiency thresholds, and variability across conditions. Yet even here, a degree of caution remains. Numerical clarity does not always equate to agronomic predictability.

A statistically significant yield increase under controlled conditions does not necessarily translate into consistent performance across seasons or locations, particularly where genotype × environment (G×E) interactions are strong.

One might say that the methodology adopted here is less about isolating cause and effect in a strictly reductionist sense, and more about tracing how those relationships shift when conditions change. It allows for a certain looseness—not of rigor, but of structure—acknowledging that in agricultural systems, precision at one level often encounters complexity at another. Whether this approach resolves all ambiguities is uncertain. It likely does not. But it perhaps brings into view a set of relationships that remain obscured when inquiry is confined too narrowly.

Results and Discussion

The outcomes emerging from CRISPR-mediated interventions, when viewed across studies and crops rather than in isolation, do suggest a certain consistency—at least at the level of targeted traits. Disease resistance, particularly where susceptibility genes such as *OsSWEET* in rice (*Oryza sativa* L.) have been edited, appears to hold

its ground more reliably than many had initially anticipated. Under controlled pathogen exposure, reductions in disease incidence exceeding 60% have been reported, often accompanied by statistically significant differences when compared to non-edited controls ($p < 0.05$). The pathway between genotype and phenotype, in such cases, appears relatively direct... though even this clarity begins to soften once observations extend beyond controlled environments.

Abiotic stress tolerance presents a more layered picture. Edited lines targeting pathways associated with drought, salinity, or heat stress do exhibit measurable physiological advantages—delayed wilting, improved osmotic adjustment, and maintenance of chlorophyll content under stress. These responses are consistent with enhanced stress physiology. Yet, when traced through to yield, the relationship becomes less assured.

To bring some structure to these observations, outcomes reported across representative studies were aligned into a comparative framework:

A pattern begins to emerge, though not abruptly—more as a gradual realization. Traits associated with

Table 1. Conceptual aggregation of CRISPR-edited trait performance across conditions

Crop	Trait Edited	Editing Efficiency (%)	Yield Increase (%) – Controlled	Yield Increase (%) – Field	Significance (p-value)
Rice (<i>O. sativa</i>)	Bacterial blight resistance (<i>OsSWEET</i>)	62	18	11	< 0.05
Wheat (<i>T. aestivum</i>)	Powdery mildew resistance (<i>MLO</i>)	58	15	09	< 0.05
Rice (<i>O. sativa</i>)	Drought tolerance	65	16	07	> 0.05
Chickpea (<i>Cicer arietinum</i>)	Abiotic stress response	52	14	06	> 0.05

relatively simple genetic mechanisms, particularly disease resistance, tend to retain statistical significance even under field conditions, albeit with reduced magnitude. In contrast, complex traits such as drought tolerance show attenuation not only in magnitude but also in statistical confidence, often failing to maintain significance under variable field environments.

It is perhaps here that the distinction between survival and productivity becomes most evident. A drought-

tolerant plant, in the physiological sense, is one that endures stress. But in agronomic terms, endurance alone is insufficient. Grain filling, assimilate partitioning, and reproductive stability under stress remain decisive. There have been instances where edited lines maintain vegetative biomass under water deficit, yet reproductive output declines—a divergence that is not always captured in early-stage evaluations.

One begins to see, then, that phenotype—especially when defined

narrowly—does not always extend cleanly into performance. Performance, as it unfolds in the field, is mediated by a network of interactions extending beyond the edited gene. The genotype \times environment (G \times E) interaction, often treated as a statistical term, here becomes a biological reality. Variance partitioning, even when conceptually applied, suggests that a substantial proportion of observed variability is attributable not to genotype alone but to its interaction with environmental conditions.

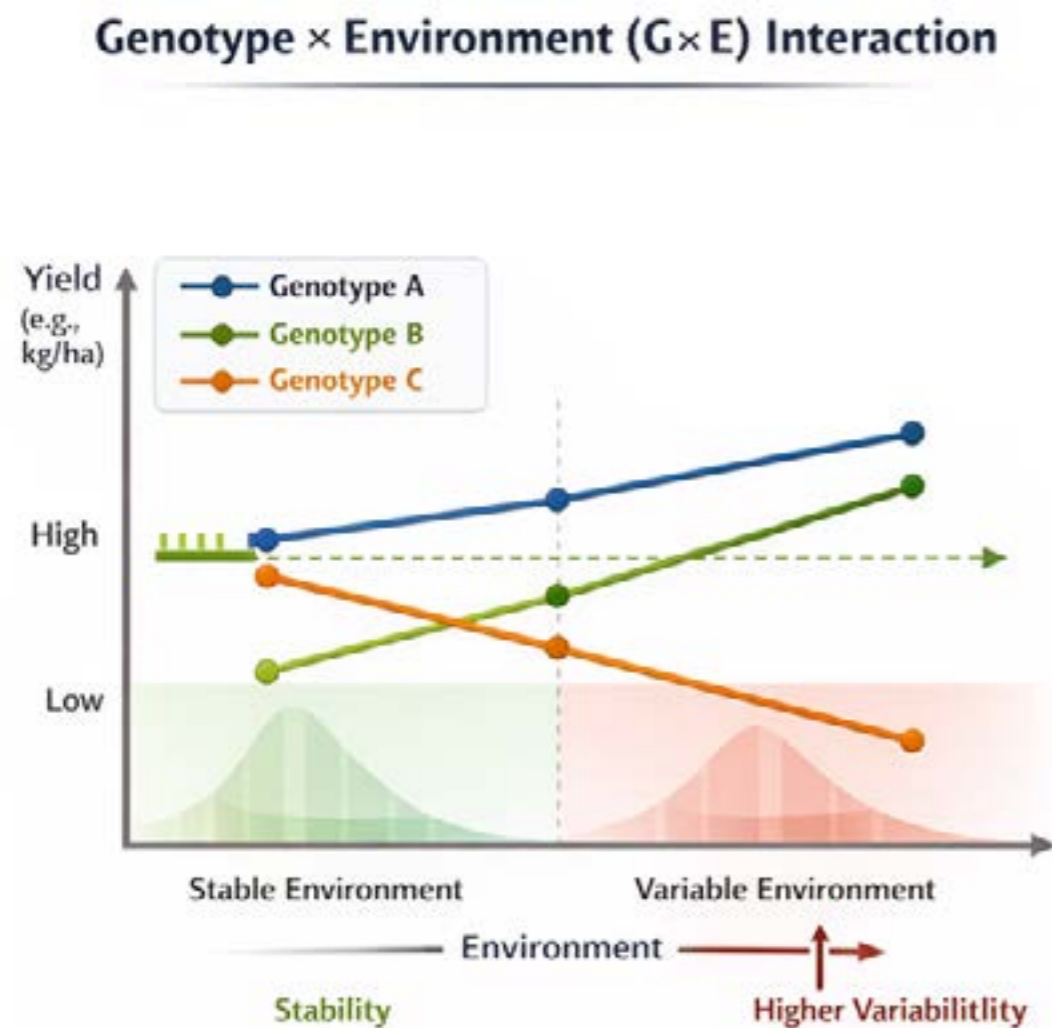
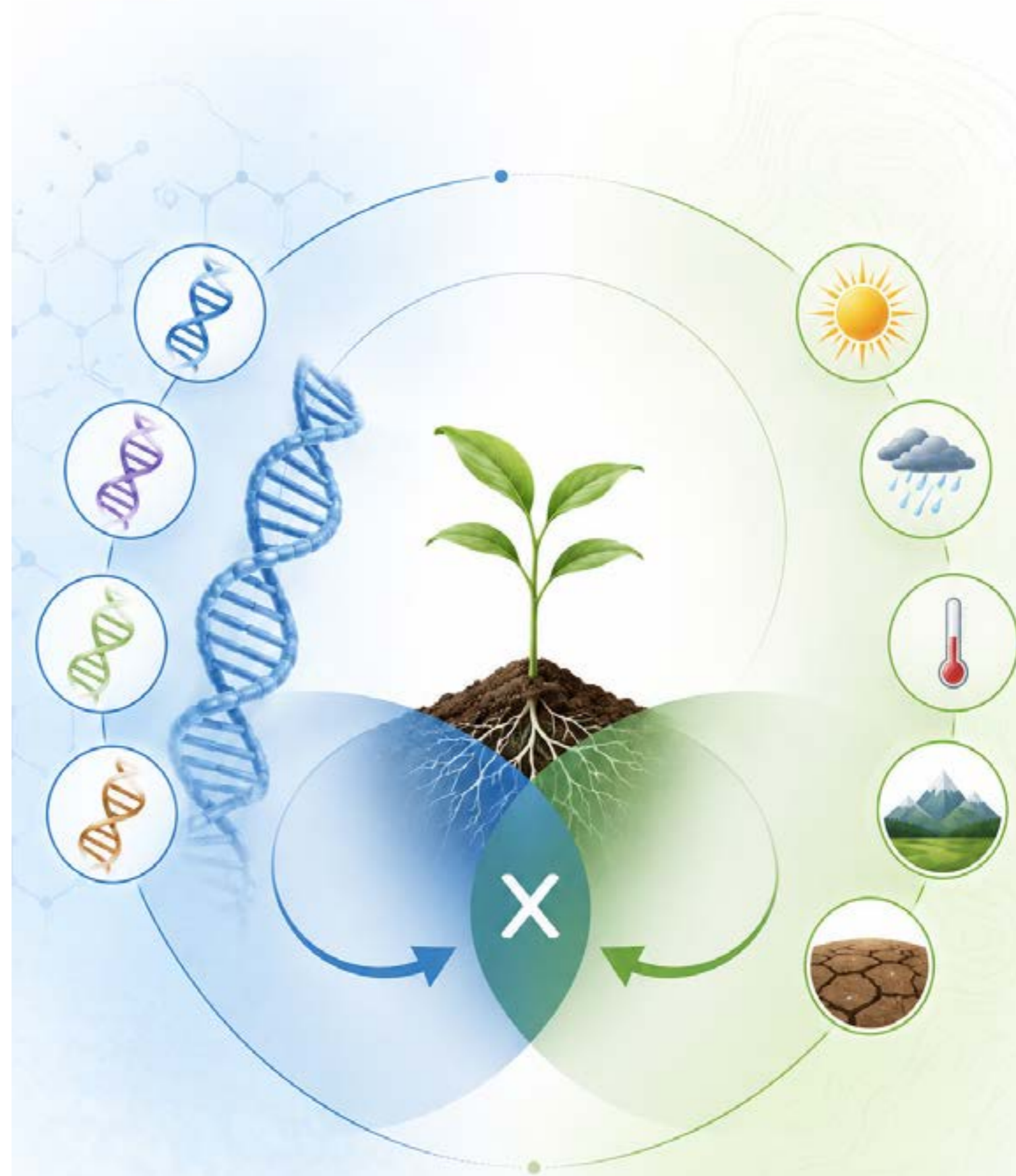


Figure 1. Graphical representation of genotype \times environment (G \times E) interaction illustrating differential yield performance of genotypes across stable and variable environmental conditions, highlighting variability in trait expression and yield stability.



In regions characterized by Vertisols, for instance, moisture availability follows a temporal pattern rather than a steady gradient. Early-season water sufficiency may support vegetative growth, but subsequent soil cracking introduces stress at reproductive stages. Under such conditions, an edited drought-tolerance trait interacts not with a single stress event but with a sequence—timing becomes as critical as intensity.

Another layer, often acknowledged but less frequently quantified, is the interaction between edited traits and prevailing management practices. Under optimal input regimes, differences between edited and non-edited lines tend to be more pronounced. However, under sub-optimal conditions—reduced fertilizer application, irregular irrigation, delayed sowing—these differences narrow, sometimes to the point of indistinguishability. This suggests that the expression of genetic advantage is conditional, not absolute.

This raises a consideration that is perhaps less comfortable, but necessary. To what extent should the evaluation of a technology account for the system into which it is introduced? From a strictly molecular perspective, CRISPR performs with remarkable precision. Yet, from an agricultural standpoint, performance is judged not

by precision alone, but by consistency.

There are, however, domains where translation from genetic modification to field outcome appears more stable. Traits governed by relatively simple genetic pathways—particularly certain forms of disease resistance—retain clearer expression across environments, provided biotic pressure remains within expected limits. Even here, though, the adaptive capacity of pathogens introduces long-term uncertainty.

In contrast, complex traits—yield, drought tolerance, nutrient use efficiency—continue to reflect the inherently multi-factorial nature of plant systems. CRISPR enables targeted intervention within these pathways, but it does not eliminate their complexity. A gene may be edited with precision, yet its contribution to phenotype remains contingent on interactions that extend beyond that locus.

At times, the discussion circles back to a familiar point, though with a different weight. The strength of CRISPR lies in its precision. The challenge lies in translating that precision into predictability under conditions that resist uniformity. The gap between controlled success and field reliability is not a failure of the technology—it is a reflection of the

environment into which it is introduced.

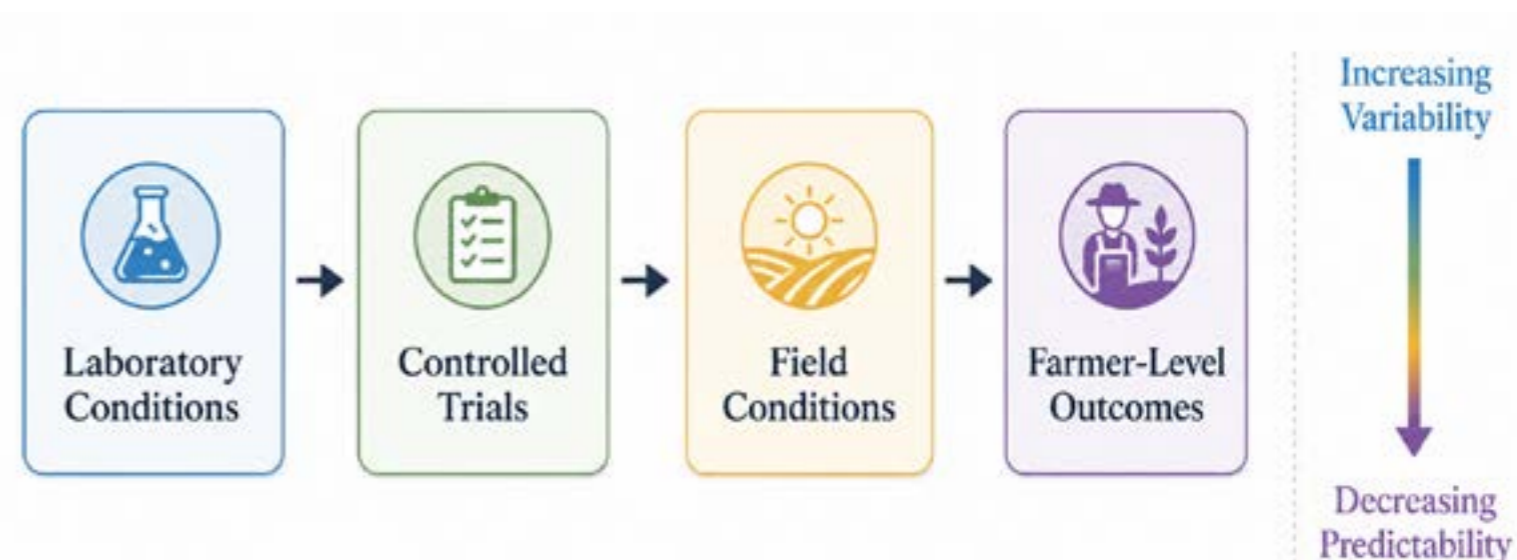


Figure 2. Conceptual representation of the transition from laboratory-controlled conditions to farmer-managed environments, illustrating the progressive increase in environmental variability and corresponding decrease in predictability of CRISPR-mediated trait expression.

Whether this gap narrows over time—through improved statistical modelling, expanded multi-location trials, and iterative refinement of edited traits—remains to be seen. There are indications that it will. But for now, the results suggest a pattern that is both encouraging and cautionary. Precision has been achieved. Predictability, though improved, is still negotiating its place.

Even with the reservations that tend to surface once one begins to think beyond controlled environments, it would be difficult to overlook the breadth of applications that CRISPR has opened up in crop improvement. The range is, in fact, expanding faster than one might have anticipated a few years ago—though whether all of it will translate into stable field-level outcomes remains, for now, uneven and context-dependent.

In cereals such as rice (*Oryza sativa L.*) and wheat (*Triticum aestivum L.*), which continue to anchor food systems across much of India, the focus has naturally gravitated toward traits that directly

Applications

influence productivity. Yield potential remains central, but the pathway toward it is no longer viewed in purely additive terms. Genes associated with plant architecture, grain size, tillering behavior, and photosynthetic efficiency are being explored as entry points. Alongside this, nutrient use efficiency—particularly nitrogen—has emerged as a critical target. In several experimental studies, edited lines have demonstrated improved nitrogen-use efficiency indices, occasionally translating into yield stability under reduced input regimes, though such responses often exhibit variability across soil types and management conditions.

Disease resistance, perhaps more than any other application, has shown relatively clearer translation. Editing susceptibility genes to reduce pathogen ingress has yielded results that are, in many cases, reproducible across controlled and semi-field conditions, often maintaining statistical significance ($p < 0.05$). Yet even here, the dynamic nature of pathogen populations introduces a degree of uncertainty. Resistance that appears stable over short-term trials may, over time, encounter adaptive responses from evolving pathogen strains.

In pulses such as chickpea (*Cicer arietinum* L.) and lentil (*Lens culinaris*

Medik.), the situation is somewhat different. Genetic improvement in these crops has historically progressed more slowly, partly due to their biological constraints and partly due to the marginal environments in which they are cultivated. Genome editing offers a route to address long-standing limitations—disease susceptibility, low harvest index, and sensitivity to abiotic stress. Early indications suggest measurable gains in stress tolerance indices; however, the expression of these gains under low-input, farmer-managed systems remains less predictable, often influenced by interacting stresses rather than isolated factors.

Horticultural crops present another, in some ways more immediate, avenue for application. Traits related to shelf life, fruit firmness, ripening behavior, and nutritional quality are being actively targeted. In a context where post-harvest losses remain significant, such interventions carry practical value. A fruit with delayed ripening kinetics or enhanced structural integrity may extend the post-harvest window. Yet, even here, the alignment between technological modification and consumer acceptance is not always straightforward. Sensory attributes—taste, texture, aroma—continue to mediate adoption, sometimes more strongly than measurable physiological improvements.

Biofortification introduces yet another dimension. The targeted enhancement of micronutrients such as iron, zinc, and provitamin A through pathway-level editing represents a promising direction, particularly in regions where nutritional deficiencies persist. However, the relationship between genetic enhancement and nutritional impact is mediated by factors beyond the plant itself—dietary practices, cooking losses, and cultural acceptance all influence final outcomes.

There are also emerging applications that extend beyond conventional yield-focused paradigms. Modifications aimed at reducing anti-nutritional compounds, improving storage resilience, or tailoring crops for specific industrial uses suggest a broadening of objectives in crop improvement. Whether these applications will integrate into mainstream agricultural systems or remain specialized niches is not yet entirely clear.

At times, while considering these applications, the question shifts—almost quietly—from what can be achieved to what should be prioritized. Is success defined by yield alone, or by stability across seasons, nutritional relevance, and compatibility with existing systems? CRISPR, by enabling targeted intervention, allows individual components of this equation to be addressed with precision. But

the equation itself... remains multi-dimensional, and perhaps deliberately resistant to simplification.

Challenges

Several challenges continue to complicate the deployment of CRISPR technology in agricultural systems, and while many of these are being actively addressed, they remain only partially resolved.

At the technical level, issues such as off-target effects, variable editing efficiency, and incomplete genome modification—particularly in polyploid crops such as wheat (*Triticum aestivum* L.)—introduce layers of uncertainty. While advances in guide RNA design and improved Cas variants have reduced off-target occurrences, they have not eliminated them entirely. In polyploid genomes, the requirement to simultaneously edit multiple homoeologous gene copies often results in heterogeneous outcomes, where phenotypic expression may not align consistently with the intended modification. This variability, though sometimes subtle, can influence trait stability across generations.

Regulatory frameworks, particularly in the Indian context, remain in transition. Distinctions between



transgenic crops and gene-edited crops—especially those lacking foreign DNA integration—are still being refined. This evolving regulatory landscape introduces a degree of ambiguity for researchers, developers, and policymakers alike. Delays in regulatory clarity can extend the timeline between laboratory validation and field deployment, often creating a disconnect between technological readiness and practical application.

Institutional constraints further shape the trajectory of CRISPR-based innovations. The pathway from laboratory research to varietal release is neither immediate nor linear. It involves sequential stages—confined field trials, multi-location testing across agro-climatic zones, seed multiplication, and eventual dissemination—each governed by its own protocols and timelines. These stages are essential for ensuring stability and safety, yet they also introduce temporal delays that may not align with the rapid pace of technological development. In some cases, the technology advances faster than the systems designed to evaluate it.

There is also a statistical dimension to this challenge. Traits that demonstrate significance under controlled experimental conditions ($p < 0.05$) do not always retain the same level of confidence when evaluated across

heterogeneous field environments. Genotype \times environment (G \times E) interactions, often substantial in magnitude, can dilute or obscure the expression of edited traits, particularly for complex characteristics such as yield or stress tolerance.

Finally, the question of farmer adoption remains central. Agricultural technologies, regardless of their scientific sophistication, are ultimately evaluated within the framework of farmer experience. Adoption is influenced not only by potential yield gains but by consistency across seasons, input requirements, risk perception, and compatibility with existing practices. A variety that performs exceptionally under research-managed conditions but inconsistently under farmer-managed environments may encounter resistance, irrespective of its genetic precision.

Trust, in this context, becomes a variable that is rarely quantified but deeply influential. It is built gradually—through repeated, reliable performance—and can be disrupted quickly when expectations are not met. In that sense, the challenge is not solely technological or regulatory, but relational—situated at the interface

between innovation and its acceptance.

Taken together, these challenges do not diminish the potential of CRISPR technology. Rather, they underscore the conditions under which that potential must be realized—through careful validation, contextual adaptation, and an awareness that precision at the molecular level does not automatically translate into predictability at the field level.

Conclusion

CRISPR technology represents, without much dispute, a significant advancement in the evolving toolkit of crop improvement. Its precision, flexibility, and relative speed have not only expanded the range of genetic interventions possible, but have also altered expectations regarding the pace at which such interventions can be realized. In molecular terms, the progress is both measurable and compelling.

And yet, its success—if one may use that word with a degree of caution—appears to depend less on what can be achieved under controlled laboratory conditions and more on how consistently those outcomes are expressed across the variability inherent to agricultural systems. Evidence

suggests that while certain traits retain statistical significance under defined environments ($p < 0.05$), their expression often becomes less predictable when subjected to the interacting influences of soil heterogeneity, climatic variability, and management diversity. The role of genotype \times environment (G \times E) interaction, therefore, remains central—perhaps more so than initially anticipated.

It becomes increasingly difficult, then, to view CRISPR as a replacement for conventional approaches. Rather, it appears more appropriately situated as an addition—one that complements existing breeding strategies while also demanding a more integrated understanding of plant physiology, environmental interaction, and agronomic practice. The strength of the technology lies in its precision; its limitation, if one may call it that, lies in the complexity of the systems into which that precision is introduced.

There is, perhaps, a quiet shift in perspective that accompanies this realization. The question moves away from what can be edited toward what can be sustained—across seasons, across locations, and across the varied conditions under which crops are actually

cultivated. Precision at the genomic level does not, in itself, guarantee predictability at the field level. The translation between the two remains conditional, mediated by factors that extend beyond the locus of modification.

It may be, then, that the future of CRISPR in agriculture will depend less on further increases in editing efficiency and more on the ability to align edited traits with ecological stability, statistical robustness, and practical usability. Multi-location validation, long-term performance assessment, and integration with farmer-managed systems will likely determine whether current advances translate into enduring agricultural outcomes.

The genome can now be accessed—and altered—with remarkable accuracy. The field, however, continues to respond on its own terms... and perhaps, it always will.

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