

# MERIDIAN: Cognitive-First Adaptive Learning Through Spaced Repetition, Calibrated Confidence, and Behavioral Economics

*Working Paper*

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## **Abstract**

This paper presents the cognitive science and behavioral economics framework underlying Meridian, a domain-agnostic adaptive learning platform. Meridian integrates five empirically validated learning mechanisms—spaced repetition scheduling, calibrated confidence assessment, generation-before-recognition sequencing, interleaved cross-domain practice, and self-explanation prompting—with a behavioral economics layer designed to sustain learner engagement over the weeks and months required for durable knowledge formation. The system is designed to produce knowledge that survives the transition from study session to real-world application, targeting use cases where retention failures have operational, clinical, or professional consequences. This paper details each mechanism, its empirical foundation, and its specific implementation within the Meridian architecture.

**Keywords:** adaptive learning, spaced repetition, FSRS, desirable difficulty, interleaved practice, metacognition, confidence calibration, behavioral economics, loss aversion, cognitive science, educational technology

## **1. The Problem: Content Delivery Is Not Learning**

The dominant paradigm in educational technology treats learning as content delivery. Material is presented, recall is tested, and the learner moves on. This model optimizes for throughput—courses completed, modules finished, certificates issued—while ignoring the central finding of over a century of memory research: without systematic reinforcement, most learned material is forgotten within days.

Hermann Ebbinghaus established in 1885 that memory decays exponentially in the absence of reinforcement, with the steepest losses occurring in the first hours after encoding. His forgetting curve demonstrated that learners retain roughly 58% of newly learned material after 20 minutes, 44% after one hour, and just 34% after 24 hours. Modern replications have confirmed these findings across diverse populations and content types.

The implication is stark: any learning system that delivers content without engineered reinforcement is building on sand. Completion rates and test scores measured at the point of delivery tell us almost nothing about what the learner will know in two weeks, two months, or two years. Yet these are precisely the metrics that most platforms optimize for, because they are easy to measure and satisfying to report.

Meridian takes a fundamentally different approach. Rather than optimizing for the moment of learning, it optimizes for the moment of application—which may be weeks or months after the study session. Every architectural decision in the system is informed by a specific, empirically validated principle from cognitive psychology or behavioral economics. The result is a platform where the learning science is not an afterthought bolted onto a content management system, but the structural foundation upon which everything else is built.

## **2. Cognitive Architecture**

Meridian’s cognitive layer comprises five mechanisms, each targeting a distinct failure mode of human memory and learning. No single mechanism is sufficient; their power lies in their interaction.

### **2.1 Spaced Repetition Scheduling (FSRS)**

Spaced repetition exploits the spacing effect—the finding that distributed practice produces superior long-term retention compared to massed practice. The effect has been documented continuously since Ebbinghaus and is among the most robust findings in all of experimental psychology.

Meridian implements the Free Spaced Repetition Scheduler (FSRS), a modern algorithm that maintains a per-item model of memory stability and retrievability. Unlike fixed-interval systems, FSRS computes each item’s review interval from its individual forgetting curve—a function of the item’s review history, the learner’s performance, and a target retention probability (default: 0.90). Items the learner struggles with are scheduled for earlier review; mastered material recedes to longer intervals. The algorithm continuously recalibrates as performance data accumulates, converging on the minimum number of reviews needed to maintain target retention.

The practical consequence is significant: the system eliminates both under-review (which leads to forgetting) and over-review (which wastes time without producing learning gains). Each review session is composed of precisely the items that need reinforcement at that moment.

### **2.2 Calibrated Confidence Assessment**

After each answer is revealed, the learner rates their own confidence on a three-level scale: didn’t know, partially knew, and knew it. This self-assessment—not the objective correctness of the answer—drives the FSRS scheduling algorithm. The separation is intentional and grounded in metacognitive monitoring theory.

The confidence  $\times$  correctness matrix produces four pedagogically distinct states, each requiring a different intervention strategy. A learner who selects the correct answer but reports low confidence has exposed a fragile memory trace requiring reinforcement. A learner who selects incorrectly but reports high confidence has exposed a dangerous illusion of knowing—a phenomenon extensively documented by Dunning and Kruger and by calibration researchers—that needs active disruption. In both cases, the raw correctness score alone would produce the wrong scheduling decision.

By placing the learner’s self-assessment at the center of the scheduling loop, Meridian creates a continuous feedback signal on metacognitive accuracy. Over time, learners develop better-calibrated awareness of what they actually know versus what they think they know—a skill with direct real-world consequences in any domain where confidence-weighted judgment matters.

### **2.3 Desirable Difficulty and the Generation Effect**

Bjork and Bjork’s desirable difficulties framework demonstrates that conditions which slow initial learning often accelerate long-term retention. Meridian operationalizes this through the generation effect—the finding that self-generated answers are retained significantly better than passively received ones, even when the generated answer is incorrect.

In practice, questions are presented without answer choices visible. The learner first encounters a free-response field and is prompted to generate their own answer from memory before optionally expanding the multiple-choice options. This forced retrieval attempt—even when it fails—activates the memory trace more deeply than passive recognition. The multiple-choice options remain available as scaffolding for learners who need them, but they are secondary to the generation attempt.

The deliberate sequencing of generation before recognition is a direct application of Bjork and Bjork’s principle that “conditions of learning that make performance improve rapidly often fail to support long-term retention and transfer, whereas conditions that create challenges and slow the rate of apparent learning often optimize long-term retention and transfer.” The learner may experience the generation phase as more difficult and less efficient than immediately viewing the answer choices—and they would be correct about the difficulty, but wrong about the efficiency.

### **2.4 Interleaved Practice**

Study sessions in Meridian draw from multiple knowledge domains rather than blocking by topic. A session might interleave questions spanning several subject areas, forcing the learner to identify which knowledge schema applies to each question. This design is grounded in research by Rohrer and Taylor and Kornell and Bjork, who demonstrated that interleaved practice produces superior transfer and discrimination ability compared to blocked study.

The mechanism operates through discrimination training. When practice is blocked by topic, the learner knows the strategy before reading the problem—the context supplies the answer category for free. Interleaving removes this cue, requiring the learner to first identify what kind of problem they are facing before selecting a solution strategy. This discrimination skill is precisely what blocked practice never develops and what real-world application always requires.

A particularly relevant finding from this literature is the metacognitive illusion it exposes: learners consistently perceive blocked practice as more effective than interleaving, even when their own test performance demonstrates the opposite. Meridian incorporates interleaving by design rather than leaving it to learner preference, because learner preference in this case is demonstrably miscalibrated.

## **2.5 Self-Explanation Prompting**

After the correct answer is revealed, Meridian prompts the learner to explain in their own words why it is correct. This self-explanation protocol is grounded in the work of Chi and colleagues, who established that self-explanation produces learning gains that persist across transfer tasks.

The mechanism operates through gap-filling. When learners attempt to construct a causal explanation, they encounter gaps in their own understanding that passive reading of an expert explanation would never expose. The act of generating an explanation activates relational reasoning and connects new information to existing knowledge structures—an elaborative encoding process that strengthens both storage and retrieval pathways.

Chi et al. (1989) found that “good” learners—those who achieved higher transfer performance—generated significantly more self-explanations while studying worked examples than “poor” learners. The follow-up study demonstrated that prompting self-explanations improved understanding even among learners who would not self-explain spontaneously, confirming that the effect is trainable rather than trait-dependent.

## **3. Behavioral Economics Layer**

The cognitive architecture described above is necessary but insufficient. Spaced repetition, desirable difficulty, and interleaving produce durable learning only if the learner returns to the system repeatedly over weeks and months. Retention is, at its core, a behavior problem—and behavior is the domain of behavioral economics.

Meridian incorporates four behavioral mechanisms, each grounded in established findings from decision science and behavioral psychology.

### **3.1 Streak Mechanics and Loss Aversion**

A visible day-streak counter leverages loss aversion—the empirically demonstrated tendency for losses to be experienced roughly twice as intensely as equivalent gains. Kahneman and Tversky’s prospect theory established that the value function for losses is steeper than for gains, meaning that the psychological cost of breaking an established streak exceeds the effort cost of a brief study session required to maintain it.

The streak is the first signal the learner sees upon return. Once established above a threshold (typically 3–5 days), the streak becomes self-reinforcing: each additional day raises the perceived loss associated with breaking it. This mechanism does not require gamification, points, badges, or extrinsic rewards—it operates on the learner’s inherent sensitivity to loss, which Kahneman and Tversky showed to be a universal feature of human decision-making under risk.

### **3.2 Mastery Percentage and the Goal-Gradient Effect**

A mastery percentage—computed from FSRS stability thresholds across the full question bank—provides a long-arc progress signal that complements the streak’s daily signal. The theoretical basis is the goal-gradient effect, first described by Hull and later validated in consumer behavior by Kivetz, Urminsky, and Zheng, which demonstrates that effort increases as perceived progress toward a goal increases.

The mastery metric is deliberately conservative. A question is counted as “mastered” only when its FSRS stability parameter exceeds 21 days—meaning the algorithm predicts the learner will retain the answer for at least three weeks without review. This ensures the number is always honest, always has room to grow, and always reflects genuine retention rather than recency of exposure. The slow-building nature of the metric prevents the ceiling effect that undermines engagement when progress indicators saturate too quickly.

### **3.3 Session Sizing and Commitment Reduction**

Meridian offers multiple study modes with different session lengths—ranging from 10 to 40 questions. The existence of a short-session option reduces the perceived commitment barrier that prevents learners from starting. This design draws on the foot-in-the-door effect and Cialdini’s commitment and consistency principle: once a person has committed to a small action, they are significantly more likely to continue with a larger one.

Behavioral research consistently demonstrates that starting is the hardest part of any effortful activity. The “quick” session (10 questions, approximately 5 minutes) exists not because 10 questions produce optimal learning, but because the option to study for only 5 minutes removes the activation energy barrier that would otherwise prevent the session from occurring at all. Most learners who begin a quick session continue beyond the minimum.

### **3.4 Open-Loop Tension and the Zeigarnik Effect**

At the end of each session, the summary screen displays outstanding review counts and upcoming due cards—creating an open cognitive loop. This leverages the Zeigarnik effect, the finding that incomplete tasks create a mild cognitive tension that biases toward task resumption, a phenomenon further developed by Ovsiankina.

The information is factual, not manufactured. There is no artificial urgency, no countdown timer, no “you’re falling behind” messaging. The open loop works precisely because it reflects genuine task incompleteness: there are cards due for review, and the learner’s own memory traces will decay without reinforcement. The Zeigarnik effect operates on authentic incompleteness, not on fabricated scarcity.

## **4. Mechanism Interaction**

The mechanisms described above are not independent interventions applied in parallel. They form an integrated system in which each mechanism amplifies the others. Soderstrom and Bjork’s review of the learning-versus-performance distinction provides the theoretical frame: conditions that maximize current performance (ease, fluency, blocked practice) are systematically different from conditions that maximize long-term learning (difficulty, effort, interleaving). Meridian is designed to operate on the learning side of this distinction.

The interaction effects are specific and identifiable. Spaced repetition determines *when* a learner encounters each item; the generation effect determines *how* they engage with it; interleaving determines *what* surrounds it; confidence calibration determines how the encounter is *classified*; and self-explanation determines how deeply the answer is *processed* after reveal. The behavioral economics layer then ensures the learner returns tomorrow to repeat the cycle.

No single mechanism in this system is novel. Spaced repetition has been studied since 1885. The generation effect was formalized in 1978. Interleaving benefits were demonstrated in 2007. Loss aversion was established in 1979. What Meridian contributes is not the discovery of these effects but their integration into a unified system architecture that is both domain-agnostic and immediately deployable.

## 5. Domain Agnosticism

Every mechanism in Meridian's architecture operates on the structure of human memory and motivation, not on the content of any particular domain. The cognitive principles apply identically whether the learner is encoding acupuncture point locations, aviation emergency procedures, regulatory compliance requirements, or foreign-language vocabulary.

The system requires only a structured question bank—questions, choices, correct answers, explanations, and metadata (domain, difficulty, exam type)—to be adapted to a new field. The scheduling algorithm, confidence calibration loop, generation sequencing, interleaving logic, and behavioral engagement mechanisms all operate without modification.

This domain agnosticism is not a marketing claim. It is a direct consequence of building on cognitive science that describes how human memory works, rather than on pedagogical traditions specific to any particular discipline. The same Ebbinghaus forgetting curve operates on medical board content and on cybersecurity certification content; the same Dunning-Kruger miscalibration affects aspiring acupuncturists and aspiring financial analysts; the same Zeigarnik tension drives return behavior whether the unfinished task is a set of pharmacology flashcards or a set of export-control regulation questions.

## 6. Conclusion

Meridian is not an educational technology in the conventional sense. It is a behavioral system that happens to operate on educational content. Its architecture is designed around two questions that most learning platforms never ask: Will the learner remember this information when they need it? And will the learner come back tomorrow to continue the process that makes long-term retention possible?

The answers to both questions depend on well-understood mechanisms from cognitive psychology and behavioral economics—mechanisms that have been individually validated across thousands of studies spanning more than a century. Meridian's contribution is the integration of these mechanisms into a coherent, deployable system that can be configured for any knowledge domain where retention matters more than completion.

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