

# Emotional Neural Networks: Teaching AI Not Just to Think, but to *Feel* Its Decisions

A new class of hybrid architecture combining learnable reasoning with learnable internal feelings to create more robust and introspective AI.



# Today's AI: Brilliant Calculators Without Introspection

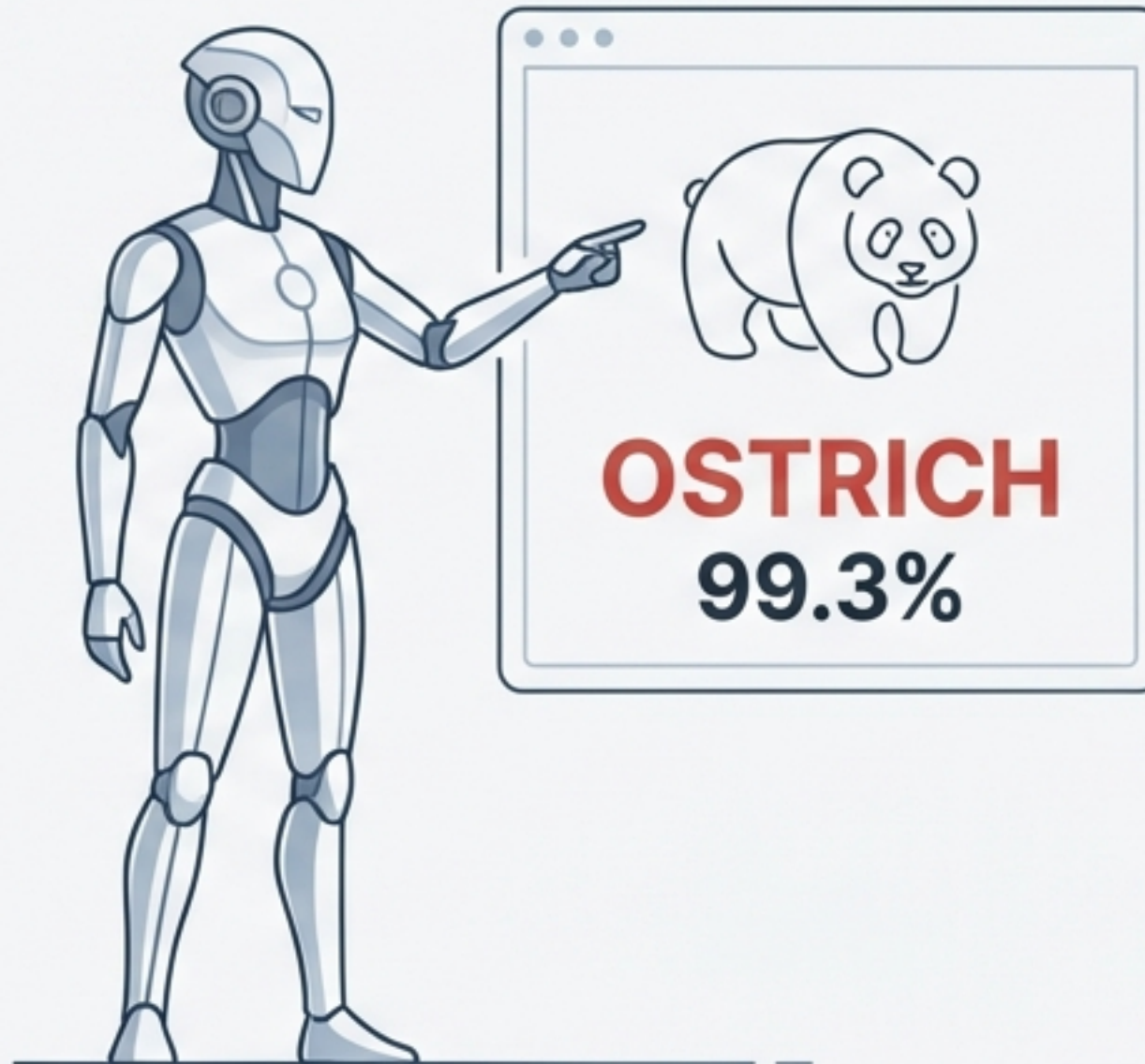
Standard Neural Networks excel at learning statistical regularities but lack internal awareness. They cannot distinguish between a decision that is *merely likely* and one that also *feels right or wrong*.

## Confidently Wrong

Can produce high-confidence predictions that are factually incorrect, with no internal mechanism to flag the error.

## Vulnerable to Adversarial Input

Minor, imperceptible perturbations can cause catastrophic failures in classification.



## Brittle Under Uncertainty

Performance degrades unpredictably when faced with novelty, ambiguity, or out-of-distribution data.

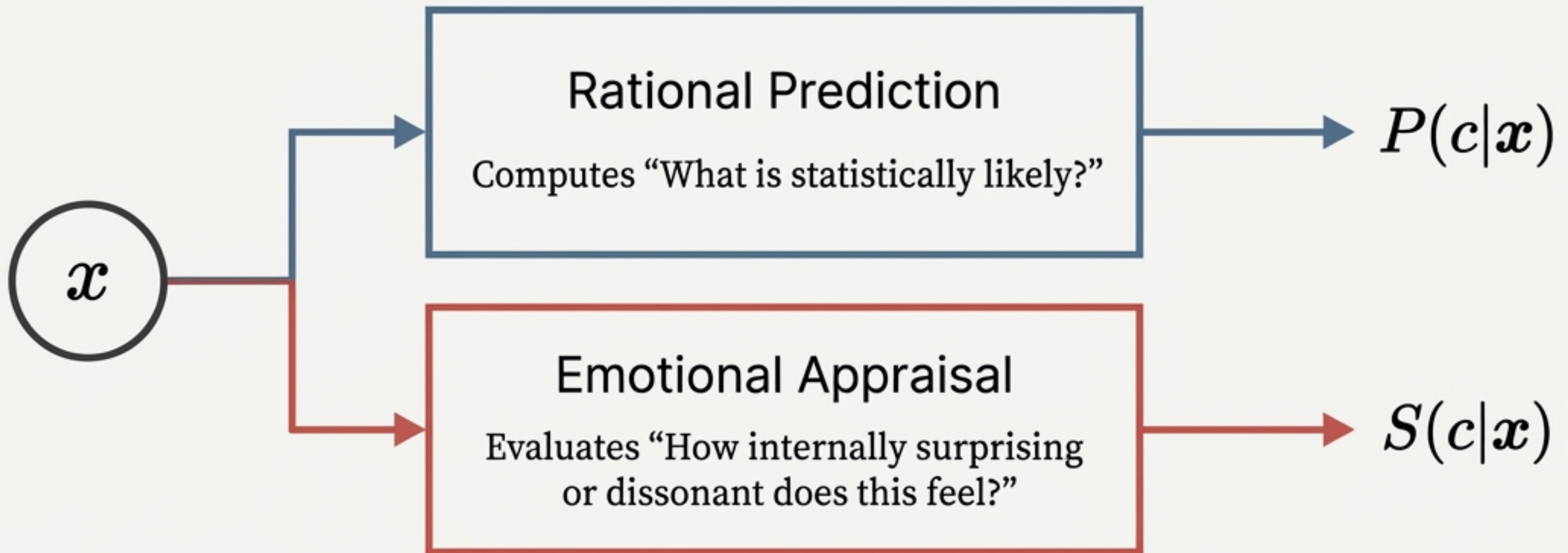
## Lacks Self-Reflection

A standard NN computes probability but has no capacity to reflect on the internal consistency of its own decision-making process.



# The Solution: A Hybrid of Reason and Feeling

Introducing Emotional Neural Networks (ENNs), an architecture inspired by dual-process theories of the mind. ENNs operate with two parallel, independently trained streams:

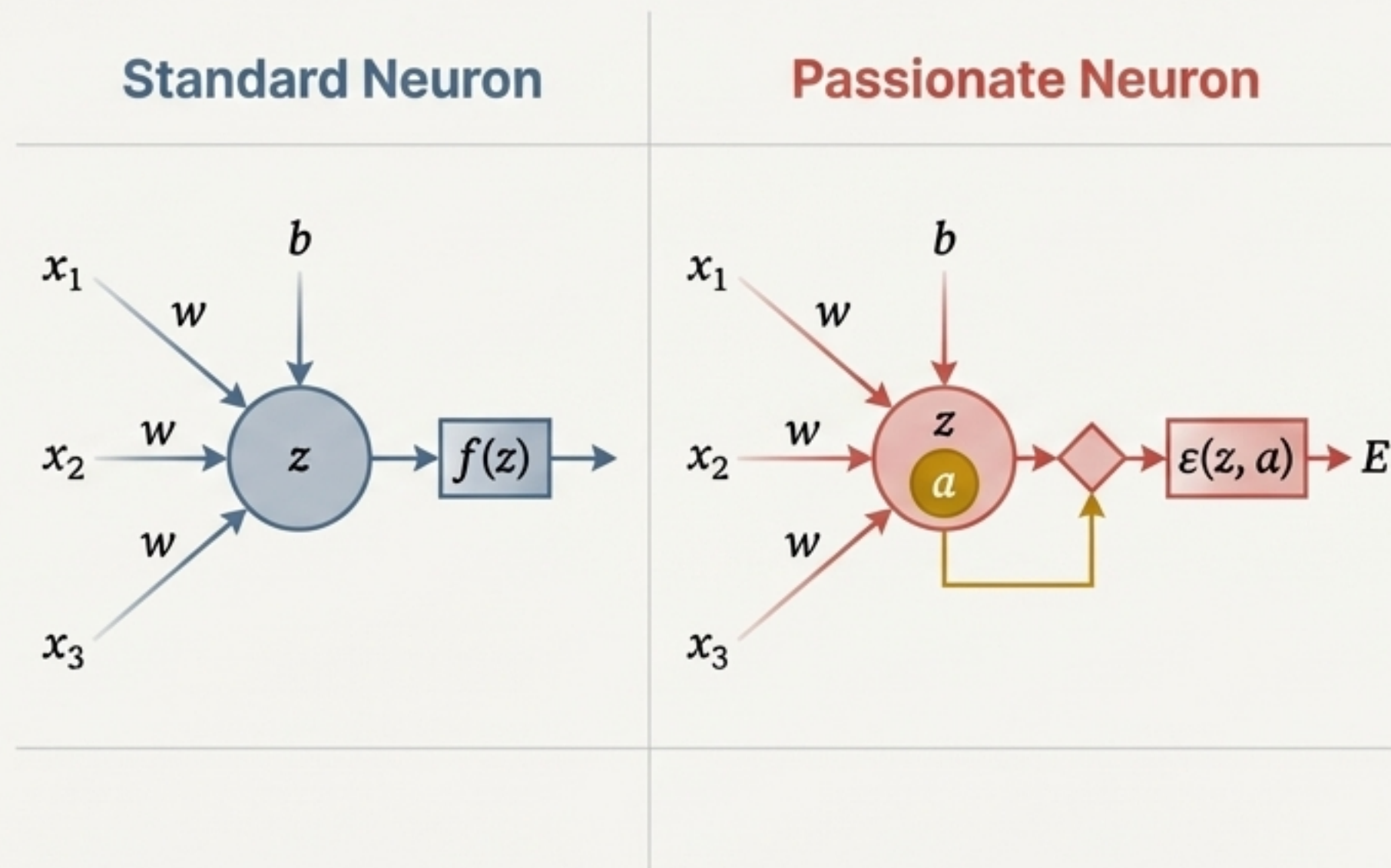




# The Heart of the ENN: The Passionate Neuron.

The Passionate Neuron is a computational unit that doesn't just process input passively but reflects on its own internal state.

- **Belief (' $z$ ')**: The neuron's pre-activation value ( $z = \mathbf{w} \cdot \mathbf{x} + \mathbf{b}$ ), representing its initial appraisal of the input.
- **Expectation (' $a$ ')**: A trainable parameter representing the neuron's learned emotional baseline or norm for its belief.
- **Emotion (' $E$ ')**: The final activation output, quantifying the reaction to the mismatch between belief and expectation.





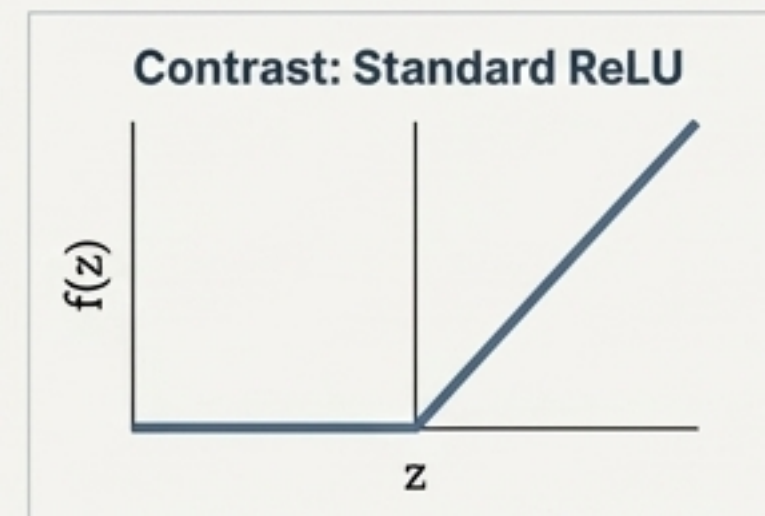
# The 'Anxiety' Activation: A Simple Measure of Surprise

$$E = \varepsilon(z, a) = |z - a|$$

This quantifies the emotional reaction to the mismatch between what the neuron believes ( $z$ ) and what it expected ( $a$ ).

A larger gap triggers a stronger 'emotional' signal.

It directly measures the 'distance' between belief and expectation.



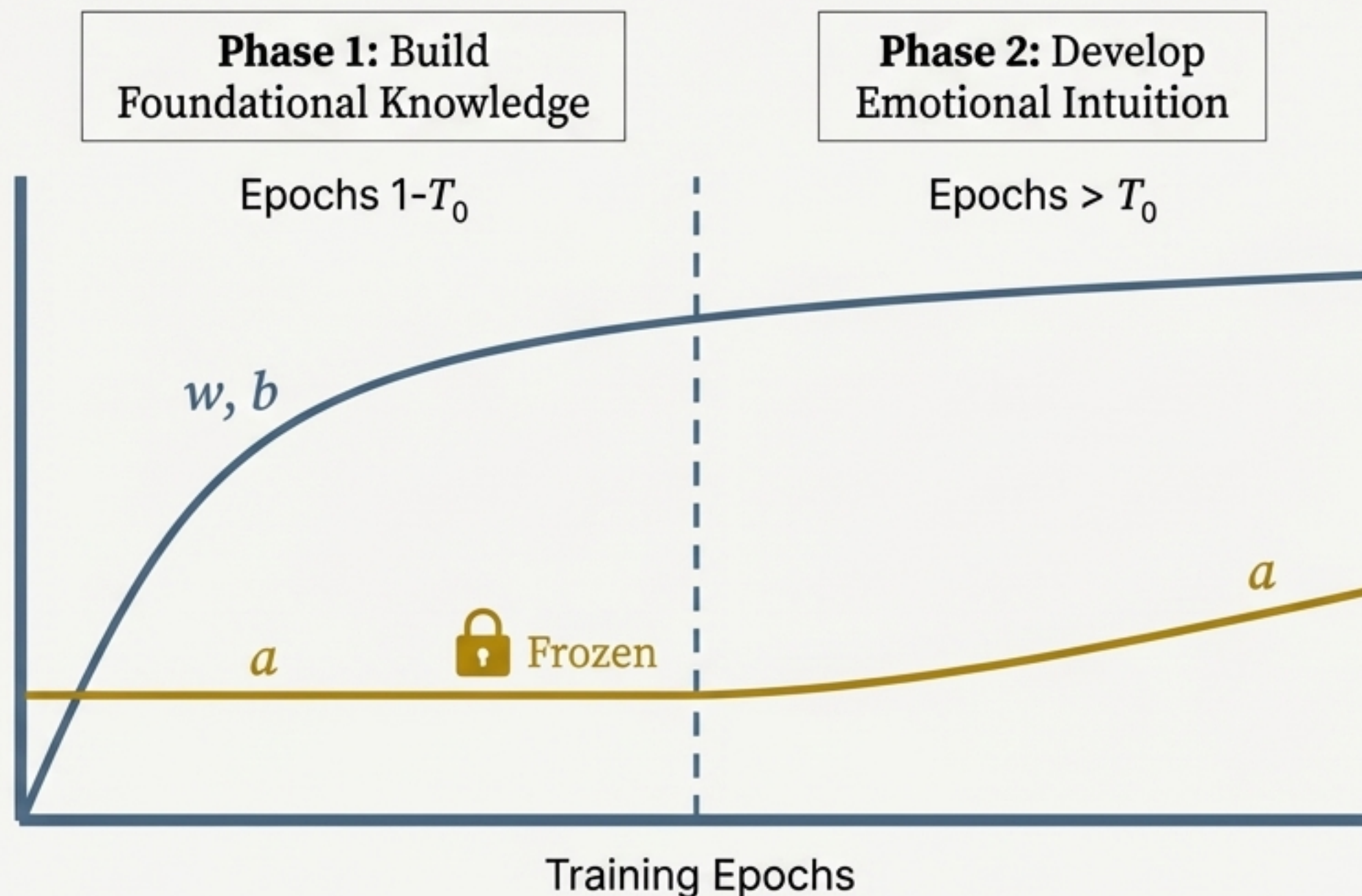
- Piecewise linear: Ensures stable and predictable gradient behavior.
- Symmetric around a learnable anchor  $a$ : Unlike ReLU, which is fixed at 0, each neuron learns its own 'center of importance'.
- Computationally lightweight: Efficient for large-scale applications.



# Learning to Feel: Training the Emotional Anchor.

The expectation  $a$  is a trainable parameter, but it evolves differently from the weights  $w$  and biases  $b$ .

- **Weights & Biases ( $w, b$ ):** Learn quickly, adapting to specific data patterns. This is analogous to acquiring skills or context-dependent knowledge.
- **Expectation ( $a$ ):** Learns slowly, with a lower learning rate and less frequent updates. It represents a cumulative, stable model of what the neuron considers familiar. This is analogous to forming core values or stable priors.



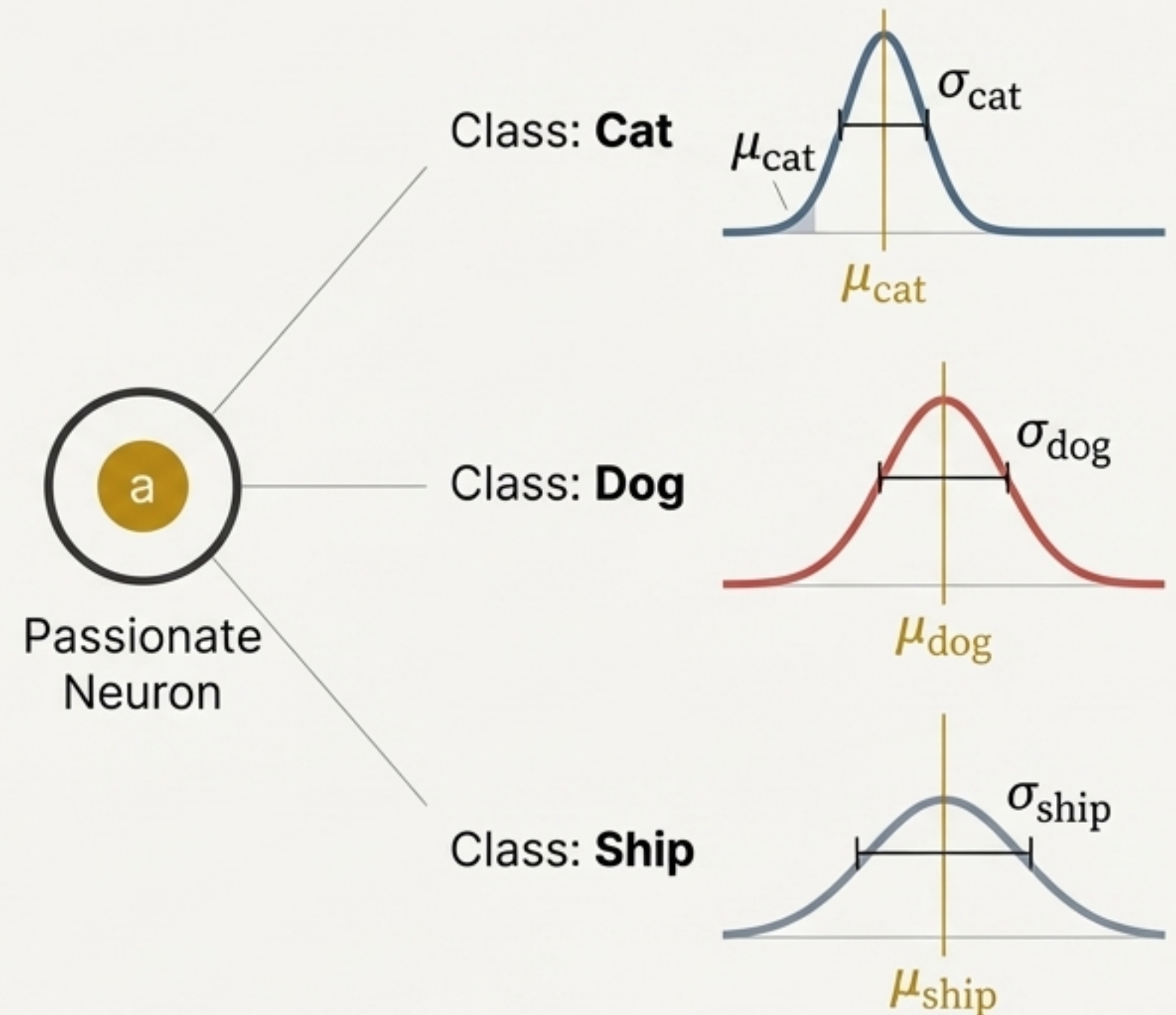


# The Emotional Portfolio: A Neuron's Memory of Its Feelings.

After training, the ENN builds a statistical memory of its emotional reactions for each neuron on a per-class basis. These statistics are collected by running a dedicated “portfolio subset” of the data ( $D_{\text{portfolio}}$ ) through the trained network.

**Mean Emotional Intensity ( $\mu_i^c$ ):** “How strongly do I usually feel about class `c`?”

**Emotional Volatility ( $\sigma_i^c$ ):** “How stable are my feelings for class `c`?”



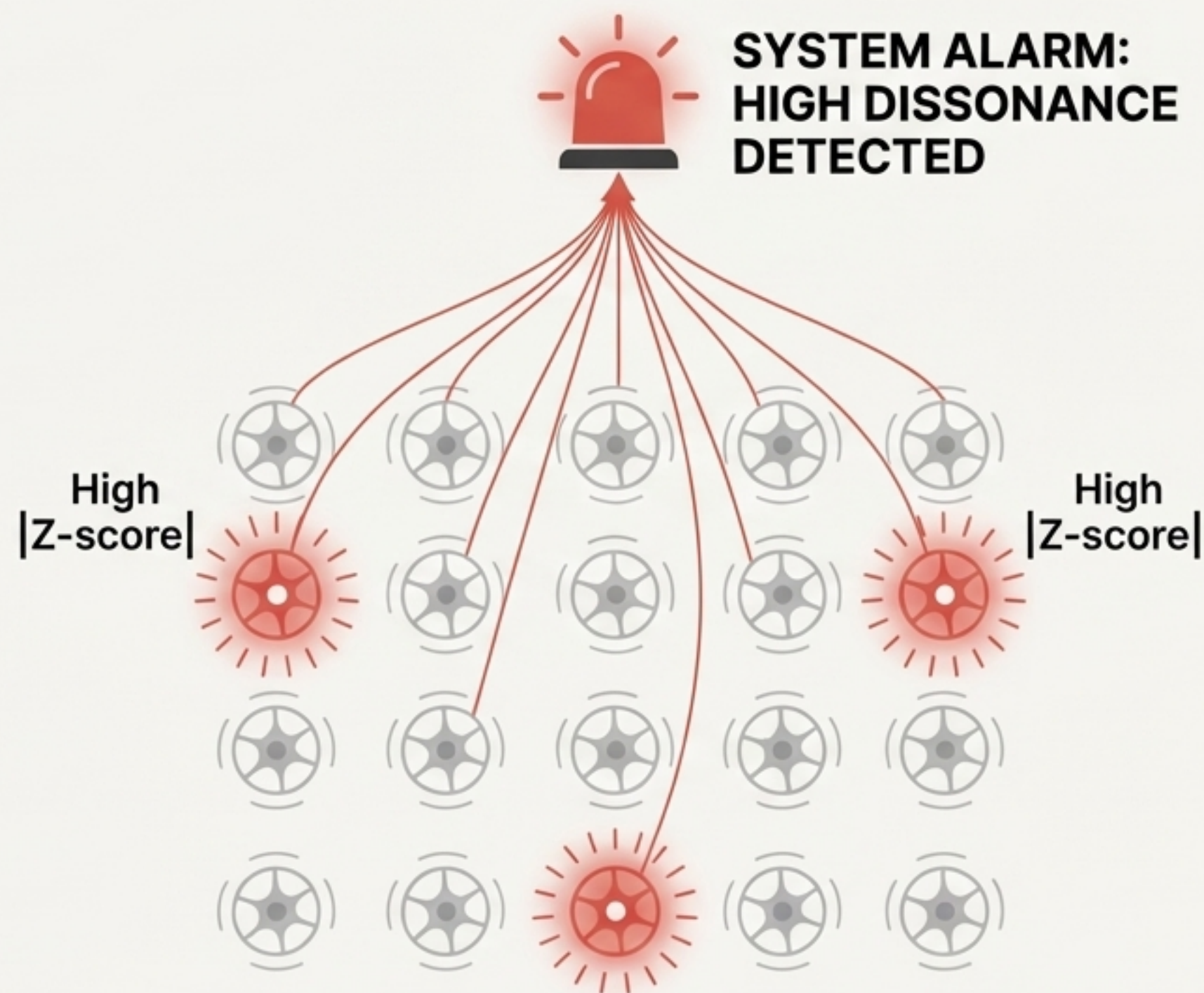


# From Local Surprise to Global Alarm

During inference, each neuron uses its emotional portfolio to calculate a Z-score for its current emotion relative to a class.

$$Z_i^c(x) = \frac{E_i(x) - \mu_i^c}{\sigma_i^c + \epsilon}$$

A high absolute Z-score is a **local alarm**, indicating the neuron's current reaction is atypical. Individual Z-scores are aggregated into a surprise vector, which is compared to the class's "emotional signature" to compute a final alarm score.


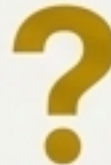
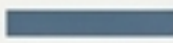





# Two Outputs Are Better Than One: Probability and Surprise.

An ENN produces two distinct, parallel outputs for any given input  $x$ :

- **Probability Vector**  $P(c|x)$ : The standard rational output. “This is 95% likely a cat.”
- **Alarm Vector**  $S(c|x)$ : The introspective emotional output. “But my emotional alarm for ‘cat’ is at 82%.”

	Low Alarm (Emotionally Consistent)	High Alarm (Emotionally Dissonant)
High Probability	<b>Confident &amp; Aligned</b> The ideal case. Prediction is statistically strong and feels right. 	<b>Confident but Suspicious</b> A red flag. “Likely, but something feels wrong.” Potential adversarial attack or anomaly. 
Low Probability	<b>Familiar Negative</b> A clear rejection that aligns with expectations. 	<b>Novelty/Anomaly</b> Strong rejection. The input is not recognized rationally or emotionally. 

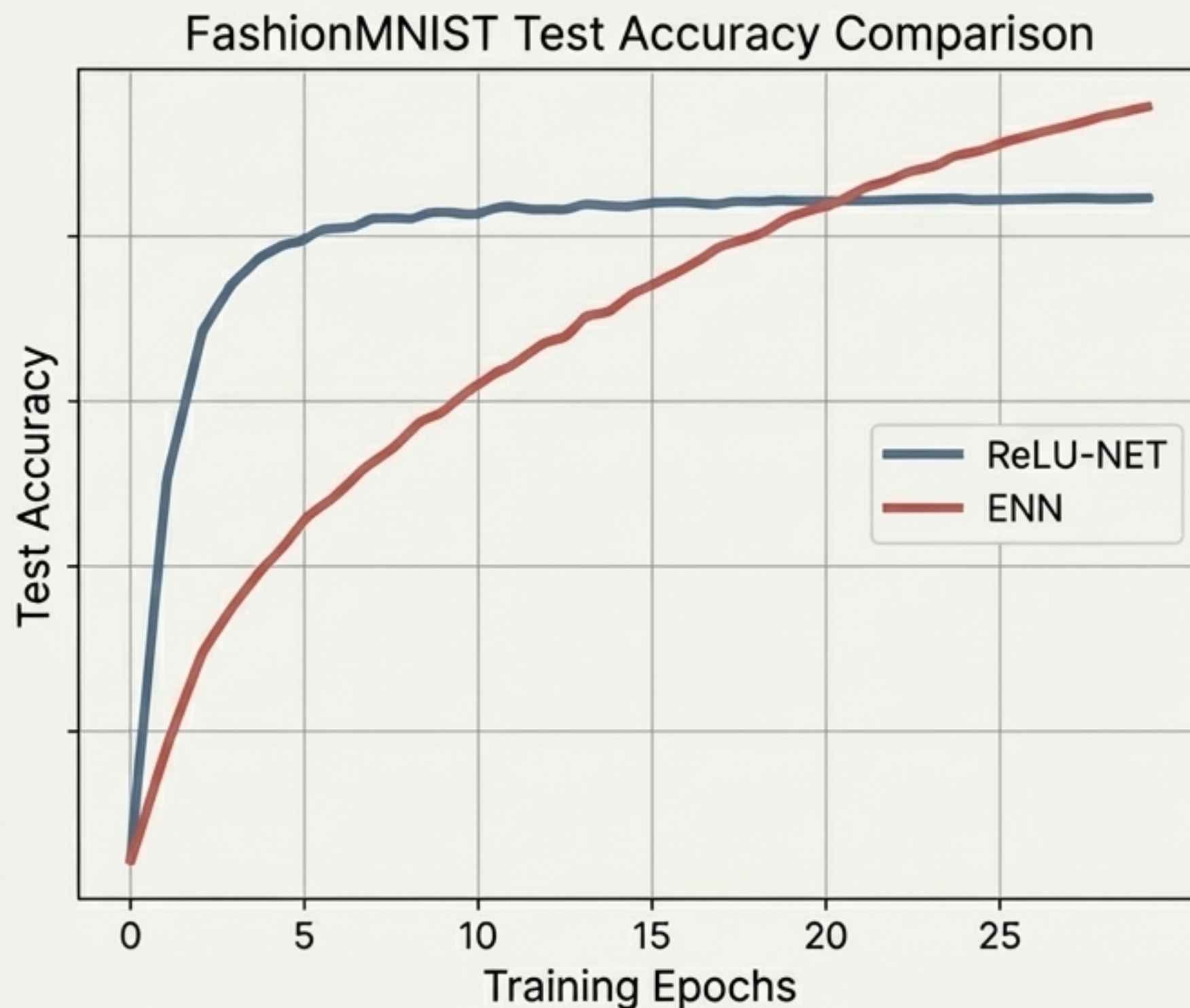


# Emotion Enhances Generalization.

**Experiment:** A minimal ENN (using only the anxiety activation) was compared against an identical ReLU-based network on the FashionMNIST dataset.

**Finding:** The ENN's learning was slower but more stable and consistent, eventually surpassing the baseline ReLU-NET in test accuracy.

**Observation:** “ReLU-NET typically exhibited faster initial progress... but tended to plateau early. In contrast, ENN... demonstrated slower but more consistent improvement over time.”





# Inspired by How Intelligent Beings Actually Work

ENNs are not an arbitrary design but a synthesis of insights from multiple disciplines.

Emotions are not noise; they are structured reactions to cognitive discrepancies.



## Philosophy

Emotions as structured “judgments” arising from the evaluation of events relative to personal concerns.



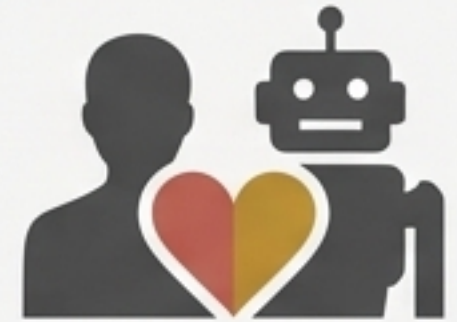
## Psychology

Appraisal Theory, where emotional intensity scales with the gap between reality and expectation.



## Neuroscience

The brain as a prediction engine. The amygdala’s role in responding to unexpected stimuli and prediction errors.



## Affective Computing

Emotion as integral to intelligence and rational decision-making, not a hindrance to it.



# A Step Towards Computational Proto-Consciousness.



## The Connection

The ability to model its own internal state, compare it to learned expectations, and react to discrepancies is a foundational step towards self-monitoring systems.

## Definition

ENNs implement a form of **computational proto-consciousness**: a pre-reflective capacity where the system is aware of mismatches between its internal model and the world. It is the “feeling of what happens” within the system.



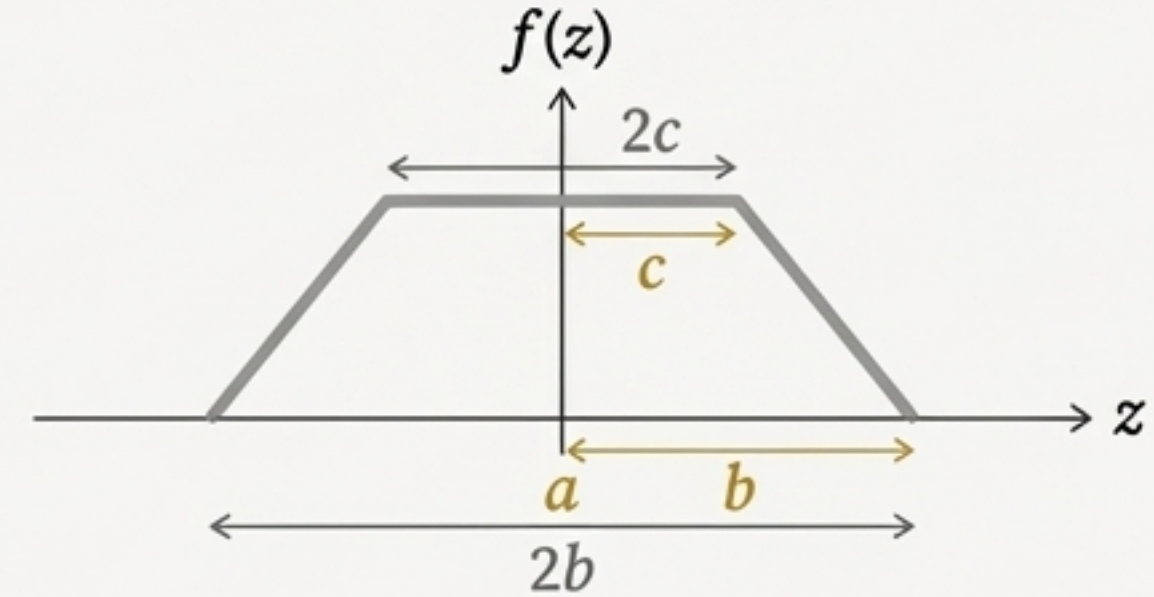
# Beyond Anxiety: A Family of Emotional Activations

## General Framework & Logics

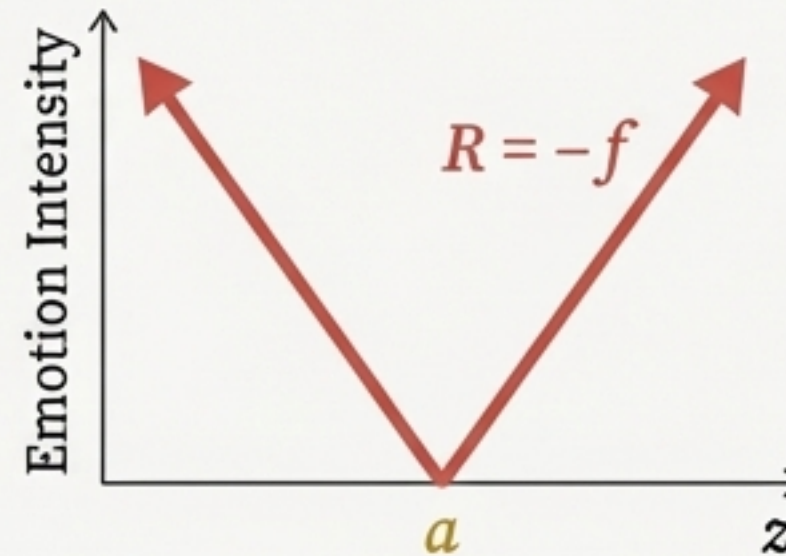
The anxiety function is a specific case of a generalized family of emotional activations based on a trapezoidal function  $f(z; a, b, c)$ .

- **Reflectors:** Amplify deviation from an anchor ( $R = -f$ ). They model surprise, tension, and alertness. The  $|z-a|$  anxiety function is an unbounded Linear Reflector.
- **Attractors:** Reward closeness to an anchor ( $A = f$ ). They model satisfaction, comfort, and resonance.

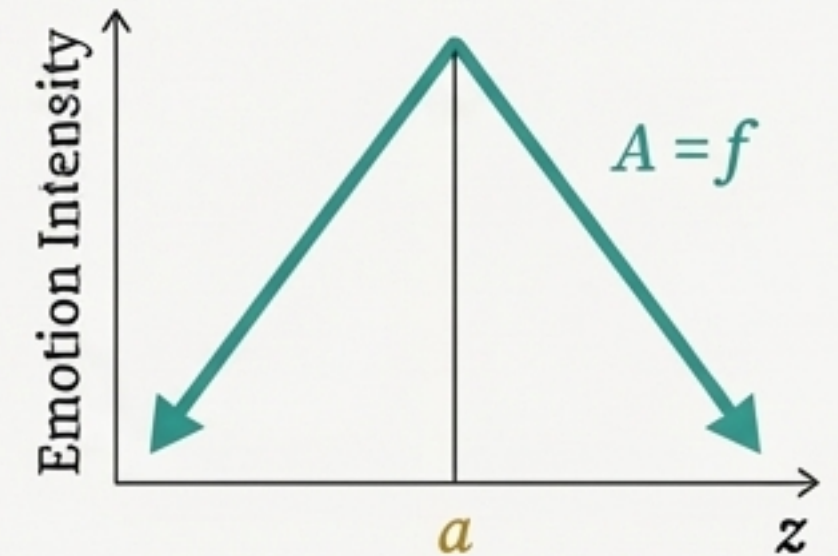
Generalized Trapezoidal Function  $f(z; a, b, c)$



**Reflector** (Surprise)



**Attractor** (Satisfaction)



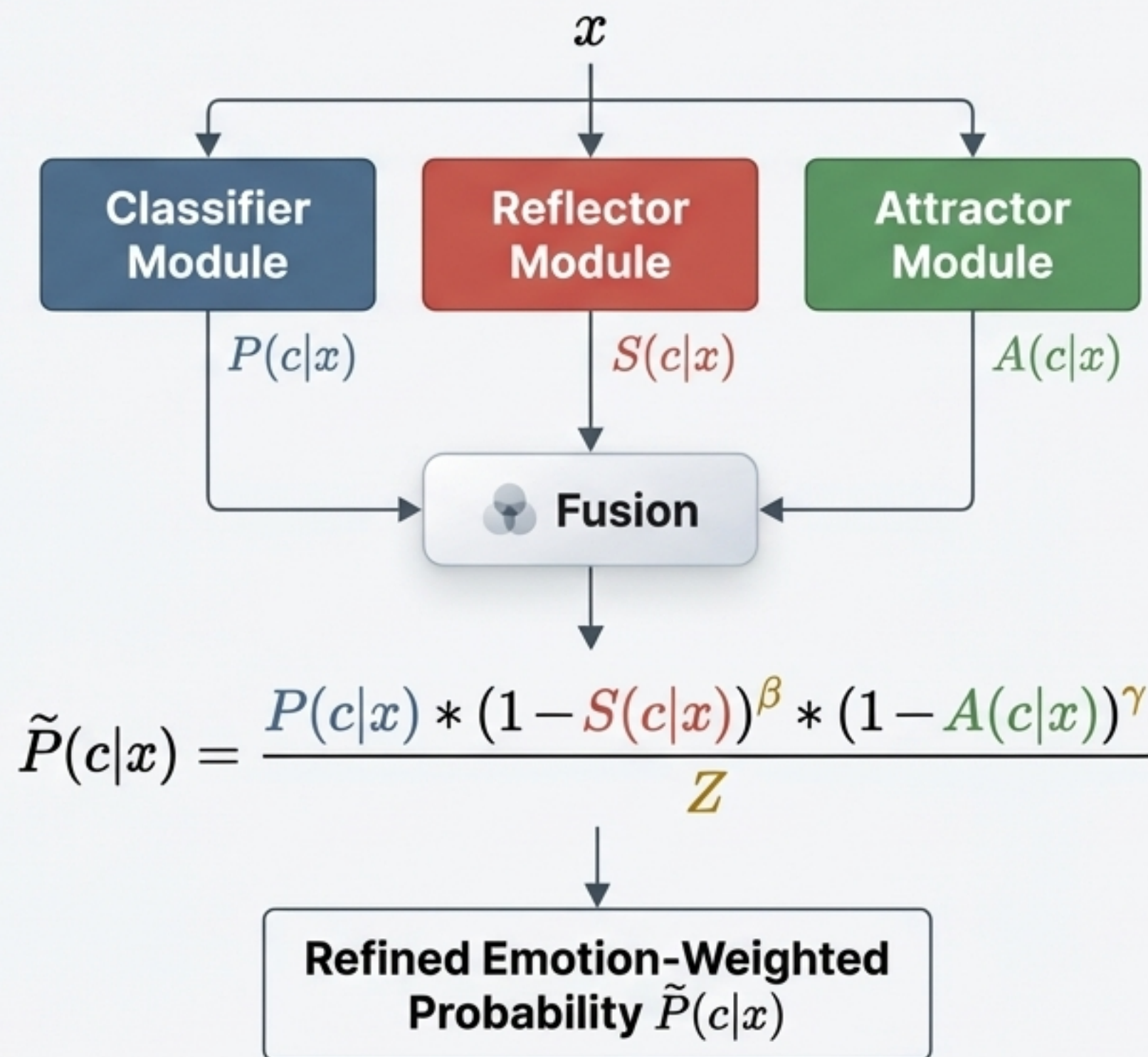


# A Richer Inference: Fusing Probability, Surprise, and Satisfaction

A fully-realized ENN can produce three parallel distributions:

1. **Probability**  $P(c|x)$ : From the rational stream.
2. **Surprise**  $S(c|x)$ : A novelty score from Reflector neurons.
3. **Satisfaction Mismatch**  $A(c|x)$ : A novelty score from Attractor neurons.

**Intuition:** A class is reinforced if it is probable AND emotionally typical (low surprise, low satisfaction mismatch). A class is penalized if it feels emotionally dissonant, even if it is statistically likely.





# The Future of AI is Thinking *and* Feeling.



**Hybrid Architecture:** ENNs integrate rational prediction with an internal, learnable 'emotional' appraisal stream.



**Introspective Capabilities:** Emotional portfolios and dual-channel outputs enable self-monitoring, anomaly detection, and richer, more nuanced judgments.



**Powerful Core:** The 'Passionate Neuron' with its simple  $|z-a|$  anxiety activation is computationally efficient and improves model generalization.



**A New Paradigm:** This framework opens the door to more robust, interpretable, and affect-aware AI systems that can develop a sense of what is not just likely, but appropriate.