

Veritas Consensus Protocol

Abstract

A consensus mechanism where voice equals risk. Participants stake capital to gain influence. Belief Decomposition aggregates opinions, Bayesian Truth Serum scores accuracy, and stake redistribution rewards truth-tellers.

1 Stake System

Influence in consensus is proportional to capital at stake.

1.1 Stake and Lock Definitions

Each participant maintains:

- **Global stake** S_i = total at-risk capital across all markets
- **Per-market lock** $L_{i,m}$ = locked capital in market m

Lock amount = 2% of last buy amount in that market.

1.2 Solvency Invariant

Total stake must cover all lock requirements:

$$S_i \geq \sum_{\text{all markets}} L_{i,m} \quad (1)$$

1.3 Automatic Collateralization

On trade execution:

$$\text{Skim} = \max \left(0, \sum_m L_{i,m}^{\text{new}} - S_i \right) \quad (2)$$

Skim is collected atomically with the trade only when total lock requirement exceeds current stake.

1.4 Epistemic Weight

Voting weight in each market equals locked capital:

$$w_i = L_{i,m} \quad (3)$$

2 Belief Decomposition

Extracts shared knowledge from diverse opinions.

2.1 Input Requirements

Each participant provides:

- Belief p_i (e.g., 0.7 for 70% confidence)
- Meta-prediction m_i (prediction of population average)

2.2 Local Expectations Matrix

Calculate belief-prediction correlations:

$$w_{11} = \text{weighted avg}(p_i \times m_i \mid p_i = 1) \quad (4)$$

$$w_{21} = \text{weighted avg}(p_i \times m_i \mid p_i = 0) \quad (5)$$

2.3 Common Prior

Extract shared baseline before evidence:

$$\pi = \frac{w_{21}}{w_{21} + (1 - w_{11})} \quad (6)$$

2.4 Aggregate Belief

Combine beliefs with prior correction:

$$A = \frac{\prod_i p_i^{\hat{w}_i} / \pi^{\sum \hat{w}_i}}{\prod_i p_i^{\hat{w}_i} / \pi^{\sum \hat{w}_i} + \prod_i (1 - p_i)^{\hat{w}_i} / (1 - \pi)^{\sum \hat{w}_i}} \quad (7)$$

2.5 Certainty Measure

Agreement level from Jensen-Shannon divergence:

$$c = 1 - [H(A) - \sum_i \hat{w}_i H(p_i)] \quad (8)$$

where $H(p) = -p \log_2 p - (1 - p) \log_2 (1 - p)$. High certainty amplifies redistribution.

3 Bayesian Truth Serum

Measures information contribution of each participant.

3.1 Score Formula

$$s_i = D_{\text{KL}}(p_i \parallel \bar{m}_{-i}) - D_{\text{KL}}(p_i \parallel \bar{p}_{-i}) - D_{\text{KL}}(\bar{p}_{-i} \parallel m_i) \quad (9)$$

where \bar{p}_{-i} and \bar{m}_{-i} are leave-one-out aggregates excluding agent i .

Positive scores indicate informative, accurate contributions. Negative scores indicate noise.

3.2 Zero-Sum Property

Weighted scores sum to zero:

$$\sum_i w_i \times s_i = 0 \quad (10)$$

Creates natural winner and loser sets for redistribution.

4 Stake Redistribution

4.1 P90 Normalization

Compute 90th percentile scale factor:

$$k = \max(\text{P90}(|s_1|, |s_2|, \dots, |s_n|), 0.1) \quad (11)$$

Normalize scores to $[-1, 1]$:

$$\hat{s}_i = \text{clamp}\left(\frac{s_i}{k}, -1, +1\right) \quad (12)$$

Prevents outliers from dominating redistribution.

4.2 Slashes and Rewards

For losers (negative scores):

$$\text{Slash}_i = c \times |\hat{s}_i| \times w_i \quad (13)$$

For winners (positive scores):

$$\text{Reward}_i = \frac{\text{your signal strength}}{\text{total signal strength}} \times \text{total slashed} \quad (14)$$

4.3 Bounded Risk

You can never lose more than your lock in any single market:

$$\text{Slash}_i \leq L_{i,m} \quad (15)$$

5 ICBS Prediction Markets

5.1 Market Structure

Each belief market has LONG and SHORT tokens. Traders buy LONG to express confidence, SHORT to express skepticism.

5.2 Inverse Coupling

When you buy LONG, SHORT gets cheaper.

When you buy SHORT, LONG gets cheaper.

This prevents pure hype spirals.

5.3 Market Prediction

The market's consensus forecast:

$$q = \frac{\text{LONG reserves}}{\text{LONG reserves} + \text{SHORT reserves}} \quad (16)$$

5.4 Settlement

At epoch end, compare market prediction q to BD score x :

If LONG was right ($x > q$):

- LONG reserves grow by factor x/q
- SHORT reserves shrink by factor $(1 - x)/(1 - q)$

If SHORT was right ($x < q$):

- SHORT reserves grow
- LONG reserves shrink

Total value locked stays constant. No tokens minted or burned.

6 Game Theory

6.1 Nash Equilibrium

Rational traders buy LONG when they believe:

$$\mathbb{E}[\text{BD score}] > q \quad (17)$$

At equilibrium, the market prediction equals expected truth:

$$q = \mathbb{E}[x] \quad (18)$$

6.2 Optimal Strategy

Rational participants trade based on expected BD score relative to market prediction:

$$\mathbb{E}[x] > q \implies \text{buy LONG} \quad (19)$$

$$\mathbb{E}[x] < q \implies \text{buy SHORT} \quad (20)$$

$$\mathbb{E}[x] = q \implies \text{hold} \quad (21)$$

6.3 Sybil Resistance

Identity fragmentation provides no advantage. Total epistemic weight and expected rewards remain invariant under capital splitting:

$$\sum_{j \in \text{sybils}} w_j = w_{\text{original}}, \quad \sum_{j \in \text{sybils}} \mathbb{E}[\Delta S_j] = \mathbb{E}[\Delta S_{\text{original}}] \quad (22)$$

7 Protocol Lifecycle

7.1 Trade Execution

Agent buys tokens, triggering lock calculation ($L = 0.02 \times \text{amount}$). If total lock requirement exceeds current stake, skim is collected atomically. Lock replaces previous lock for that position.

7.2 Epoch Processing

Participants submit beliefs p_i and meta-predictions m_i . BD computes aggregate A and certainty c . BTS scores s_i are calculated. P90 normalization applied. Stakes redistributed via certainty-weighted slashes and proportional rewards.

7.3 Position Closure

Agent sells tokens to zero balance. Lock $L_{i,m}$ is released. Excess stake becomes withdrawable:

$$\text{Withdrawable} = S_i - \sum_{m \in \text{active}} L_{i,m} \quad (23)$$

8 Key Properties

The protocol exhibits the following properties:

- **Proportional influence:** Epistemic weight directly proportional to capital at risk
- **Automatic collateralization:** Seamless stake management via conditional skimming
- **Bounded risk:** Maximum loss per market bounded by lock amount
- **Zero-sum dynamics:** Total stake preserved across redistribution
- **Incentive compatibility:** Nash equilibrium at truthful reporting
- **Sybil resistance:** No advantage from identity fragmentation