# Mathematical Formalism for Voting Process 

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## Voting process

Voting is one of approaches to reach consensus in a decentralized environment (e.g., Bitcoin).

Problems of existing voting models (e.g., $x$-th percentile voting) :

- not all votes are taken into account
- some votes are given extraordinary weight, which theoretically allows $x \%$ attacks
- parameters of the voting process are not always grounded

Our proposal: rigorous mathematical model allowing to

- determine voting results algorithmically (with either an explicit expression or simple computational methods)
- take all votes into consideration
- maximize all voters' satisfaction with the voting result.

Additionally, our model is tunable and can be implemented for other voting processes in Bitcoin and beyond.

## Function

Lets maximize all voters satisfaction with the voting result via defining dissatisfaction function $F(s, v)$,
where

- $s$ is the target parameter chosen in voting (e.g., block size limit)
- $v$ is the vote

Suppose we have votes $v_{1}, v_{2}, \ldots, v_{n}$.

$$
L(s)=\sum_{i=1}^{n} w_{i} F\left(s, v_{i}\right) \rightarrow \min _{s}
$$

where $w_{i}$ are (optional) vote weights.

Then, we can solve for target $s$ by differentiating:

$$
L^{\prime}(s)=0
$$

## Two types of dissatisfaction functions:

$$
F(s, v)=D(s-v)
$$

A) absolute difference between a vote and the target

| $V$ | 1 | 2 | 4 |
| :---: | :---: | :---: | :---: |
| 1 mb | $D(0)$ | $D(1)$ | $D(3)$ |
| 2 mb | $D(-1)$ | $D(0)$ | $D(2)$ |
| 3 mb | $D(-2)$ | $D(-1)$ | $D(1)$ |
| 4 mb | $D(-3)$ | $D(-2)$ | $D(0)$ |
| 5 mb | $D(-4)$ | $D(-3)$ | $D(-1)$ |

$$
L(s=1)=\sum_{i=1}^{5} D\left(1-v_{i}\right)
$$

$$
F(s, v)=D(s / v-1)
$$

B) relative difference

| $V$ | 1 | 2 | 4 |
| :---: | :---: | :---: | :---: |
| 1 mb | $D(0 \%)$ | $D(+100 \%)$ | $D(+300 \%)$ |
| 2 mb | $D(-50 \%)$ | $D(0 \%)$ | $D(+100 \%)$ |
| 3 mb | $D(-67 \%)$ | $D(-33 \%)$ | $D(+33 \%)$ |
| 4 mb | $D(-75 \%)$ | $D(-50 \%)$ | $D(0 \%)$ |
| 5 mb | $D(-80 \%)$ | $D(-60 \%)$ | $D(-20 \%)$ |

- $V$ miners vote
- $S$ voting result
* $D$ should be a non-negative, continuously differentiable, convex function with the only zero at 0.


## Example

Example: quadratic dissatisfaction function

$$
D_{s q}(x)=\left\{\begin{array}{ll}
x^{2}, & \text { if } x \geq 0, \\
\alpha x^{2}, & \text { if } x<0 .
\end{array} \quad F(s, v)=\{1, \alpha\}(s-v)^{2}\right.
$$



Example of votes distribution

| Vote, Mb | 1 | 2 | 4 | 8 |
| :---: | :---: | :---: | :---: | :---: |
| Weight | $21 \%$ | $25 \%$ | $25 \%$ | $29 \%$ |

Block-size result with $\alpha$...

| $\alpha$ | 1 | 0.5 | 0.25 | 0.125 |
| :---: | :---: | :---: | :---: | :---: |
| Vote result, Mb | 4.03 | 3.25 | 2.59 | 2.13 |

Parameter $\alpha$ allows to select how much the voting result will be skewed:

- $\alpha=1$ corresponds to a weighted average of the votes
- $\alpha<1$ skews the target to lower values
- $\alpha>1$ skews the target to higher values


## Example

Example: exponential dissatisfaction function

$$
D_{\text {exp }}(x)=e^{\alpha x}+\alpha e^{-x}-\alpha-1
$$



Example of votes distribution

| Vote, Mb | 1 | 2 | 4 | 8 |
| :---: | :---: | :---: | :---: | :---: |
| Weight | $21 \%$ | $25 \%$ | $25 \%$ | $29 \%$ |

Resulting block-size with $\alpha$...

| $\alpha$ | 1 | 2 | 3 | 4 | 5 |
| :---: | :---: | :---: | :---: | :---: | :---: |
| Vote result, Mb | 4.47 | 3.40 | 2.82 | 2.46 | 2.22 |

Again, $\alpha$ determines skewness. Unlike the quadratic function, exponential function is infinitely continuously differentiable; however, optimization task requires more complex computations.

## Thank You

To Bitfury

Calculating voters' dissatisfaction for the quadratic function with skew parameter $\alpha=0.5$. Total dissatisfaction is the weighted sum of voters' dissatisfactions.



The optimal vote result 3.25 can be obtained by solving $L^{\prime}(s)=0$ on intervals $(1,2),(2,4)$ and $(4,8)$.

On interval (1, 2):

$$
\begin{aligned}
& L(s)=0.21(s-1)^{2}+0.25 \cdot 0.5(s-2)^{2}+0.25 \cdot 0.5(s-4)^{2}+0.29 \cdot 0.5(s-8)^{2}=0.605 s^{2}-4.24 s+11.99 \\
& L^{\prime}(s)=1.21 s-4.24=0 ; \quad s=4.24 / 1.21 \approx 3.504
\end{aligned}
$$

On interval (2, 4):

$$
\begin{aligned}
& L(s)=0.21(s-1)^{2}+0.25(s-2)^{2}+0.25 \cdot 0.5(s-4)^{2}+0.29 \cdot 0.5(s-8)^{2}=0.73 s^{2}-4.74 s+12.49 \\
& L^{\prime}(s)=1.46 s-4.74=0 ; \quad s=4.74 / 1.46 \approx 3.247
\end{aligned}
$$

On interval (4, 8):

$$
\begin{aligned}
& L(s)=0.21(s-1)^{2}+0.25(s-2)^{2}+0.25(s-4)^{2}+0.29 \cdot 0.5(s-8)^{2}=0.855 s^{2}-5.74 s+14.49 \\
& L^{\prime}(s)=1.71 s-5.74=0 ; \quad s=5.74 / 1.71 \approx 3.357
\end{aligned}
$$

Only the second value 3.247 is within the corresponding interval $(2,4)$, so we may skip checking other two values.

