A $O(2^{n/2})$ Universal Maximization Algorithm Research Paper • X-HEC Entrepreneurs 2023-2024

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Finite Sum Theorem.

Any real function with n boolean variables can be expressed as a unique finite sum of 2^n real coefficients multiplied by a finite product of boolean variables.

$\forall f: \{0,1\}^n \to \mathbb{R}$ $\forall p \subseteq \{1, \dots, n\} \exists ! a_p \in \mathbb{R}$ $\forall x \in \{0,1\}^n$ $f(x) = \int a_p$ $p \subseteq \{1, \ldots, n\}$ $l \in p$

Finite Sum Theorem. Proof. Using Lagrange Polynomials and the binary bijection between $y \in \{0,1\}^n$, can be expressed as $P_v(x)$, a polynomial evaluation of the natural integer expression of x.

matter f.

coefficient.

 $\{0,1\}^n$ and $\{0,1,2,...,2^n - 1\}$, we show that the indicator $1_{x=v}$ for a given

Considering $x = x_1 + 2x_2 + 4x_3 + ... + 2^{n-1}x_n$ and $x_i^k = x_i$, it becomes clear that the binomial formula leads to the existence of the formula no

Unicity comes from contradiction by subtracting two potential candidates, and evaluating the result for a given x that would highlight a non-zero









Finite Sum Theorem. What can be said about the real coefficients that compose the sum?

$\forall p \subseteq \{1, \dots, n\} : a_p =$

Can we derive an estimator for the coefficients without relying on exponential computation? Empirical results suggests that it is possible in most cases, but as we will discuss in this defense, this may not be the most efficient approach for constituting an actionable formula for a significant number of boolean variables.

$$\sum_{q \subseteq p} (-1)^{|p| - |q|} f\left(\sum_{i \in q} x_i 2^{i-1}\right)$$



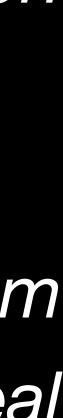
The relationship with WV the set of real-weighted MAXSAT instances



What is W?

W is the collection of positively-real weighted clauses on the boolean variables, plus a real constant. We can show a couple of properties on W, using the argument: $\bigvee x_i + \bigwedge \neg x_i = 1$ $i \in p$ $i \in p$ ${f: {0,1}^n \to \mathbb{R}} \subset \mathbb{W}$. This result is a consequence of the finite sum theorem, note that there can be multiple candidates w to model a real function with n boolean variables, depending on the clauses.

 $\mathbb{W} = \left\{ w = c + \sum_{i=1}^{n} w_i C_i : w_i \ge 0 \right\}$



Why using W?

W is the collection of positively-real weighted clauses on the boolean variables, plus a real constant. Assuming you have constructed an instance $w \in \mathbb{W}$, such that $\forall x \in \{0,1\}^n$, f(x) = w(x), you can use MAXSAT logic to determine x^* the optimum of f from the study of w.

MAXSAT is NP-HARD, in practice large instances can only approximate the optimum.

binomial distribution, the expected cardinal of p is $\frac{n}{2}$.

Overall the algorithm has a complexity $O(2^{n/2})$ because each iteration of the while loop relies on the brute-force computation of $2^{|p|}$ coefficients.

There is no need for constructing the entire instance in order to start deriving an approximation algorithm. The idea is to sample a subset $p \subseteq \{1, \ldots, n\}$ at random. From









Appleatons.

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such that $f(x^*) \ge f(x)$ in $O(2^{n/2})$ complexity. the tradeoff being between precision and computational efficiency.

$$\forall p \subseteq \{1, \dots, n\} : a_p \approx \frac{2^{|p|}}{|H(p)|} \sum_{q \in H(p)} (-1)^{|p| - |q|} f\left(\sum_{i \in q} x_i 2^{i-1}\right) : H(p) \subset \mathcal{P}(p)$$

Let $f: \{0,1\}^n \to \mathbb{R}$, be a cost function. According to the Finite Sum Theorem, and by definition of W we know that there exists a model $w \in W$ from which we can derive x^*

Although multiple algorithms already exist and are documented on MAXSAT, we present an alternative algorithm that relies on the ability to approximate efficiently a_p through the following formula. Note that because this is an approximation, it can have linear complexity,



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coefficients $\hat{a}_{p_1}, \ldots, \hat{a}_{p_N}$, ensuring that : $\forall i : a_{p_i} \leq 0 \Leftrightarrow \hat{a}_{p_i} \leq 0$ the sign is preserved. $\forall i, j : a_{p_i} \leq a_{p_i} \Leftrightarrow \hat{a}_{p_i} \leq \hat{a}_{p_i}$ the order relationship is preserved. $\forall i, j : |a_{p_i}| \le |a_{p_i}| \Leftrightarrow |\hat{a}_{p_i}| \le |\hat{a}_{p_i}|$ the 'abs' order relationship is preserved.

- Let $\hat{a}_p \approx a_p$ be the O(poly(n)) approximation of the exact coefficient a_p . One possible approach to solving the model $w \in \mathbb{W}$ derived, is to sample N
- We will now assume $|\hat{a}_{p_1}| \ge |\hat{a}_{p_2}| \ge ... \ge |\hat{a}_{p_n}|$ that the coefficients are sorted.

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Idea:

$$\hat{a}_{p_1} > 0 \Rightarrow \forall t \in p_1, x_t^* = 1$$

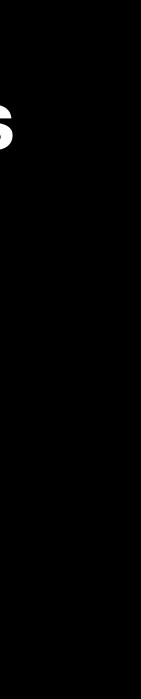
We derive a logic condition from \hat{a}_{p_1}

$$\hat{a}_{p_2} > 0 \Rightarrow \forall t \in p_2, x_t^* =$$

We derive another logic condition from \hat{a}_{p_2} , if conflicting with an above condition, the above condition is preferred, until the total conflicting contributions of below conditions are greater than the initial one. In our example, if $|\hat{a}_{p_2}| + |\hat{a}_{p_3}| > |\hat{a}_{p_1}|$ it can be rational to drop the logic condition derived from \hat{a}_{p_1} .

 $and \hat{a}_{p_1} \leq 0 \Rightarrow \exists t \in p_1, x_t^* = 0$

 $1 \text{ and } \hat{a}_{p_2} \leq 0 \Rightarrow \exists t \in p_2, x_t^* = 0$



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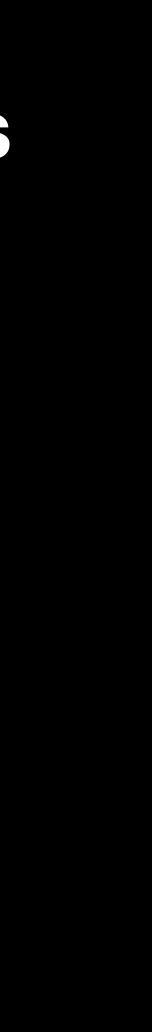
Mitigations:

$$\begin{aligned} \forall i, j: |a_{p_i}| \leq |a_{p_j}| \Leftrightarrow |\hat{a}_{p_i}| \leq |\hat{a}_{p_j}| \text{ doesn't imply} \\ \forall i, j, k: |a_{p_i}| + |a_{p_j}| > |a_{p_k}| \Leftrightarrow |\hat{a}_{p_i}| + |\hat{a}_{p_j}| > |\hat{a}_{p_k}| \end{aligned}$$

The sample p_1, \ldots, p_N must be representative of the whole.

Even more so as you keep stacking approximation errors.

There are still 2^n candidates for p, a high entropy function close to $1_{x=x^*}$ would not be solvable through this method.



Computing Complex Polynomial Roots A framework for Universal Polynomial Root Approximation

Let $z \in \mathbb{C}$, we call N the precision parameter. We introduce a function $\phi_{\mathbb{C}}$: $\{0,1\}^{4(N+1)} \to \mathbb{C}$, that satisfies the property:

$$|z| < 2^{N} \Rightarrow \min_{\substack{x \in \{0,1\}^{4(N+1)} \\ \text{where:}}} |z - \phi_{\mathbb{C}}(x)| \le 2^{-N}$$

where:
$$\phi_{\mathbb{C}}(x) = (-1)^{x_{1}} \sum_{k=-N}^{N} x_{N+k+2} 2^{k} + i(-1)^{x_{2N+3}} \sum_{l=-N}^{N} x_{3N+4+l} 2^{l}$$

Computing Complex Polynomial Roots A framework for Universal Polynomial Root Approximation

Let $P: X \mapsto \sum_{k=1}^{n} z_k X^k$ be a complex polynomial of $\mathbb{C}[X]$.

 $f: x \mapsto f$

 $f(x^*) \ge \inf_{\substack{\theta \in [-\pi] \\ \lambda \in [0, \infty]}} f(x^*)$

One way of seeing this is to call $Z = 2^{-N} + \max |z^* - z'|$ to obtain the following upper bound to the approximation z':P(z')=0 $|f(x^*)| \le Z^{n-1}2^{-N}$

Practical use for higher degree Polynomials suggest using the alternative logarithmic cost function:

$$f: x \mapsto -\log$$

$$- |(P \circ \phi_{\mathbb{C}})(x)|$$

According to the Finite Sum Theorem, under the assumption that there exists a complex root $z^* \in \mathbb{C}$ to $P \in \mathbb{C}[X]$ following $|z^*| < 2^N$

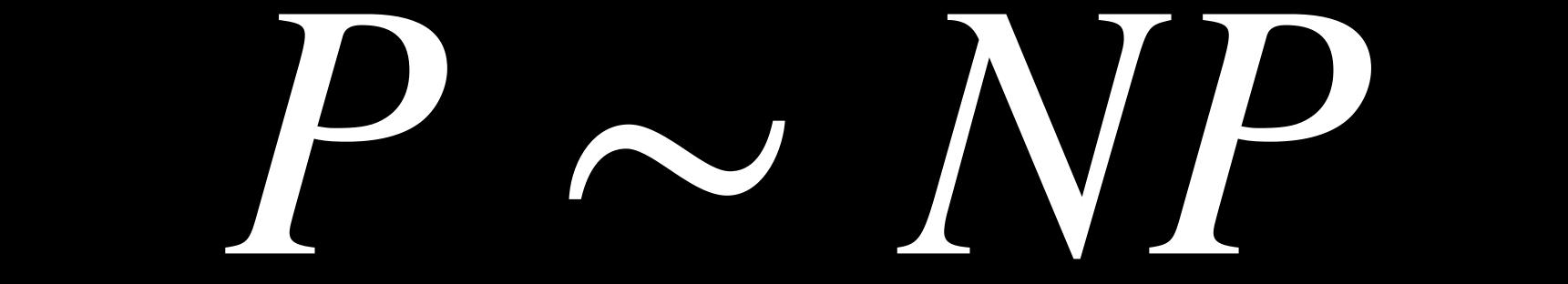
$$\prod_{\substack{\lambda,\pi \\ 1}} - \left| P(z^* + \lambda 2^{-N} e^{i\theta}) \right|$$

Such a quantity rapidly converges to 0^- as $N \to +\infty$

 $(1 + |(P \circ \phi_{\mathbb{C}})(x)|)$

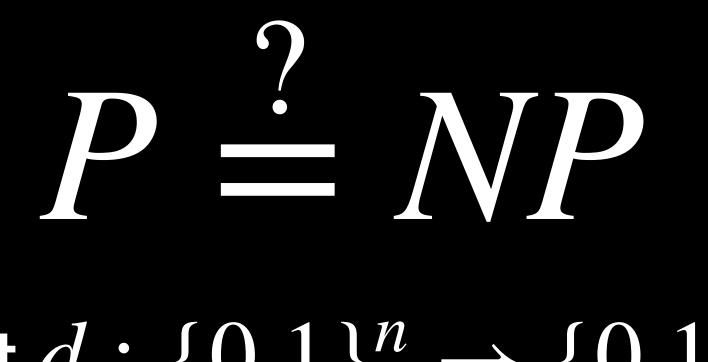


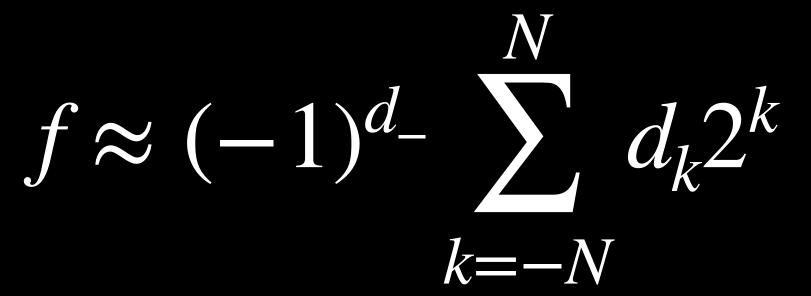
We have an A^* framework for Maximising $f: \{0,1\}^n \to \mathbb{R}$.



Directions.

- Let $d: \{0,1\}^n \to \{0,1\}$.
- I: There exists a SAT instance that models d.
- II : The complexity of generating such a SAT Instance is O(poly(n)).
- III : Real functions with n boolean variables are equivalent to a finite collection of SAT Instances.







Exploration on the practical Business Implications of the Algorithm

Research & Large-Scale Experiment

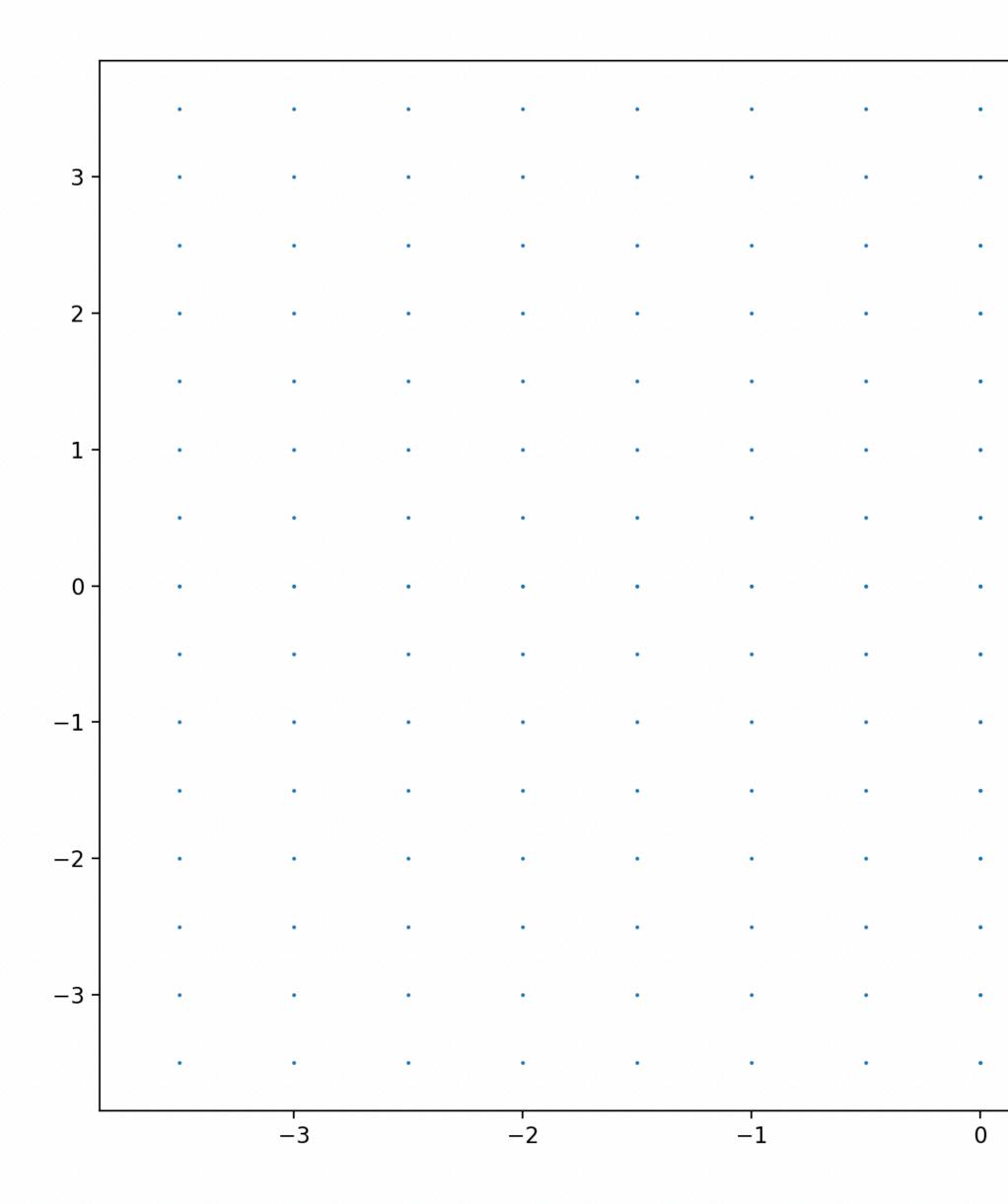
What next?

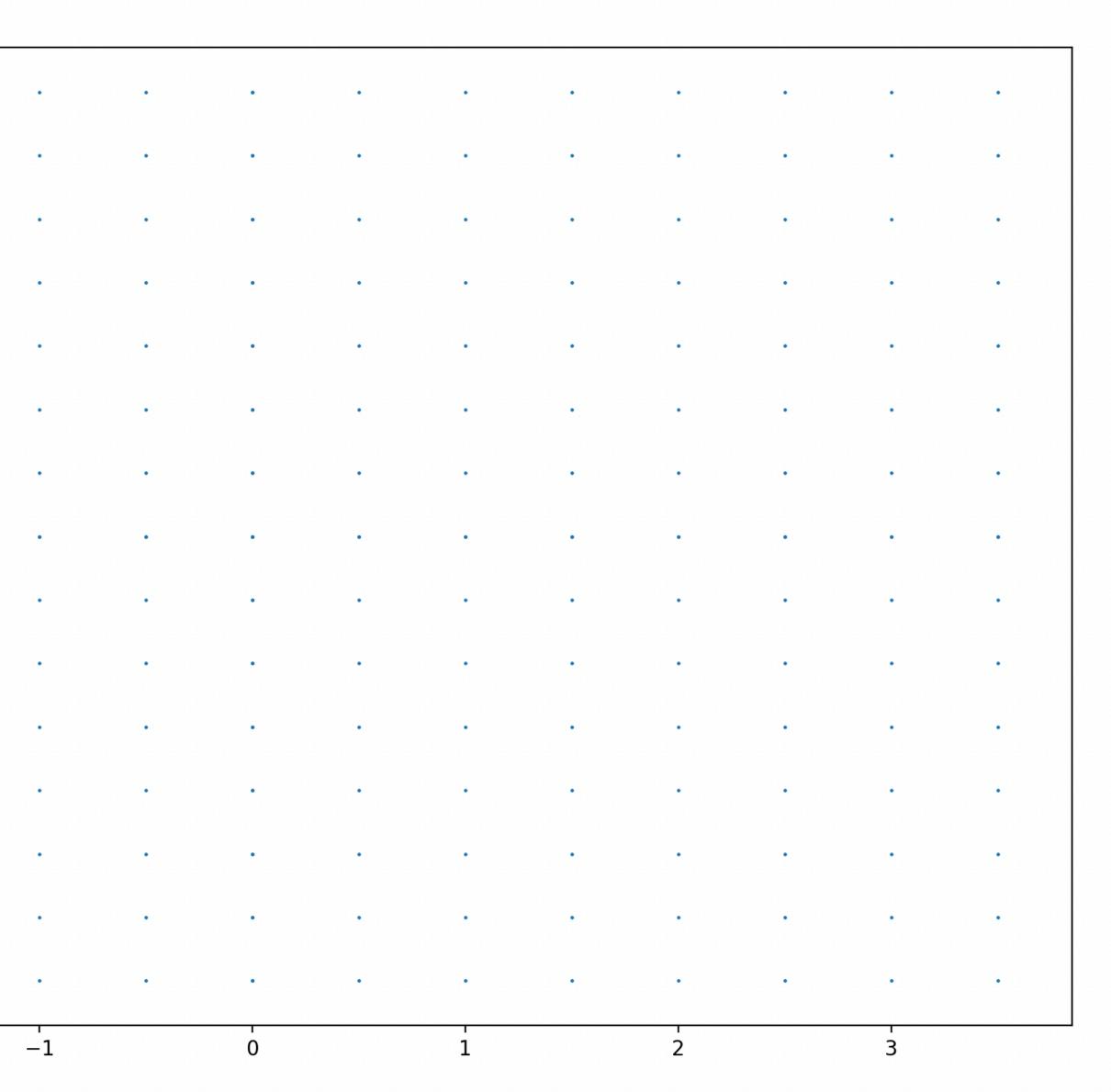
EU Funding Application

Graphical Intuition for $\phi_{\mathbb{C}}$

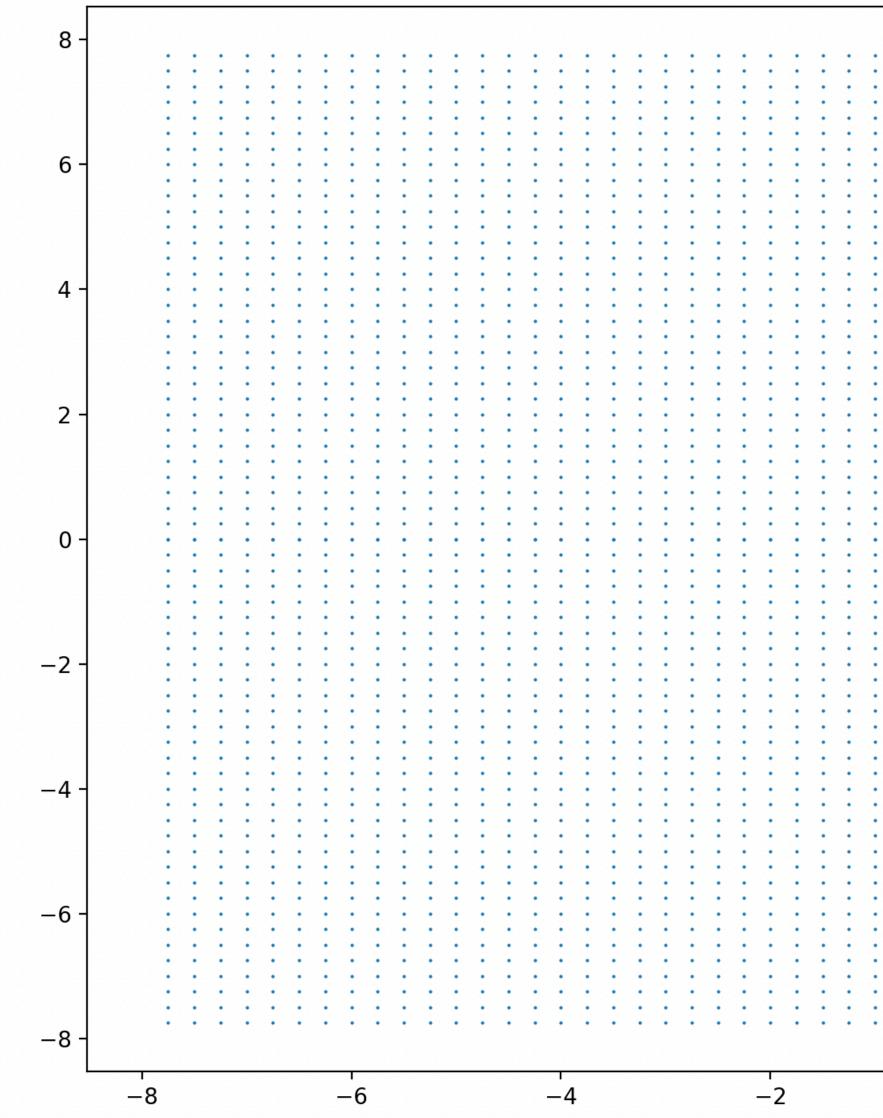
Precision parameter from N = 1...4

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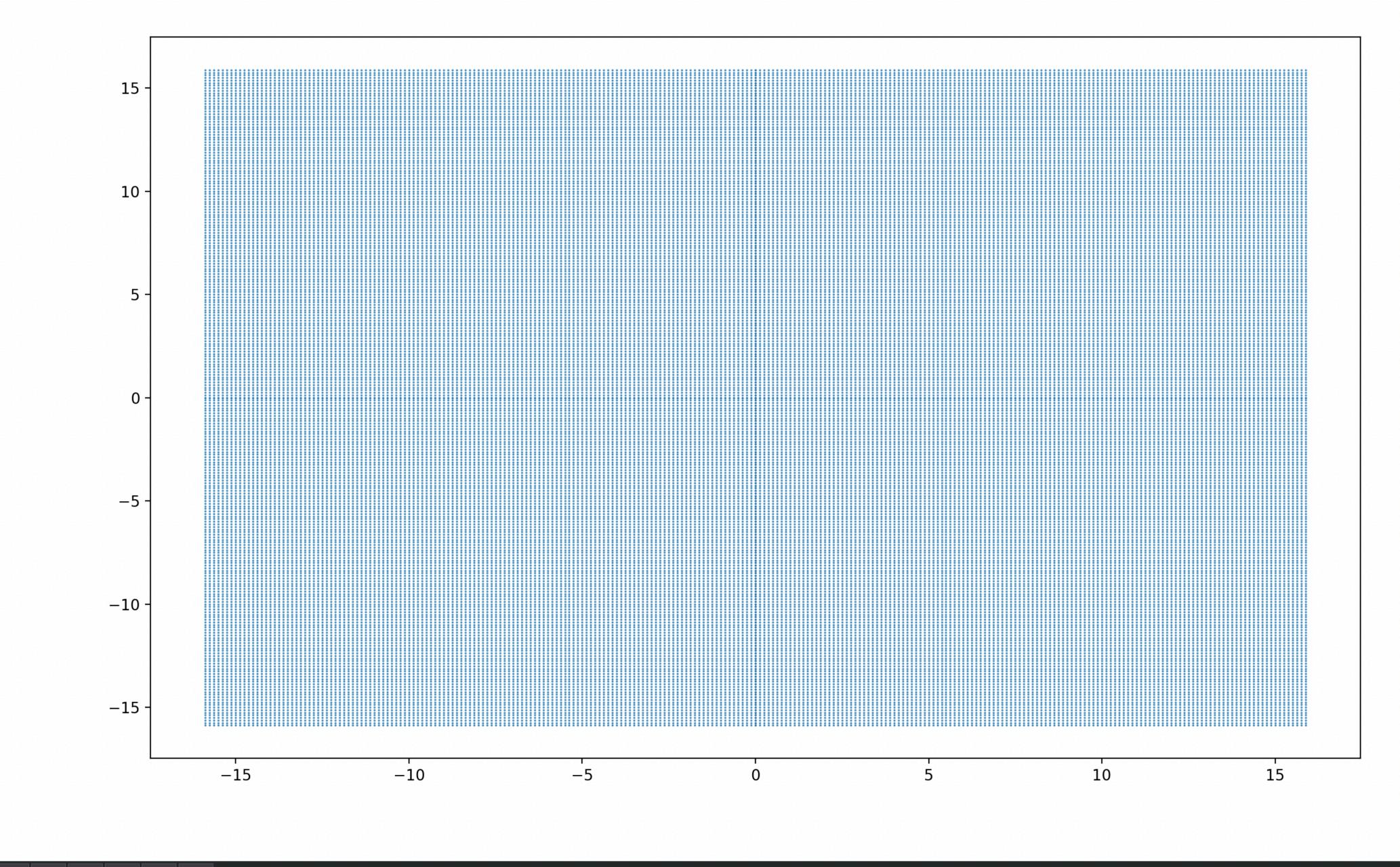








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