# **STARK Math**

Introduction

Explanation by Example

Commit to the Trace Polynomials

Commit to the Constraint Polynomial

Probabilistic Verifier Queries

FFT and IFFT Introduction

# Introduction

This document serves as an overview of STARK math, providing deep insights whilst not going too much into formal verification of axioms. The idea is to provide an implementation specification for developers who seek to code STARKs from scratch, or for mathematicians who are looking for a summary of the methods involved in constructing basic STARKs.

# **Explanation by Example**

Most of the information in this section stems from the <a href="Stark 101">Stark 101</a> series by <a href="Starkware">Starkware</a>. The content has been modified to convey the core principles in a more straight forward manner, eliminating the need to read their Python code line-byline. Some details related to domain choices have been omitted whilst other sections of the source material have been extended.

#### STARK 101

STARK 101 is a hands-on tutorial on how to write a STARK prover from scratch (in Python).

https://starkware.co/stark-101/

#### We start with an example of a trace table:

column	Х	у
0	10	20

1	30	40
2	70	80

The trace table represents the trace of an arbitrary computation. This could be the output of a domain specific program written in a DSL like Circom, or the trace of a Virtual Machine as is the case with most STARK proving systems (Risco, SP1, ZKEVMS).

We interpolate our trace polynomials for x and y using either FFT or lagrange interpolation:

```
g(x) = Interpolate((0,10),(1,30),(2,70))
```

$$h(x) = Interpolate((0,20), (1,40), (2,80))$$

Next we define a constraint polynomial that evaluates to o for our domain:

$$C(x) = h(x) - g(x) - 10$$

This constraint polynomial may not be inherently useful, however it conveys the idea of how we can enforce constraints. Multiple constraint polynomials (c1, c2, ..., cn) can be combined into a single constraint polynomial:

$$C(x) = c1(x) + c2(x) + ... + cn(x)$$

## **Commit to the Trace Polynomials**

We commit to all trace polynomials by evaluating them over an extended domain and inserting the evaluations into a merkle tree.

The extended domain is constructed from powers of g. g is a generator of a finite field:

$$D = g^0, g^1, g^2..g^N$$

Where  $\mathbb{N}$  is the size of the multiplicative subgroup and should be a power of  $\mathbb{Z}$ . The advantages of choosing a power of  $\mathbb{Z}$  will be explained later in this document.

For each trace polynomial we usually construct an individual merkle tree from the evaluations over the extended domain. The evaluations could in theory be combined into a single tree, but this adds overhead when performing probabilistic queries for specific constraints.

Since we utilize **FFT** under the hood of **FRI** for efficient polynomial math it is best to choose an **extended domain** that is a power of 2.



When the size of the input is a power of two,

$$N=2^n$$

it allows FFT to evenly split the problem in half at each recursive step. This leads to a natural and efficient division, reducing the time complexity of the algorithm's complexity to:

$$O(N \log(N))$$

is not a power of two, splitting evenly becomes more complex, leading to less efficient computation.

### **Commit to the Constraint Polynomial**

The process of committing to the constraint polynomial is more complex than that of committing to the trace polynomials since it involves [FR]:

Anatomy of a STARK, Part 3: FRI

STARK tutorial with supporting source code in python.

https://aszepieniec.github.io/stark-anatomy/fri.html

For the first round of our FRI folding we evaluate the constraint polynomial over the extended domain and once again commit the results to a merkle tree. For each additional folding step, until we reach a low degree polynomial, we fold the degree of the polynomial and the extended domain in half.

We evaluate the <code>folded polynomial</code> over the <code>folded domain</code> in each step and commit the results to <code>merkle trees</code>. Ultimately we are left with a <code>merkle tree</code> for each folding step.

### **Probabilistic Verifier Queries**

Now that we have constructed our merkle trees for the evaluations of the trace polynomials and the folded constraint polynomial, the verifier can issue random challenges (interactive proving). Fiat-shamir transformation can be utilized to make this protocol non-interactive, but for the scope of this section we will assume an interactive setup.

For each query the prover must reveal not only the evaluations of the trace polynomial over the originally extended domain, but also the corresponding values from the FRI merkle trees. If the values are persistent for each step in FRI and therefore satisfy the rules of our constraint polynomial, the query is considered valid.

The verifier is not required to fold the polynomials but instead employs a consistency check for each step in the queried evaluations. The formula for the consistency check looks like this:

$$f'(x) = rac{f(x) + f(g \cdot x)}{2}$$

Where g is a generator of the finite field over which the extended domain is defined. f'(x) is the next folded polynomial and f(gx) is the preceding polynomial.

A generator g can be brute-forced and must satisfy:

$$g^{(q-1)/p_i} 
ot\equiv 1 \pmod q$$

Where  $\mathbf{q}$  is the size of a field and  $\mathbf{pi}$  are all possible prime factors of  $\mathbf{q-1}$ . In commonly used fields  $\mathbf{q}$  is chosen such that factoring the prime factors of  $\mathbf{q-1}$  is feasible.

The amount of queries that are necessary for the verifier to trust the prover can be calculated according to this probability formula:

 $(1-\frac{1}{m})^k$  , where  ${f k}$  is the amount of queries and  ${f m}$  the amount of columns in the trace .



The more columns we have in the trace the more queries we need to perform in order to be able to trust the prover with negligible fraud tolerance.

#### **FFT and IFFT Introduction**

We utilize FFT to convert a polynomial from its coefficient representation to its value representation.

IFFT on the other hand is the inverse process of converting the polynomial back to the coefficient representation.

The size of the value representation depends on the size of the extended domain we choose to evaluate our polynomial over.

By using the formula for each point in our value representation, we can get the set of points required to interpolate the next folded polynomial using IFFT:

$$f'(x) = rac{f(x) + f(g \cdot x)}{2}$$

As explained earlier the verifier will only need to check the consistency of the folding steps in combination with verifying the merkle queries and won't need to re-compute the entire folding process.