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Lucrare de licență

FILE CORRUPTION REPAIR USING REED-SOLOMON CODES

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Abstract

The aim of this project is to implement file corruption detection and repair using Reed-Solomon error-correcting codes.

The code used is an erasure code over the field GF(2^{64}), implemented using $O(n \log n)$ transforms in a polynomial basis introduced by Sian-Jheng Lin, Wei-Ho Chung, and Yunghsiang S. Han in [3].

Finite field multiplication is implemented using carry-less multiplication, also known as XOR multiplication, and division is implemented using the extended Euclidean algorithm.

The implementation is a command-line utility which can generate parity data for a file, and later detect and repair corruption in that file.

The file is split into N blocks, and an arbitrary number of parity blocks M can be generated, for a total of $N + M$ blocks. Any N blocks are sufficient to recover the original data, so up to M corrupted blocks can be repaired.

As erasure codes require known error locations - the term 'erasure' refers to an error at a known location - a hash of each block is stored in the file header to detect corruption.

The file is not a single Reed-Solomon code, as that would require reading the entire file into memory, limiting the maximum file size which can be processed. Instead, the file is interpreted as a matrix with $N + M$ rows, and each column is an separate Reed-Solomon code.

Source code available at https://github.com/theo543/rsarc

Rezumat

Acest proiect are ca scop implementarea detecției și reparării coruperii fișierelor folosind coduri Reed-Solomon de corectare a erorilor.

Codul folosit este un cod de ștergere peste corpul GF(2^{64}), implementat folosind transformări în timp $O(n \log n)$ într-o bază polinomială introdusă de Sian-Jheng Lin, Wei-Ho Chung, și Yunghsiang S. Han în [3].

Înmulțirea în corp finit este implementată folosind înmulțire fără retenție ('carry-less'), uneori cunoscută ca înmulțire XOR, iar împărțirea este implementată folosind algoritmul extins al lui Euclid.

Implementarea este un utilitar de linie de comandă care poate genera date de paritate pentru un fișier și, ulterior, să detecteze și repare corupție în fișier.

Fișierul este împărțit în N blocuri, și un număr arbitrar de blocuri de paritate M pot fi generate, pentru un total de $N + M$ blocuri. Orice N blocuri sunt suficiente pentru a recupera datele originale, deci cel mult M blocuri corupte pot fi reparate.

Deoarece un cod de ștergere necesită cunoașterea locațiilor erorilor - termenul 'ștergere' înseamnă o eroare la o locație cunoscută - un hash al fiecărui bloc este stocat în antetul fișierului pentru a detecta corupție.

Fișierul nu este un singur cod Reed-Solomon, deoarece ar necesita citirea întregului fișier în memorie, limitând dimensiunea maximă de fișier care poate fi procesată. De fapt, fișierul este interpretat ca o matrice cu $N + M$ rânduri, iar fiecare coloană este un cod separat Reed-Solomon.

Cod sursă disponibil la https://github.com/theo543/rsarc

Contents

Chapter 1

Introduction

1.1 Reed-Solomon Codes

Reed-Solomon codes are a well-known class of error-correcting codes used in a wide range of applications, from data storage to radio communication. They are based on polynomials over finite fields. [4]

The code used for this project is an erasure code over the field GF(2^{64}), implemented using $O(n \log n)$ algorithms introduced in [3].

The basic working principle of this type of Reed-Solomon code is the interpretation of data as values of a polynomial evaluated at points $\omega_0, \omega_1, \ldots, \omega_{k-1}$ in a finite field GF(2^n). As there is only one polynomial of degree $k-1$ or smaller passing through k points, any combination of at least k of the original and redundant points uniquely determines the original polynomial.

Any amount of redundancy can be added, limited only by the field size. As the chosen field is GF(2^{64}), the limit is effectively infinite.

An erasure code is a type of error-correcting code which requires that the locations of corrupted data are known. The code cannot be used to discover the locations of corrupted data by itself. In this case, hashes stored in the metadata of the parity file are used to determine error locations.

Other Reed-Solomon codes do locate errors without requiring hashes, but they are not used in this project, as hashes are a simpler and more efficient solution.

One common code is RS$(255, 223)$, which is used in CDs and DVDs, and uses 8-bit symbols

(in the field GF(2^8)). The notation RS(n, k) denotes a code with n total symbols, with k data symbols and $n - k$ parity symbols. The code used in this project has no fixed n or k, they are specified by the user.

The code used in this project is also systematic, meaning that the original data is included in the output. Non-systematic codes do not include the original data, so the receiver must decode the code to obtain the original data, even if no corruption occurred.

1.2 Finite Fields

Finite fields, also known as Galois fields, are mathematical structures which define addition, multiplication, subtraction, and division over a finite set of elements [2] (as opposed to the better-known infinite fields, such as the rationals, reals, and complex numbers).

A field must satisfy the following properties:

- Associativity of addition and multiplication: $(a+b)+c = a+(b+c)$ and $(a \cdot b) \cdot c = a \cdot (b \cdot c)$
- Commutativity of addition and multiplication: $a + b = b + a$ and $a \cdot b = b \cdot a$
- Additive and multiplicative identity elements: $a + 0 = a$ and $a \cdot 1 = a$
- Additive inverses: for every a, there exists $-a$ such that $a + (-a) = 0$
- Multiplicative inverses: for every $a \neq 0$, there exists a^{-1} such that $a \cdot a^{-1} = 1$
- Distributivity of multiplication over addition: $a \cdot (b + c) = a \cdot b + a \cdot c$

The theorems of polynomial mathematics used in Reed-Solomon codes only hold in a field, however standard computer arithmetic does not form a field. Typical arithmetic supported natively by CPUs is fixed-size binary arithmetic with overflow, which is equivalent to arithmetic modulo a power of 2. Modular arithmetic only forms a field with a prime modulus, so it cannot be used directly for Reed-Solomon codes.

For example, the operation $x \cdot 2$ is not invertible, as it is equivalent to a left shift, from which the most significant bit of x cannot be recovered.

Fortunately, it is possible to construct a field based on fixed-size integers, such as 64-bit integers.

In a finite field GF(p^m), where p is a prime number and m is a positive integer, the elements are polynomials of degree $m - 1$, with coefficients in GF(p). For the GF(2^n) case, an element

in the field is a polynomial with n coefficients, where each coefficient is a bit (i.e. a value in GF$(2) = \{0, 1\}$). For the purposes of this project, n is always 64, so the field is GF(2^{64}).

The notation ω_i is used to denote the integer i converted to an element of the field GF(2^{64}) by interpreting its bits as a polynomial, which is a no-op in code, as elements of GF(2^{64}) are stored as 64-bit integers.

It is important to note that these polynomials are not the same as the ones used in Reed-Solomon codes to represent data and parity information. Elements of GF(2^n) are simply n bit integers with more complex arithmetic. They are polynomials over GF(2), with n coefficients. Reed-Solomon polynomials can be arbitrarily long. They are polynomials over GF(2^{64}), with an arbitrary number of coefficients, and each coefficient is itself a polynomial over GF(2^2) with 64 coefficients.

Finite field addition is defined as polynomial addition. In a general field GF(p^m), this would be implemented as pairwise addition of the coefficients of two polynomials, modulo p.

In binary finite fields, addition is equivalent to XOR, as the coefficients are bits. Therefore, $x + x = 0$, and $x = -x$ (the field has characteristic 2).

Multiplication is defined as polynomial multiplication, followed by reduction modulo an irreducible polynomial of degree 64 (with 65 coefficients, where the highest coefficient is 1).

The irreductible polynomial used for this project is $x^{64} + x^4 + x^3 + x + 1$ [5]. The choice of irreducible polynomial does not affect correctness, and the fields obtained from different choices are isomorphic.

A simple but inefficient formula for the multiplicative inverse, used in the early stages of this project, is $x^{-1} = x^{2^{64}-2}$, computed using exponentiation by squaring.

The extended Euclidean algorithm is more efficient, and is used in the final implementation.

Chapter 2

Finite Field Arithmetic

As previously mentioned, Reed-Solomon codes require the use of non-standard arithmetic - arithmetic over finite fields - because modular arithmetic with a non-prime modulus does not have an inverse for all elements.

Addition in GF(2^{64}) is extremely simple, as it is equivalent to XOR. Multiplication, however, is less efficient than standard multiplication, and divison even less so.

The constant POLYNOMIAL refers to the irreducible polynomial $x^{64} + x^4 + x^3 + x + 1$, with x^{64} omitted, as it would not fit in a 64-bit integer.

2.1 Russian Peasant Algorithm

The Russian peasant algorithm multiplies two values in GF(2^{64}) without requiring 128-bit integers. It incrementally performs the multiplication by adding intermediate values into an accumulator, and slowly shifting the values to be multiplied and applying polynomial reduction.

The state of the algorithm consists of the two values to be multiplied a and b, and an accumulator.

The algorithm must be executed at most 64 times, before b is guaranteed to become zero. Then, the accumulator contains the result.

At each iteration, if the low bit of b is set, the accumulator is XORed with a. Then, a is shifted left, and b is shifted right.

This is justified because, at each step, we multiply the lowest coefficient of b with a, and

add the result (either 0 or a) to the accumulator. Then, moving on to the next coefficient of b, we divide b by x and multiply a by x, which is equivalent to shifting a left and b right.

If the high bit of a was set before shifting, a is XORed with the irreducible polynomial. This is because, conceptually, a now has a 65th bit (a coefficient x^{64}), which requires reduction, done by subtracting the irreducible polynomial using XOR.

Algorithm 1 Russian Peasant Multiplication

```
function MULTIPLY(a, b)
    acc ← 0
    for i ← 1 to 64 do
        if b & 1 then
            acc ← acc ⊕ a
        end if
        carry ← a & (1 ≪ 63)
        a ← a ≪ 1
        b ← b ≫ 1
        if carry then
            a ← a ⊕ POLYNOMIAL
        end if
    end for
    return acc
end function
```

This algorithm is fairly simple and easy to implement, but multiplication can be done more efficiently on modern CPUs with special instructions. Still, this algorithm is necessary as a fallback, for CPUs which don't support 128-bit carry-less multiplication.

2.2 Carry-less Multiplication

GF(2^{64}) multiplication can be performed using three 128-bit carry-less multiplication operations. Modern CPUs have support for this operation, as it is useful for cryptographic algorithms, computing checksums, and other applications. [1]

The terms "upper half" and "lower half" will be used to refer to the most significant 64 bits and least significant 64 bits of a 128-bit integer, respectively.

By multiplying a and b using carry-less multiplication, we obtain a 128-bit result. We must reduce the upper half to a 64-bit result, which can then be XORed with the lower half to obtain the final result.

This can be done by multiplying the upper half of the result by the irreducible polynomial. Then, the lower half of the result is the product reduced modulo the irreducible polynomial.

To understand why this works, consider the process of reduction. The irreducible polynomial is aligned with each set bit in the upper half of the result, and XORed with the result. This is effectively what carry-less multiplication does.

There is a complication, however. A third multiplication is required to ensure full reduction, as the highest bits of the upper half can affect the lowest bits of the upper half.

For example, consider $x^{127} + x^{67} + x^{66} + x^{64}$. After aligning the irreducible polynomial with the highest bit and XORing, all bits in the upper half are zero. At this point, the reduction is complete, but the multiplication does not know to stop here. The irreducible polynomial will also be aligned with the other three bits, and the lower half is XORed with some unnecessary values.

The upper half of the reduced result indicates if and where this happened. A third multiplication is used to correct this. The unnecessary XORs are undone by XORing with the lower half of the third multiplication.

For fields where $x^n + 1$ is irreducible, the algorithm simplifies to carry-less multiplication followed by XORing the upper and lower halves of the result. This is the case for $x^{63} + 1$, but is unfortunately not the case for $x^{64} + 1$ [5].

The justification for the algorithm may seem somewhat complex, but the algorithm itself is very short, simple, and efficient.

Algorithm 2 Carry-less Multiplication

function MULTIPLY(a, b)
    result $\leftarrow$ CLMUL(a, b)
    result_partially_reduced $\leftarrow$ CLMUL(upper(result), POLYNOMIAL)
    result_fully_reduced $\leftarrow$ CLMUL(upper(result_partially_reduced), POLYNOMIAL)
    return lower(result) $\oplus$ lower(result_partially_reduced) $\oplus$ lower(result_fully_reduced)
end function

2.3 Extended Euclidean Algorithm

The polynomial extended Euclidean algorithm, given polynomials a and b, computes s and t such that $a \cdot s + b \cdot t = \gcd(a, b)$. When b is set to the irreducible polynomial, t is the multiplicative inverse of a. s does not need to be computed.

The algorithm uses repeated Euclidean division. Because the irreducible polynomial is of degree 64, the first Euclidean division iteration, in the first iteration of the Euclidean algorithm, is a special case. As a 65-bit polynomial cannot fit in the 64-bit variable b, the first iteration is done manually, outside the loop.

In the following pseudocode, leading_zeros(x) returns the number of leading zero bits in x. Modern CPUs have a dedicated instruction for counting leading zeros.

Algorithm 3 Extended Euclidean Algorithm

```
function EXTENDEDEUCLIDEAN(a)
    assert(a ≠ 0)
    if a = 1 then return 1 endif
    t ← 0
    new_t ← 1
    r ← POLYNOMIAL
    new_r ← a
    r ← r ⊕ (new_r ≪ (leading_zeros(new_r) + 1))
    quotient ← 1 ≪ (leading_zeros(new_r) + 1)
    while new_r ≠ 0 do
        while leading_zeros(new_r) >= leading_zeros(r) do
            degree_diff ← leading_zeros(new_r) − leading_zeros(r)
            r ← r ⊕ (new_r ≪ degree_diff)
            quotient ← quotient|(1 ≪ degree_diff)
        end while
        (r, new_r) ← (new_r, r)
        (t, new_t) ← (new_t, t ⊕ gf64_multiply(quotient, new_t))
        quotient ← 0
    end while
    return t
end function
```

Chapter 3

Polynomial Oversampling and Recovery

Standard algorithms for polynomial interpolation and evaluation, such as Newton interpolation and Horner's method, require $O(n^2)$ time. Efficient $O(n \log n)$ algorithms are used instead, based on FFT-like transforms introduced in [3].

3.1 Polynomial Basis

The polynomial basis $\mathbb{X} = \{X_0, \ldots, X_{2^{64}-1}\}$ admits transforms Ψ_h^l and $(\Psi_h^l)^{-1}$ which convert between values at h contiguous points with an arbitrary offset l and coefficients in $\mathbb{X}$, with h a power of two.

To encode a $RS(n, k)$ code, the data polynomial coefficients are obtained by applying $(\Psi_h^0)^{-1}$ to the input values, then additional values are obtained using Ψ_h^l $\frac{n}{k}$ times at offsets $l = k, 2k, \ldots, n - k$.

The basis polynomials X_i are defined as the products of polynomials $\hat{W}_j$ corresponding to the bits of the index i:

$$X_i = \prod_{j \in \text{bits}(i)} \hat{W}_j$$

$\hat{W}_i = \frac{W_i}{W_i(2^i)}$ is a normalized vanishing polynomial of degree 2^i, which vanishes (i.e. evaluates to zero) at the points $\omega_0, \omega_1, \ldots, \omega_{2^i-1}$, and evaluates to 1 at ω_{2^i}.

$$\hat{W}_i(x) = \frac{W_i(x)}{W_i(2^i)} = \frac{\prod_{j=0}^{2^i-1}(x - \omega_j)}{\prod_{j=0}^{2^i-1}(\omega_{2^i} - \omega_j)}$$

$\hat{W}_i$ has degree 2^i, as it is the product of 2^i degree one factors divided by a constant. Therefore, X_i has degree i, since it the product of W_j corresponding to the bits set in i. Since $\mathbb{X}$ contains 2^{64} polynomials with all degrees from 0 to $2^{64}-1$, it automatically is a valid basis for representing polynomials of degree up to $2^{64}-1$.

All W_i are linearized polynomials, which means they only have non-zero coefficients at power-of-two indices and are additive:

$$W_i(x+y) = W_i(x) + W_i(y)$$

Note that the standard monomial basis $\{1, x, x^2, \ldots, x^{2^{64}-1}\}$ could also be defined in a similar way, with $\hat{W}_i = X^{2^i}$, but that would not allow $O(n \log n)$ FFT-like transforms.

3.2 Forward and Inverse Transforms

Let D_h be the data polynomial with h coefficients $d_0, d_1, \ldots, d_{h-1}$. It can be expressed as a recursive function $\Delta_i^m(x)$, with $D_h(x) = \Delta_0^0(x)$:

$$\Delta_i^m(x) = \begin{cases} \Delta_{i+1}^m(x) + \hat{W}_i(x)\Delta_{i+1}^{m+2^i}(x) & 0 \le i \le \log_2(h) \\ d_m & i = \log_2(h) \end{cases}$$

At each step, the polynomial is split into coefficients whose index has the i-th bit set and those which don't. The final steps select the coefficient corresponding to the selected index m.

Because of the properties of the basis polynomials, the vector of $\frac{h}{2^i}$ evaluations of Δ_i^m can be computed from two vectors of size $\frac{h}{2^{i+1}}$: the evaluations of Δ_{i+1}^m and $\Delta_{i+1}^{m+2^i}$ at points with the $i+1$ least significant bits unset.

Let $\Phi(i, m, l) = [\Delta_i^m(\omega_c + \omega_l)$ for c in $[0, 2^i, \ldots, h-2^i]]$ be the vector of $\frac{h}{2^i}$ evaluations of Δ_i^m at all points $\omega_c + \omega_l$ where c has the i most significant bits unset, with l an arbitrary offset.

$\Phi(i, m, l)$ can be computed in $O(n)$ time from $\Phi(i+1, m, l)$ and $\Phi(i+1, m+2^i, l)$.

For each pair of values at index x in the two smaller vectors, the values at indices $2x$ and $2x+2^i$ in the larger vector can be computed. The values will be denoted as a, b, a', b' for clarity.

a' is straightforwardly computed as:

$$a' = \Delta_i^m(\omega_c + \omega_l) = \Delta_{i+1}^m(\omega_c + \omega_l) + \hat{W}_i(\omega_c + \omega_l)\Delta_{i+1}^{m+2^i}(\omega_c + \omega_l) = a + \hat{W}_i(\omega_c + \omega_l)b$$

The calculation of b' relies on the properties of the vanishing polynomials:

$$b' = \Delta_i^m(\omega_c + \omega_l + \omega_{2^i}) = \Delta_{i+1}^m(\omega_c + \omega_l + \omega_{2^i}) + \hat{W}_i(\omega_c + \omega_l + \omega_{2^i})\Delta_{i+1}^{m+2^i}(\omega_c + \omega_l + \omega_{2^i})$$

The term ω_{2^i} vanishes in both Δ_{i+1}^m and $\Delta_{i+1}^{m+2^i}$, since both contain only vanishing polynomials W_j with $j \geq i + 1$.

As $\hat{W}_i$ is normalized, $\hat{W}_i(\omega_c + \omega_l + \omega_{2^i}) = \hat{W}_i(\omega_c + \omega_l) + \hat{W}_i(\omega_{2^i}) = \hat{W}_i(\omega_c + \omega_l) + 1$.

Therefore, b' is computed as:

$$b' = a + (\hat{W}_i(\omega_c + \omega_l) + 1)b = a + \hat{W}_i(\omega_c + \omega_l)b + b = a' + b$$

The reverse calculation is also straightforward, and does not require division:

$$b' + a' = (a' + b) + a' = b$$

$$a' + \hat{W}_i(\omega_c + \omega_l)b = (a + \hat{W}_i(\omega_c + \omega_l)b) + \hat{W}_i(\omega_c + \omega_l)b = a$$

The vectors can be stored interleaved in a single array, initialized to $[d_0, d_1, \ldots, d_{h-1}]$ (h single-element vectors), and then updated in-place in $log_2(h)$ steps, each step requiring $O(n)$ time.

See the butterfly diagram in [3] for a visual representation of the transforms.

In total, $n - 1$ unique factors are needed - one evaluation of $\hat{W}_{\log_2(n)}$, two of $\hat{W}_{\log_2(n)-1}$, $\ldots$, $\frac{n}{2}$ evaluations of $\hat{W}_0$ - which can be computed in $O(n \log n)$ time.

The inverse and forward transforms are almost identical, except the outer loop direction and the inner operations are reversed.

Notice the transforms can use factors of a greater power than needed. To compute multiple transforms of different sizes with the same offset, only the factors for the largest size must be computed, and can be used for all smaller sizes.

Algorithm 4 Transform Algorithms

function PRECOMPUTEFACTORS(pow, offset)
  factors $\leftarrow$ new array of GF(2^{64}) values of size $2^{\text{pow}} - 1$
  factor_idx $\leftarrow 0$
  for step $\leftarrow 0$ to pow $- 1$ **do**
    groups $\leftarrow 2^{\text{pow}-\text{step}-1}$
    for group $\leftarrow 0$ to groups $- 1$ **do**
      factors[factor_idx] $\leftarrow \hat{W}_{\text{step}}(\omega_{\text{group}\cdot 2^{\text{step}+1}} + \omega_{\text{offset}})$
      factor_idx $\leftarrow$ factor_idx $+ 1$
    end for
  end for
  return factors
end function

function INVERSETRANSFORM(data, factors)
  for step $\leftarrow 0$ to $\log_2(\text{len(data)}) - 1$ **do**
    group_len $\leftarrow 2^{\text{step}}$
    group_factors_start $\leftarrow$ len(factors) $+ 1 - \frac{\text{len(factors)}+1}{2^{\text{step}}}$
    for group $\leftarrow 0$ to $\frac{\text{len(data)}}{2^{\text{step}+1}} - 1$ **do**
      for x $\leftarrow 0$ to group_len $- 1$ **do**
        $a \leftarrow$ group $\cdot$ group_len $\cdot 2 + x$
        $b \leftarrow a +$ group_len
        data[b] $\leftarrow$ data[b] $+$ data[a]
        data[a] $\leftarrow$ data[a] $+$ data[b] $\cdot$ factors[group_factors_start $+$ group]
      end for
    end for
  end for
end function

function FORWARDTRANSFORM(data, factors)
  for step $\leftarrow \log_2(\text{len(data)}) - 1$ down to 0 **do**
    group_len $\leftarrow 2^{\text{step}}$
    group_factors_start $\leftarrow$ len(factors) $+ 1 - \frac{\text{len(factors)}+1}{2^{\text{step}}}$
    for group $\leftarrow 0$ to $\frac{\text{len(data)}}{2^{\text{step}+1}} - 1$ **do**
      for x $\leftarrow 0$ to group_len $- 1$ **do**
        $a \leftarrow$ group $\cdot$ group_len $\cdot 2 + x$
        $b \leftarrow a +$ group_len
        data[a] $\leftarrow$ data[a] $+$ data[b] $\cdot$ factors[group_factors_start $+$ group]
        data[b] $\leftarrow$ data[b] $+$ data[a]
      end for
    end for
  end for
end function

3.3 Formal Derivative

The error correction algorithm cannot directly use the inverse transform for interpolation, as the received data is not at contiguous points, and the number of non-corrupted points is likely not a power of two.

Instead, an algorithm based on the formal derivative is used which can recover the original polynomial from any k intact points regardless of error location.

In all fields, the formal derivative of a polynomial is well-defined and the standard power and product rules apply, despite the concepts of limits and continuity not existing in finite fields.

$$(f \cdot g)' = f' \cdot g + f \cdot g'$$
$$(\sum_{i=0}^{n} a_i x^i)' = \sum_{i=1}^{n} (i \cdot a_i) x^{i-1}$$

The multiplication $i \cdot a_i$ between an integer and a field element is defined as repeated addition, which in GF(2^n) is either zero or a_i, as $a_i + a_i = 0$.

$$i \cdot a_i = \begin{cases} a_i & i \text{ odd} \\ 0 & i \text{ even} \end{cases}$$

Therefore, the formal derivative of a polynomial f in GF(2^n) written in the standard monomial basis is:

$$f' = a_1 + a_3 x^2 + a_5 x^4 + \dots$$

As the normalized vanishing polynomial $\hat{W}_i$ only has coefficients at power-of-two indices, the derivative will be a constant:

$$\hat{W}_i' = \frac{\prod_{j=1}^{2^i - 1} \omega_j}{W_i(2^i)}$$

To find the derivative of the basis polynomial X_i, which is a product of up to 64 polynomials, the product rule generalized to a product of n polynomials is used:

$$(\prod_{i=0}^{n} f_i)' = \sum_{j=0}^{n} f_j' \cdot \prod_{i \neq j} f_j$$

Therefore, the derivative of X_i contains $|\text{bits}(i)|$ terms, each of which is a basis polynomial with one bit of i unset, multiplied by the derivative of the vanishing polynomial corresponding to that bit:

$$X'_i = \sum_{b \in \text{bits}(i)} \hat{W}'_b \cdot X_{i-2^b}$$

Notice that X'_i only has terms with indices less than i, so the derivative of a polynomial in basis $\mathbb{X}$ can be computed in-place by iterating from the lowest degree to the highest degree coefficients, in $O(n \log n)$ time.

Algorithm 5 Polynomial Derivative

```
function PRECOMPUTEDERIVATIVEFACTORS(pow)
    assert 0 ≤ pow ≤ 64
    factors ← new array of GF(2^64) values of size pow
    for l ← 1 to pow − 1 do
        for j ← 2^(l−1) to 2^l − 1 do
            factors[l] ← factors[l] ∗ ω_j
        end for
        if l + 1 ≠ pow then
            factors[l + 1] ← factors[l]
        end if
        factors[l] ← factors[l]/W_l(2^l)
    end for
    return factors
end function
function FORMALDERIVATIVE(data, factors)
    for i ← 0 to len(data) - 1 do
        for bit ← 0 to log_2(len(data)) do
            if i & 2^bit ≠ 0 then
                data[i − 2^bit] ← data[i − 2^bit] + data[i] · factors[bit]
            end if
        end for
        data[i] ← 0
    end for
end function
```

3.4 Polynomial Recovery

In order to recover the original polynomial using the formal derivative, an error locator polynomial is constructed which vanishes at the points where errors occurred.

Let ERASURES be the set of indices where an error occurred. As previously mentioned, erasure codes require knowledge of the location of all errors, which, for this application, will be obtained using hashing.

$$e = \prod_{i \in \text{ERASURES}} (x + \omega_i)$$

Since e does not depend on the actual values of the data polynomial, its values can be computed and multiplied with the received incomplete data polynomial d, to zero out all unknown values.

The product rule allows the original polynomial d to be recovered:

$$(e \cdot d)' = e' \cdot d + e \cdot d'$$

$$(e \cdot d)'(\omega_x) = e'(\omega_x) \cdot d(\omega_x) + 0 \cdot d'(\omega_x) \; \forall \; x \in \text{ERASURES}$$

$$d(\omega_x) = \frac{(e \cdot d)'(\omega_x)}{e'(\omega_x)} \; \forall \; x \in \text{ERASURES}$$

Therefore, the original polynomial is recovered by multiplying the received polynomial by the error locator polynomial, applying the inverse transform, taking the formal derivative, applying the forward transform, and finally dividing by the derivative of the error locator polynomial, at the error locations.

Algorithm 6 Reed-Solomon Decoding

t_fac $\leftarrow$ PrecomputeFactors$(\log_2(n), 0)$
d_fac $\leftarrow$ PrecomputeDerivativeFactors$(\log_2(n))$
d $\leftarrow [d_0, d_1, \ldots, d_{n-1}]$ ▷ received data
erasures $\leftarrow [i_0, i_1, \ldots, i_k]$ ▷ indices of errors
$(e, e') \leftarrow$ ComputeErrorLocator(erasures, t_fac, d_fac)
$\hat{d} \leftarrow d \cdot e$ ▷ multiply partially corrupt data by error locator polynomial
$\hat{d'} \leftarrow$ ForwardTransform(FormalDerivative(InverseTransform$(\hat{d}$, t_fac), d_fac), t_fac)
for $i \in$ erasures **do**
    $d[i] \leftarrow \hat{d'}[i]/e'[i]$
end for

For this application, (e, e') can be reused for the entire file, since all Reed-Solomon codes will have the same error locations - a missing block causes a missing value in all codes (remember the codes are 'columns' which span the entire file), and e' can be inverted in advance to reduce the number of multiplicative inverse operations.

3.5 Error Locator Computation

A $O(n \log n)$ algorithm for computing the error locator polynomial is described in [3] which uses fast Walsh-Hadamard transforms, however it requires 2^r operations where r is the power of the field, so it is not useful for GF(2^{64}).

Instead, I used a $O(n \log^2 n)$ recursive algorithm which splits the polynomial into two halves, and combines the two results by multiplying in $O(n \log n)$ time using the transforms.

Algorithm 7 Error Locator Polynomial Computation

function COMPUTEERRORLOCATOR(erasures, out_len, t_fac, d_fac)
  values ← new empty array
  coefficients ← InternalRecursion(erasures, out_len, t_fac, d_fac, values)
  FormalDerivative(coefficients, d_fac)
  ForwardTransform(coefficients, t_fac)
  return (values, coefficients) ▷ coefficients now contains values of derivative
end function
function INTERNALRECURSION(erasures, out_len, t_fac, out_values)
  if len(erasures) $= 1$ **then**
    if out_values $\neq$ null **then**
      out_values ← new array $[\omega_i + \omega_{\text{erasures[0]}}$ for i from 0 to out_len $- 1]$
    end if
    return new array $[\omega_{\text{erasures[0]}}, 1, 0, \ldots, 0]$ of size out_len
  end if
  special_case ← len(erasures) + 1 $=$ out_len
  $a \leftarrow$ InternalRecursion(erasures from 0 to $\frac{\text{len(erasures)}}{2} - 1$, $\frac{\text{out\_len}}{2}$, t_fac, null)
  ResizeWithZeros(a, out_len)
  ForwardTransform(a, t_fac)
  $b \leftarrow$ InternalRecursion(erasures from $\frac{\text{len(erasures)}}{2}$ + special_case to end, $\frac{\text{out\_len}}{2}$, t_fac, null)
  ResizeWithZeros(b, out_len)
  ForwardTransform(b, t_fac)
  $a \leftarrow a * b$ ▷ polynomial evaluations are multiplied in $O(n)$ time
  if special_case **then**
    $a \leftarrow a * [\omega_i + \omega_{\text{erasures}[\frac{\text{len(erasures)}}{2}]}$ for i from 0 to $\frac{\text{out\_len}}{2} - 1]$ ▷ multiply in extra value
  end if
  if out_values $\neq$ null **then** ▷ the top-most call must return both coefficients and values
    out_values ← Copy(a) ▷ the memory of b can be reused here for the copy
  end if
  InverseTransform(a, t_fac) ▷ convert back to coefficients after multiplications are done
  return a
end function

The special case is sometimes needed to prevent a branch where len(erasures) $=$ out_len, which would request only n coefficients for a polynomial of degree n.

Chapter 4

Implementation

The implementation is written in Rust, using some third-party libraries for I/O, multithreading, hashing, and progress reporting. No libraries were used for the finite field arithmetic, polynomial operations, or Reed-Solomon codes. See A.2 for a complete list of the libraries used.

4.1 Interleaved Codes

The data and parity files are not treated as one single Reed-Solomon codeword. Instead, the data is split into blocks. The blocks are *not* independent Reed-Solomon codes, but rather the codes are split across all blocks, with each block containing one symbol from each code. In other words, the files are treated as matrices, with the blocks as rows and codes as columns.

This is necessary for several reasons:

- Using one large Reed-Solomon code would require reading all data into memory, limiting the maximum size to the available memory, or requiring complicated saving and loading to disk as part of the processing.
- The precomputed factors for one large code would be as large as the code itself, requiring large amounts of time and space to compute and store.
- Since hashes are used to detect corruption, the file must be split into blocks anyways, limiting the granularity of error correction to the block size.

Any error occuring in a block will affect all codes, as deleting a row from a matrix affects one element of each column. Multiple errors in one block have the same effect as a single error. This is suited to common error patterns, which tend not to be uniformly distributed, but intead are burst errors.

See appendix A.1 for an visual example of how burst errors are easier to correct.

4.2 Data Storage

The parity data and metadata are stored in a separate file, specified by the user.

The metadata consists of the header, which specifies the parameters of the encoding - expected data file size, number of data and parity blocks, and block size - and hashes of all blocks, used to detect corruption. The hashes also include the first 8 bytes of each block, to allow reassembly if the blocks are somehow scrambled, such as by deletion or insertion of a byte. This is very unlikely to happen, but would cause complete failure without a way to put the blocks back in order.

As the Reed-Solomon codes are split across all blocks, reading and writing the data and parity files has a very inefficient access pattern, as reading one code requires one access to each block. The blocks are analogous to rows in a matrix, and the codes to columns. Processing a matrix stored in row-major order column-wise is inherently inefficient, since non-contiguous memory access is required.

The system call overhead and seek time can be somewhat mitigated by reading many symbols per block at once, as many as can fit into memory. This produces a large buffer of interleaved symbols, which can then be processed in memory.

Reducing the length of individual codes by increasing block size improves performance, since it allows reading more symbols at once, and therefore passing through the data file fewer times. However, there is still a penalty for the non-contiguous access, especially on a hard disk drive.

The worst case scenario is if there is not enough memory to read more than one code at once, since this will require one system call per symbol. In this case, the only option is to increase the block size, which reduces code length, allowing more symbols to be read at once.

This theoretically reduces the granularity of the error correction, causing a single-byte error to render large amounts of data useless for recovery, but since burst errors are most common, this is acceptable.

The performance could also be improved by splitting a file into small sections which fit into memory, but the sections would be independent and could not be used to repair each other.

Writing is implemented using memory mapped I/O, which is simpler to use, but relies completely on the operating system to batch writes to the disk.

Since the data symbols from each code are read in batches - many codes read from each block per pass through the data file - the writes will naturally be batched as well. Testing does not show a bottleneck in writing. If necessary, the batching could be done manually, collecting output symbols in a large buffer and using normal write calls, instead of memory mapped I/O.

The same I/O code is used for both encoding and decoding. When decoding, system calls are used to read uncorrupted symbols from both files, and memory mapped I/O is used to write recovered symbols to both files. The architecture could be extended to process an arbitrary number of files, such as multiple parity files, single-file archives combining data and parity, or arbitrary folder structures as data instead of a single file.

4.3 Multithreaded Processing

To process codes in parallel, a multithreaded pipeline is used, consisting of a reader thread, an adapter thread, multiple processor threads, and a writer thread.

The reader thread reads codes into an interleaved buffer as described in the previous section. The processor threads execute the encoding or decoding algorithm, writing the output symbols into a buffer which is sent to the writer thread. The writer thread simply copies symbols from received buffers into the output memory maps, allowing the operating system to flush pages to disk asynchronously.

In order to synchronize the threads, channels are used to send messages between them. Heap-allocated buffers are used to store input and output data, moving input data from reader to adapter to processor, and output data from processor to writer.

Used input buffers are returned back to the reader and output buffers returned to the processors using separate return channels.

Filled input buffers are sent by the reader to the adapter, which creates a task message for each code in the buffer and sends it to the processor threads through a shared channel, including an offset which specifies which code to read from the buffer, and a shared atomic counter which is decremented whenever a processor thread finishes processing a code, so that when every code has been processed, the input buffer is sent back to the reader to be reused.

There are five channels used in total for the following purposes:

- Sending filled input buffers from the reader to the adapter, along with the number of codes and the index of the first code.
- Sending task messages from the adapter to the processors, containing a reference to the input buffer, a shared atomic counter, the index of the code, offset into the buffer, and number of codes in the buffer.
- Sending filled output buffers from the processors to the writer, along with the index of the code.
- Returning input buffers to the reader after every code has been processed, which is done by decrementing the shared atomic counter and returning the buffer when it reaches zero.
- Returning output buffers to the processors after the output symbols have been copied to the memory maps by the writer.

The input buffers are protected by a read-write lock, which allows multiple processors to read codes from the buffer at once, but only one thread - the reader - can write to it at a time. This is only used to ensure thread-safety, not for synchronization, which is done only using channels and the atomic counters.

The processor threads share the same precomputed factors, which depending on the task are either transform factors at multiple offsets for encoding, or transform factors plus derivative factors and the error locator polynomial for decoding.

Chapter 5

Conclusions

The file metadata is currently not protected from corruption. This can be addressed by adding meta-parity blocks among the parity blocks. Although the metadata is generally small, and therefore unlikely to be corrupted, this is a significant flaw in the current implementation.

The implementation can successfully generate parity data and repair file corruption using Reed-Solomon codes in $O(n \log n)$ time.

Correctness is ensured using unit tests for the IO-free part of the implementation - the finite field arithmetic and polynomial algorithms - and end-to-end testing for the entire encoding and decoding sequence, using simulated file corruption.

Appendices

A.1 Recovery Figures

To visually demonstrate some limitations of this error correction scheme, the following figures show the use of Reed-Solomon codes to repair bitmap images.

Figure (a) has 12966 errors and can be repaired, yet figure (b) has only 3288 errors and cannot be repaired. This is because the errors in figure (a) are one contiguous burst, whereas the errors in figure (b) are many small bursts spread across the image, damaging more blocks than in figure (a).

(a) A recoverable bitmap image with one corrupt area.

(b) An unrecoverable bitmap image with many small corrupt areas.

(c) Figure (a), repaired.

Figure 1: Image source: `scipy.datasets.face` (derived from https://pixnio.com/fauna-animals/raccoons/raccoon-procyon-lotor)

A.1.1 Image Generation Script

```
from os import system
from scipy.datasets import face as get_face
from imageio.v3 import imwrite
from PIL import Image
import numpy as np

def main():
    face = get_face(gray=True)
    face = Image.fromarray(face).resize((256, 256))
    imwrite("face.bmp", np.array(face))
    system("cargo run encode face.bmp face.rsarc 4096 5")

    rng = np.random.default_rng(42)
    def randomize(arr):
        for i in range(0, len(arr)):
            arr[i] = rng.integers(0, 256)

    face = np.array(face)
    face_2 = face.copy()
    face_3 = face.copy()
    randomize(face_2.ravel()[7300:20300])

    for i in range(0, face_3.size, 2000):
        randomize(face_3.ravel()[i:i+100])

    print(f"Differences between face and face_2: {np.sum(face != face_2)}")
    print(f"Differences between face and face_3: {np.sum(face != face_3)}")

    imwrite("face_2.bmp", face_2)
    imwrite("face_2_repaired.bmp", face_2)
    imwrite("face_3.bmp", face_3)
    imwrite("face_3_repaired.bmp", face_3)

    system("cargo run repair face_2_repaired.bmp face.rsarc")
    system("cargo run repair face_3_repaired.bmp face.rsarc")

    face_2_repaired = Image.open("face_2_repaired.bmp")
    face_3_repaired = Image.open("face_3_repaired.bmp")
    assert np.all(np.array(face_2_repaired) == face)
    assert not np.all(np.array(face_3_repaired) == face)

    def to_png(path):
        imwrite(f"{path}.png", np.array(Image.open(f"{path}.bmp")))

    to_png("face_2")
    to_png("face_3")
    to_png("face_2_repaired")

if __name__ == "__main__":
    main()
```

A.2 Dependencies

The following Rust libraries were used in the implementation:

- `fastrand` - Random number generation for testing.
- `blake3` - Hashing for error detection.
- `crossbean-channel` - Channels for communication between threads. Although the standard library provides channels, multi-producer multi-consumer channels are not stabilized yet (as of Rust 1.87.0).
- `indicatif` - Terminal progress bars.
- `memmap2` - Cross-platform memory mapped I/O.
- `num_cpus` - Obtains of CPU cores for automatically selecting number of threads.
- `positioned-io` - Cross-platform random access file I/O.
- `sysinfo` - Obtains available memory for automatically selecting number of codes to process at once.

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