# Peroxide Guide

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# Prerequisites

- Rust
- Peroxide

# **Quick Start**

## 2.1 Cargo.toml

- To use peroxide, you should edit Cargo.toml
- Current document version is corresponding to 0.8 peroxide = "0.8"

## 2.2 Import all at once

• You can import all functions & structures at once

```
extern crate peroxide;
use peroxide::*;

fn main() {
    //Some codes...
}
```

# Vector

### 3.1 Print Vec<f64>

## 3.2 Syntactic sugar for Vec<f64>

- There is useful macro for Vec<f64>
- For R, there is c

```
\# R
a = c(1,2,3,4)
```

• For Peroxide, there is c!

```
// Rust
fn main() {
   let a = c!(1,2,3,4);
}
```

## 3.3 From ranges to Vector

• For R, there is seq to declare sequence.

```
# R
a = seq(1, 4, 1)
print(a)
# [1] 1 2 3 4
```

• For peroxide, there is seq to declare sequence.

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```
fn main() {
    let a = seq(1, 4, 1);
    a.print();
    // [1, 2, 3, 4]
```

## 3.4 Vector Operation

```
• There are some vector-wise operations
    - add(&self, other: Vec<f64>) -> Vec<f64>
    - sub(&self, other: Vec<f64>) -> Vec<f64>
    - mul(&self, other: Vec<f64>) -> Vec<f64>
    - div(&self, other: Vec<f64>) -> Vec<f64>
    - dot(\&self, other: Vec<f64>) -> f64
    - \text{norm(\&self)} \rightarrow f64
  fn main() {
      let a = c!(1,2,3,4);
      let b = c!(4,3,2,1);
      a.add(&b).print();
      a.sub(&b).print();
      a.mul(&b).print();
      a.div(&b).print();
      a.dot(&b).print();
      a.norm().print();
      // [5, 5, 5, 5]
      // [-3, -1, 1, 3]
      // [4, 6, 6, 4]
      // [0.25, 0.6667, 1.5, 4]
      // 20
      // 5.477225575051661 // sqrt(30)
• And there are some useful operations too.
    - pow(&self, usize) -> Vec<f64>
    - powf(&self, f64) -> Vec<f64>
    -  sqrt(&self) -> Vec<f64>
  fn main() {
      let a = c!(1,2,3,4);
      a.pow(2).print();
      a.powf(0.5).print();
```

#### Concatenation 3.5

There are two concatenation operations.

a.sqrt().print(); // [1, 4, 9, 16]

// [1, 1.4142, 1.7321, 2] // [1, 1.4142, 1.7321, 2] 3.5. CONCATENATION 11

```
• cat(T, Vec<T>) -> Vec<f64>
```

• concat(Vec<T>, Vec<T>) -> Vec<T>

```
fn main() {
    let a = c!(1,2,3,4);
    cat(0f64, a.clone()).print();
    // [0, 1, 2, 3, 4]

    let b = c!(5,6,7,8);
    concat(a, b).print();
    // [1, 2, 3, 4, 5, 6, 7, 8]
}
```

# Matrix

### 4.1 Declare matrix

- You can declare matrix by various ways.
  - R's way Default
  - MATLAB's way
  - Python's way
  - Other macro

### 4.1.1 R's way

- Description: Same as R matrix(Vector, Row, Col, Shape)
- Type: matrix(Vec<T>, usize, usize, Shape) where T: std::convert::Into<f64> + Copy
   Shape: Enum for matrix shape Row & Col

### 4.1.2 MATLAB's way

- Description: Similar to MATLAB (But should use &str)
- Type: ml\_matrix(&str)

```
fn main() {
    let a = ml_matrix("1 2; 3 4");
    a.print();
    //    c[0] c[1]
    // r[0]    1    2
```

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```
// r[1] 3 4
}
```

### 4.1.3 Python's way

• Description: Declare matrix as vector of vectors.

```
• Type: py_matrix(Vec<Vec<T>>) where T: std::convert::Into<f64> + Copy
```

```
fn main() {
    let a = py_matrix(vec![vec![1, 2], vec![3, 4]]);
    a.print();
    //    c[0] c[1]
    // r[0]    1    2
    // r[1]    3    4
}
```

#### 4.1.4 Other macro

- Description: R-like macro to declare matrix
- For R,

```
# R
a = matrix(1:4:1, 2, 2, Row)
print(a)
# [,1] [,2]
# [1,] 1 2
# [2,] 3 4
```

• For Peroxide,

### 4.2 Basic Method for Matrix

There are some useful methods for Matrix

- row(&self, index: usize) -> Vec<f64>: Extract specific row as Vec<f64>
- col(&self, index: usize) -> Vec<f64>: Extract specific column as Vec<f64>
- diag(&self) -> Vec<f64>: Extract diagonal components as Vec<f64>
- swap(&self, usize, usize, Shape) -> Matrix: Swap two rows or columns
- subs\_col(&mut self, usize, Vec<f64>): Substitute column with Vec<f64>
- subs\_row(&mut self, usize, Vec<f64>): Substitute row with Vec<f64>

```
fn main() {
   let a = ml_matrix("1 2; 3 4");
```

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```
a.row(0).print(); // [1, 2]
   a.col(0).print(); // [1, 3]
   a.diag().print(); // [1, 4]
   a.swap(0, 1, Row).print();
   // c[0] c[1]
   // r[0] 3 4
   // r[1]
           1
   let mut b = ml_matrix("1 2;3 4");
   b.subs_col(0, c!(5, 6));
   b.subs_row(1, c!(7, 8));
   b.print();
           c[0] c[1]
   // r[0]
           5 2
   // r[1] 7 8
}
```

### 4.3 Read & Write

In peroxide, we can write matrix to csv.

- write(&self, file\_path: &str): Write matrix to csv
- write\_with\_header(&self, file\_path, header: Vec<&str>): Write with header

Also, you can read matrix from csv.

- Type: read(&str, bool, char) -> Result<Matrix, Box<Error>>
- Description: read(file\_path, is\_header, delimiter)

### 4.4 Concatenation

There are two options to concatenate matrices.

- cbind: Concatenate two matrices by column direction.
- rbind: Concatenate two matrices by row direction.

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### 4.5 Matrix operations

• In peroxide, can use basic operations between matrices. I'll show you by examples.

```
fn main() {
   let a = matrix!(1;4;1, 2, 2, Row);
   (a.clone() + 1).print(); // -, *, / are also available
   // c[0] c[1]
   // r[0] 2 3
   // r[1]
             4
   let b = matrix!(5;8;1, 2, 2, Row);
   (a.clone() + b.clone()).print(); // -, *, / are also available
   // c[0] c[1]
   // r[0] 6 8
   // r[1] 10 12
   (a.clone() % b.clone()).print(); // Matrix multiplication
   // c[0] c[1]
   // r[0] 19 22
   // r[1] 43 50
```

## 4.6 Extract & modify components

- In peroxide, matrix data is saved as linear structure.
- But you can use two-dimensional index to extract or modify components.

```
fn main() {
    let mut a = matrix!(1;4;1, 2, 2, Row);
    a[(0,0)].print(); // 1
    a[(0,0)] = 2f64; // Modify component
    a.print();
    //    c[0] c[1]
    // r[0] 2 2
```

```
// r[1] 3 4
}
```

### 4.7 Conversion between vector

### 4.7.1 Vector to Matrix

• to\_matrix method allows conversion from vector to column matrix.

```
fn main() {
    let a = c!(1,2,3,4);
    a.to_matrix().print();
    //    c[0]
    // r[0]    1
    // r[1]    2
    // r[2]    3
    // r[3]    4
}
```

#### 4.7.2 Matrix to Vector

• Just use row or col method (I already showed at Basic method section).

```
fn main() {
    let a = matrix!(1;4;1, 2, 2, Row);
    a.row(0).print(); // [1, 2]
}
```

### 4.8 Useful constructor

- zeros(usize, usize): Construct matrix which elements are all zero
- eye(usize): Identity matrix
- rand(usize, usize): Construct random uniform matrix (from 0 to 1)

```
fn main() {
    let a = zeros(2, 2);
    assert_eq!(a, ml_matrix("0 0;0 0"));

let b = eye(2);
    assert_eq!(b, ml_matrix("1 0;0 1"));

let c = rand(2, 2);
    c.print(); // Random 2x2 matrix
}
```

# Linear Algebra

### 5.1 Transpose

• Caution: Transpose does not consume the original value.

```
fn main() {
    let a = matrix!(1;4;1, 2, 2, Row);
    a.transpose().print();
    // Or you can use shorter one
    a.t().print();
    // c[0] c[1]
    // r[0] 1 3
    // r[1] 2 4
}
```

## 5.2 LU Decomposition

- Peroxide uses complete pivoting for LU decomposition Very stable
- Since there are lots of causes to generate error, you should use Option
- lu returns Option<PQLU>

```
PQLU has four field - p, q, 1, u
p means row permutations
q means column permutations
1 means lower triangular matrix
u menas upper triangular matrix
```

• The structure of PQLU is as follows:

```
#[derive(Debug, Clone)]
pub struct PQLU {
    pub p: Perms,
    pub q: Perms,
    pub l: Matrix,
    pub u: Matrix,
}

pub type Perms = Vec<(usize, usize)>;
```

• Example of LU decomposition:

### 5.3 Determinant

• Peroxide uses LU decomposition to obtain determinant  $(\mathcal{O}(n^3))$ 

```
fn main() {
    let a = matrix!(1;4;1, 2, 2, Row);
    assert_eq!(a.det(), -2f64);
}
```

### 5.4 Inverse matrix

- Peroxide uses LU decomposition to obtain inverse matrix.
- It needs two sub functions inv 1, inv u
  - For inverse of L, U, I use block partitioning. For example, for lower triangular matrix:

$$L = \begin{pmatrix} L_1 & \mathbf{0} \\ L_2 & L_3 \end{pmatrix}$$

$$L^{-1} = \begin{pmatrix} L_1^{-1} & \mathbf{0} \\ -L_3^{-1}L_2L_1^{-1} & L_3^{-1} \end{pmatrix}$$

```
fn main() {
    let a = matrix!(1;4;1, 2, 2, Row);
    a.inv().unwrap().print();
    //    c[0] c[1]
    // r[0]   -2    1
    // r[1]   1.5 -0.5
}
```

### 5.5 Moore-Penrose Pseudo Inverse

```
• X^{\dagger} = (X^T X)^{-1} X

fn main() {

let a = matrix!(1;4;1, 2, 2, Row);

let pinv_a = a.psudo_inv().unwrap();
```

# **Functional Programming**

### 6.1 FP for Vector

There are some functional programming tools for Vec<f64>

```
pub trait FPVector {
    type Scalar;
   fn fmap<F>(&self, f: F) -> Self
    where
        F: Fn(Self::Scalar) -> Self::Scalar;
   fn reduce<F, T>(&self, init: T, f: F) -> Self::Scalar
    where
        F: Fn(Self::Scalar, Self::Scalar) -> Self::Scalar,
        T: convert::Into<Self::Scalar>;
   fn zip_with<F>(&self, f: F, other: &Self) -> Self
   where
        F: Fn(Self::Scalar, Self::Scalar) -> Self::Scalar;
   fn filter<F>(&self, f: F) -> Self
    where
        F: Fn(Self::Scalar) -> bool;
   fn take(&self, n: usize) -> Self;
   fn skip(&self, n: usize) -> Self;
```

### 6.1.1 fmap

• fmap is syntactic sugar for map

```
fn main() {
    let a = c!(1,2,3,4);

    // Original rust
    a.clone()
        .into_iter()
        .map(|x| x + 1f64)
        .collect::<Vec<f64>>()
        .print();
        // [2, 3, 4, 5]
```

```
// fmap in Peroxide
a.fmap(|x| x + 1f64).print();
// [2, 3, 4, 5]
}
```

#### 6.1.2 reduce

• reduce is syntactic sugar for fold

```
fn main() {
    let a = c!(1,2,3,4);

    // Original rust
    a.clone()
        .into_iter()
        .fold(0f64, |x, y| x + y)
        .print(); // 10

    // reduce in Peroxide
    a.reduce(0f64, |x, y| x + y).print(); // 10
}
```

### 6.1.3 zip\_with

• zip\_with is composed of zip & map

```
fn main() {
    let a = c!(1,2,3,4);
    let b = c!(5,6,7,8);

// Original rust
a.clone()
    .into_iter()
    .zip(&b)
    .map(|(x, y)| x + *y)
    .collect::<Vec<f64>>().print();
    // [6, 8, 10, 12]

// zip_with in Peroxide
a.zip_with(|x, y| x + y, &b).print();
    // [6, 8, 10, 12]
}
```

### 6.1.4 filter

• filter is just syntactic sugar for filter

```
fn main() {
    let a = c!(1,2,3,4);
    a.filter(|x| x > 2f64).print();
    // [3, 4]
}
```

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### 6.1.5 take & skip

• take is syntactic sugar for take

```
fn main() {
    let a = c!(1,2,3,4);
    a.take(2).print();
    // [1, 2]
}
```

• skip is syntactic sugar for skip

```
fn main() {
    let a = c!(1,2,3,4);
    a.skip(2).print();
    // [3, 4]
}
```

### 6.2 FP for Matrix

• Similar to FPVector

• Above functions play same roles as FPVector

## **Statistics**

### 7.1 Statistics trait

```
- To make generic code, there is {\tt Statistics} trait
```

```
- mean: just mean
- var: variance
- sd: standard deviation (R-like notation)
- cov: covariance
- cor: correlation coefficient

pub trait Statistics {
    type Array;
    type Value;

    fn mean(&self) -> Self::Value;
    fn var(&self) -> Self::Value;
    fn sd(&self) -> Self::Value;
    fn cov(&self) -> Self::Array;
    fn cor(&self) -> Self::Array;
}
```

### 7.1.1 For Vec<f64>

• Caution: For Vec<f64>, cov & cor are unimplemented (those for Matrix)

```
fn main() {
    let a = c!(1,2,3,4,5);
    a.mean().print(); // 3
    a.var().print(); // 2.5
    a.sd().print(); // 1.5811388300841898
}
```

• But there are other functions to calculate cov & cor

}

### 7.1.2 For Matrix

- For Matrix, mean, var, sd means column operations
- cov means covariance matrix & cor means also correlation coefficient matrix

```
fn main() {
   let m = matrix(c!(1,2,3,3,2,1), 3, 2, Col);
   m.mean().print(); // [2, 2]
   m.var().print(); // [1.0000, 1.0000]
   m.sd().print(); // [1.0000, 1.0000]
   m.cov().print();
       c[0]
                     c[1]
   // r[0] 1.0000 -1.0000
   // r[1] -1.0000 1.0000
   m.cor().print();
        c[0]
                     c[1]
   // r[0] 1 -1.0000
   // r[1] -1.0000
}
```

### 7.2 Simple Random Number Generator

• Peroxide uses external rand crate to generate random number

```
extern crate rand;
use self::rand::prelude::*;

fn main() {
    let mut rng = thread_rng();

    let a = rng.gen_range(0f64, 1f64); // Generate random f64 number ranges from 0 to 1
}
```

• To want more detailed explanation, see rand crate

## 7.3 Probability Distribution

- There are some famous pdf in Peroxide (not checked pdfs will be implemented soon)
  - ⊠ Bernoulli
  - ⊠ Beta
  - $\hfill\Box$  Dirichlet
  - ⊠ Gamma
  - ⊠ Normal
  - ☐ Student's t
  - $\boxtimes$  Uniform
  - □ Wishart
- There are two enums to represent probability distribution
  - OPDist<T>: One parameter distribution (Bernoulli)

- TPDist<T>: Two parameter distribution (Uniform, Normal, Beta, Gamma)
  \* T: PartialOrd + SampleUniform + Copy + Into<f64>
- There are some traits for pdf
  - RNG trait extract sample & calculate pdf
  - Statistics trait already shown above

#### 7.3.1 RNG trait

- RNG trait is composed of two fields
  - sample: Extract samples
  - pdf: Calculate pdf value at specific point

```
pub trait RNG {
    /// Extract samples of distributions
    fn sample(&self, n: usize) -> Vec<f64>;

    /// Probability Distribution Function
    ///
    /// # Type
    /// `f64 -> f64`
    fn pdf<S: PartialOrd + SampleUniform + Copy + Into<f64>>(&self, x: S) -> f64;
}
```

#### 7.3.2 Bernoulli Distribution

• Definition

Bern
$$(x|\mu) = \mu^x (1-\mu)^{1-x}$$

- Representative value
  - Mean:  $\mu$ - Var:  $\mu(1-\mu)$
- In peroxide, to generate  $Bern(x|\mu)$ , use simple algorithm
  - 1. Generate  $U \sim \text{Unif}(0,1)$
  - 2. If  $U \leq \mu$ , then X = 1 else X = 0
- Usage is very simple

```
fn main() {
    let b = Bernoulli(0.1); // Bern(x | 0.1)
    b.sample(100).print(); // Generate 100 samples
    b.pdf(0).print(); // 0.9
    b.mean().print(); // 0.1
    b.var().print(); // 0.09 (approximately)
    b.sd().print(); // 0.3 (approximately)
}
```

### 7.3.3 Uniform Distribution

• Definition

Unif
$$(x|a,b) = \begin{cases} \frac{1}{b-a} & x \in [a,b] \\ 0 & \text{otherwise} \end{cases}$$

• Representative value

```
– Mean: \frac{a+b}{2}
– Var : \frac{1}{12}(b-a)^2
```

• To generate uniform random number, Peroxide uses rand crate

```
fn main() {
    // Uniform(start, end)
    let a = Uniform(0, 1);
    a.sample(100).print();
    a.pdf(0.2).print();
    a.mean().print();
    a.var().print();
    a.sd().print();
}
```

#### 7.3.4 Normal Distribution

• Definition

$$\mathcal{N}(x|\mu,\sigma^2) = \frac{1}{\sqrt{2\pi\sigma^2}} \exp\left(-\frac{(x-\mu)^2}{2\sigma^2}\right)$$

- Representative value
  - Mean:  $\mu$
  - Var:  $\sigma^2$
- To generate normal random number, there are two famous algorithms
  - Marsaglia-Polar method
  - Ziggurat algorithm
- In peroxide, main algorithm is Ziggurat most efficient algorithm to generate random normal samples.
  - Code is based on a C implementation by Jochen Voss.

```
fn main() {
    // Normal(mean, std)
    let a = Normal(0, 1); // Standard normal
    a.sample(100).print();
    a.pdf(0).print(); // Maximum probability
    a.mean().print();
    a.var().print();
    a.sd().print();
}
```

#### 7.3.5 Beta Distribution

### 7.3.6 Gamma Distribution