

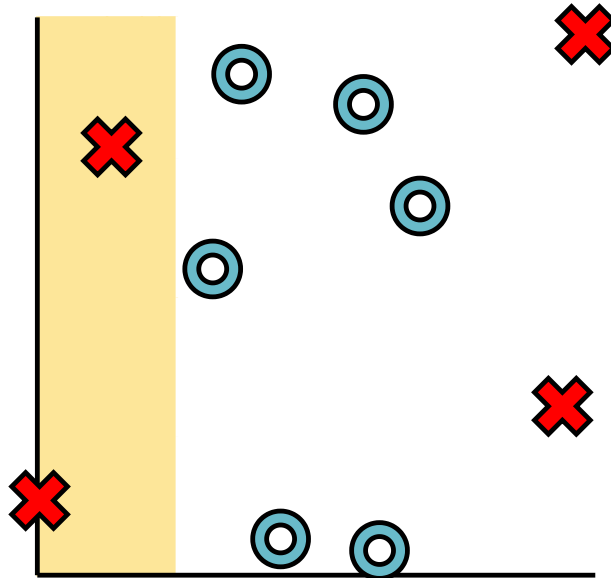


# Module 2.1 - Tensors



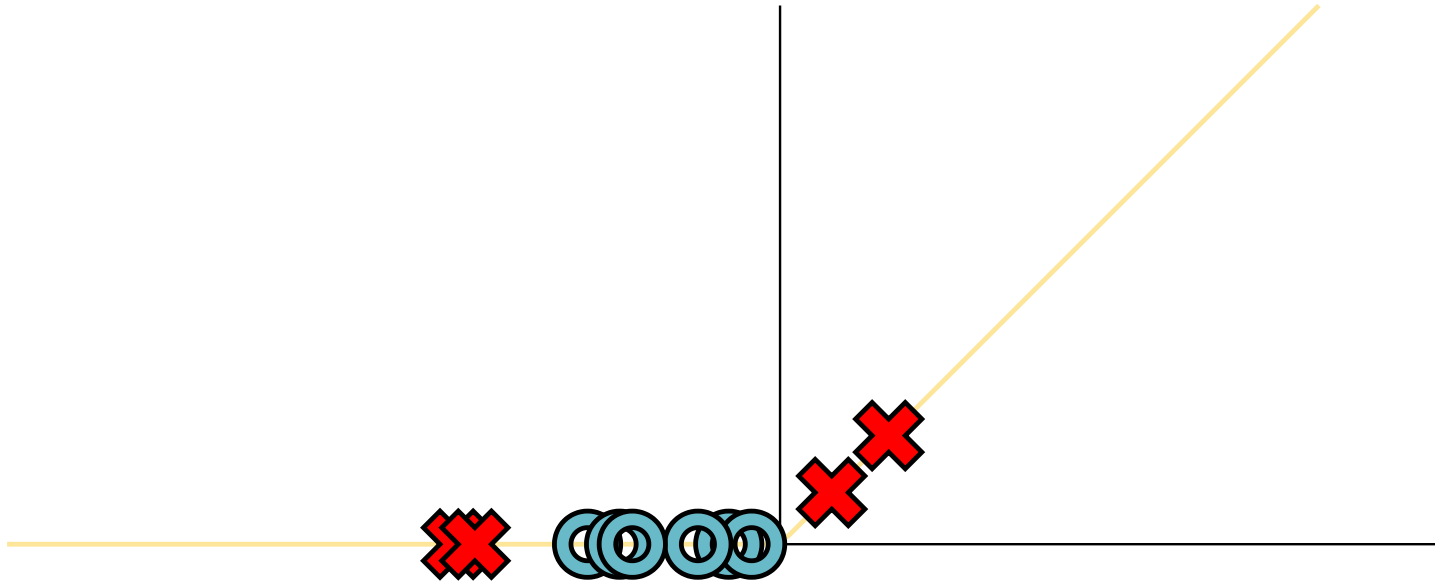
# Intuition: Split 1

```
yellow = Linear(-1, 0, 0.25)
```





# Reshape: ReLU





# Math View

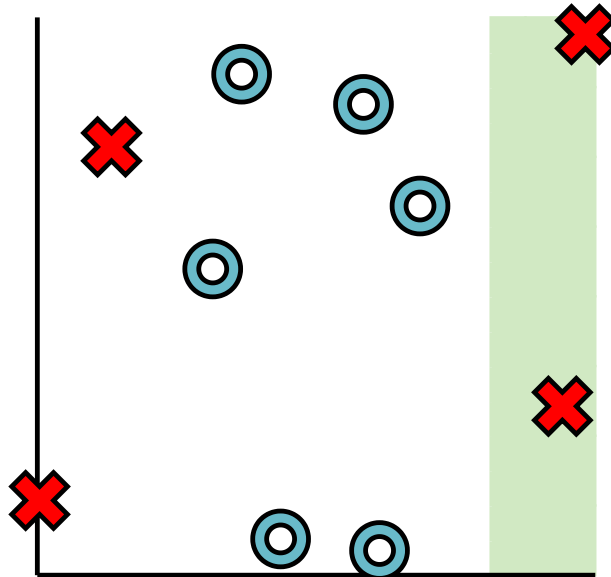
$$h_1 = \text{ReLU}(\text{lin}(x; w^0, b^0))$$





# Intuition: Split 2

```
green = Linear(1, 0, -0.8)
```



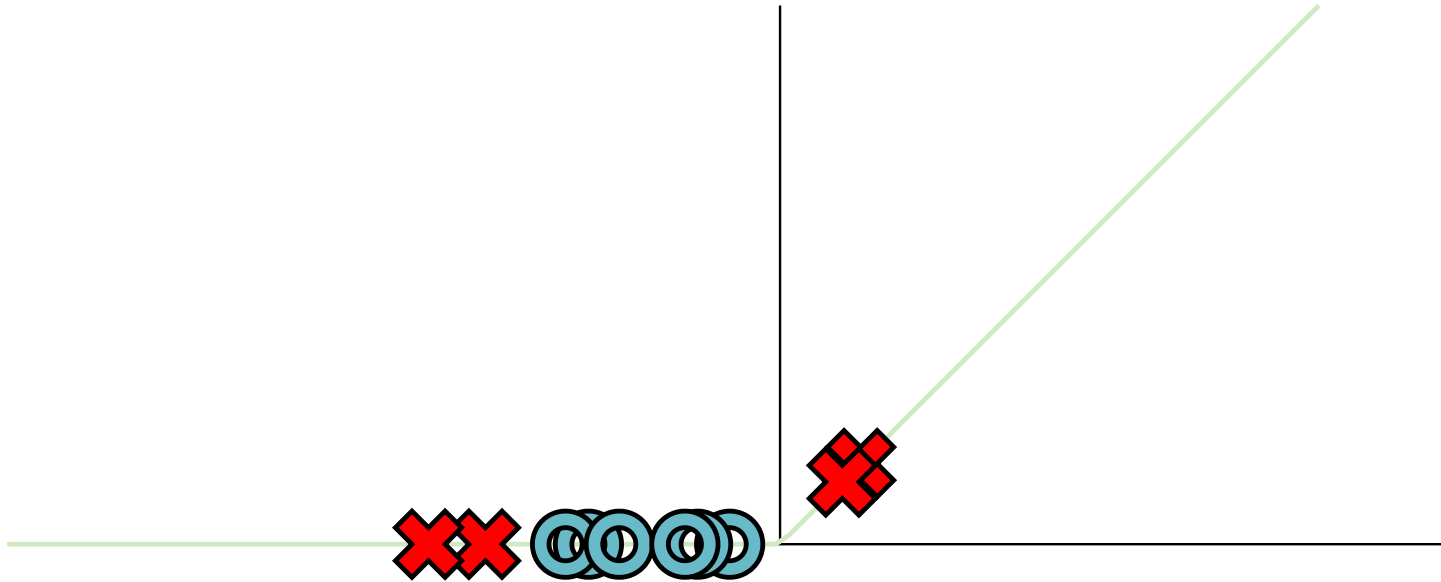


# Math View

$$h_2 = \text{ReLU}(\text{lin}(x; w^1, b^1))$$

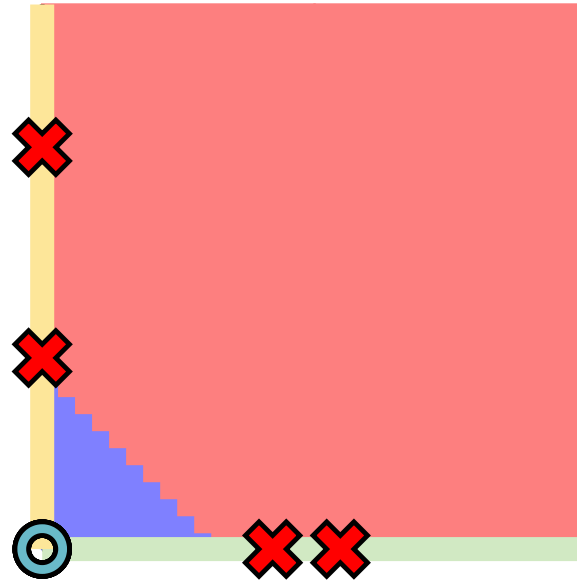


# Reshape: ReLU





# Reshape: ReLU







# Math View (Alt)

$$\text{lin}(x; w, b) = x_1 \times w_1 + x_2 \times w_2 + b$$

$$h_1 = \text{ReLU}(\text{lin}(x; w^0, b^0))$$

$$h_2 = \text{ReLU}(\text{lin}(x; w^1, b^1))$$

$$m(x_1, x_2) = \text{lin}(h; w, b)$$

Parameters:  $w_1, w_2, w_1^0, w_2^0, w_1^1, w_2^1, b, b^0, b^1$



# Quiz



# Outline

- Tensors
- Operations
- Strides



# Tensors





# Motivation

$$\text{lin}(x; w, b) = x_1 \times w_1 + x_2 \times w_2 + b$$

$$h_1 = \text{ReLU}(\text{lin}(x; w^0, b^0))$$

$$h_2 = \text{ReLU}(\text{lin}(x; w^1, b^1))$$

$$m(x_1, x_2) = \text{lin}(h; w, b)$$

Parameters:  $w_1, w_2, w_1^0, w_2^0, w_1^1, w_2^1, b, b^0, b^1$

- This is really messy!



# Matrix Form

$$\mathbf{h} = \text{ReLU}(\mathbf{W}^0 \mathbf{x} + \mathbf{b}^0)$$
$$m(\mathbf{x}) = \mathbf{W} \mathbf{h} + \mathbf{b}$$

Parameters:  $\mathbf{W}, \mathbf{b}, \mathbf{W}^{(0)}, \mathbf{b}^{(0)}$

- Need to be careful thought



# Tensors

What is it?

- Fancy multi-dimensional arrays
- Basis for an mathematical programming



# Tensors!

## Key Building Block

- Matlab
- Numpy
- Tensorflow, etc.





# Terminology

- 0-Dimensional Scalar
- `Scalar` from module-0



# Terminology

- 1-Dimensional
- Math: Vector





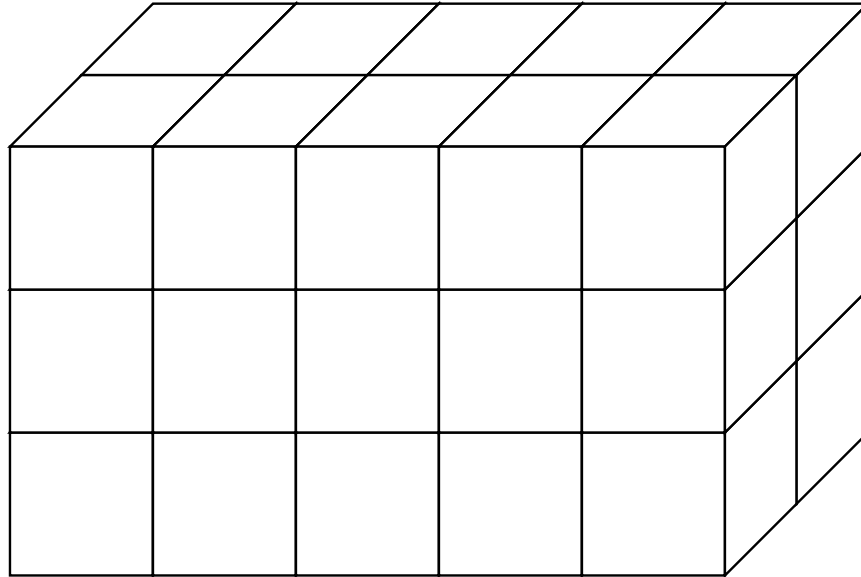
# Terminology

- 2-Dimensional
- Math: Matrix




# Terminology

- Arbitrary dimensions - Tensor (Array in numpy)







# Terminology

- Dims - # dimensions (`x.dims`)
- Shape - # cells per dimension (`x.shape`)
- Size - # cells (`x.size`)



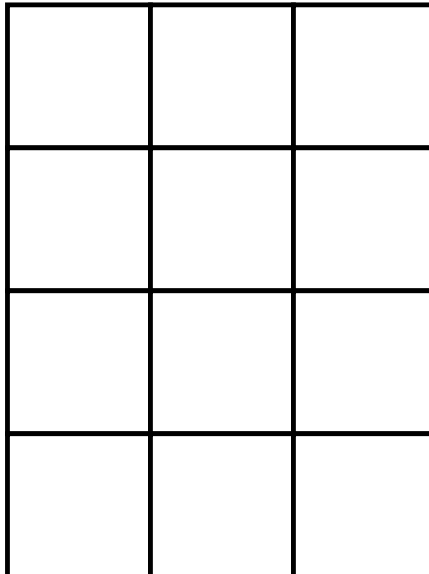
# Example

- dims: 2
- shape: (3, 5)
- size : 15




# Example

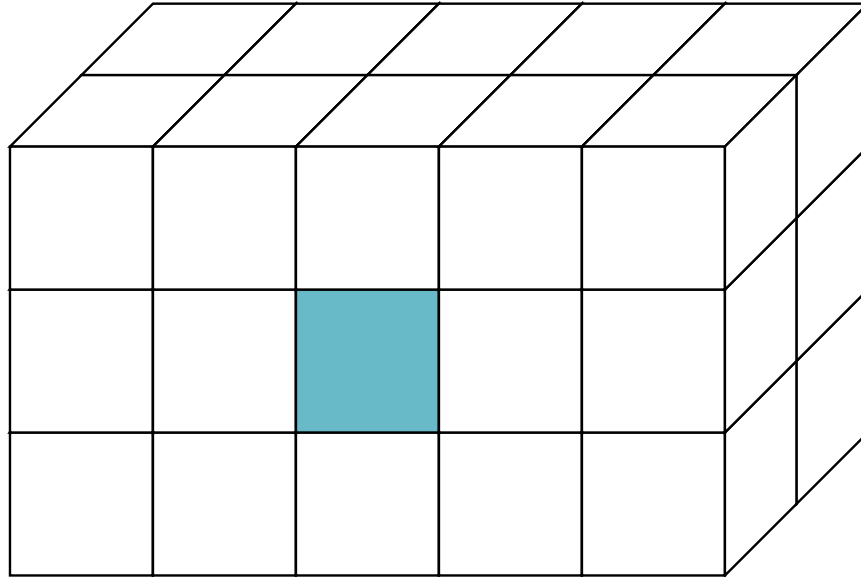
- dims: ?
- shape: ?
- size : ?





# Indexing

- Indexing syntax: `tensor[0, 1, 2]`







# Convention

- depth
- row
- columns



# Operations



# Goal

- Support user api
- Keep track of tensor properties
- Setup fast / simple Functions



# Tensor Usage

## Unary

```
new_tensor = x.log()
```

## Binary (for now, only same shape)

```
new_tensor = x + x
```

## Reductions

```
new_tensor = x.sum()
```

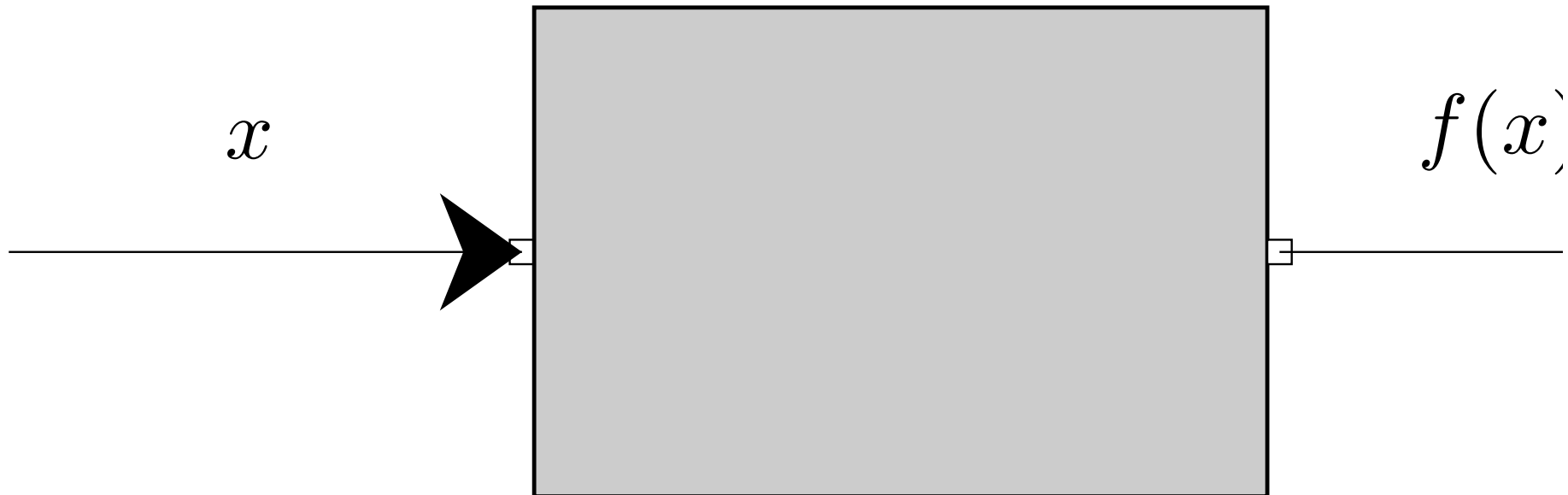




# Idea 1: Immutable Operations

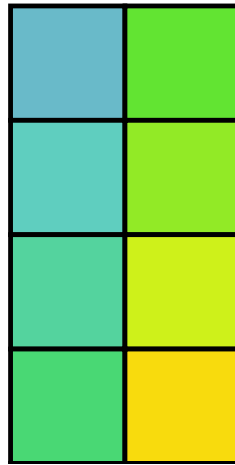
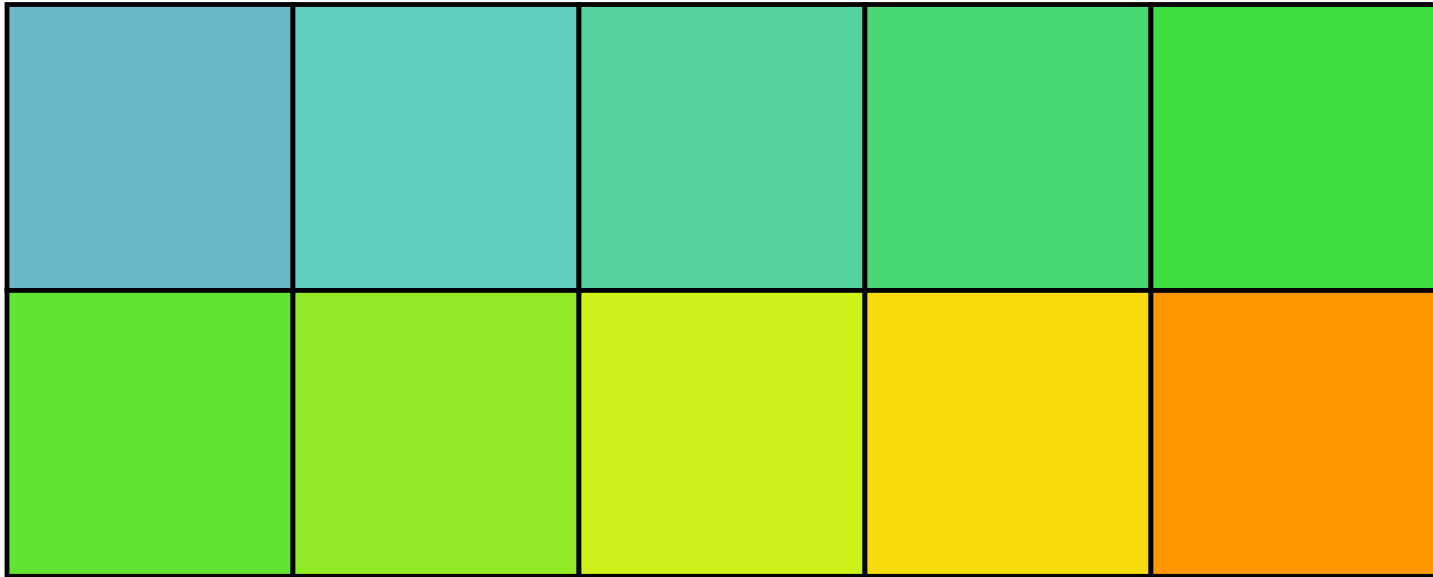
- Minitorch doesn't let you update tensors
- All operations return a new tensor (just like scalar)

```
draw_boxes(["$x$", "$f(x)$"], [1])
```



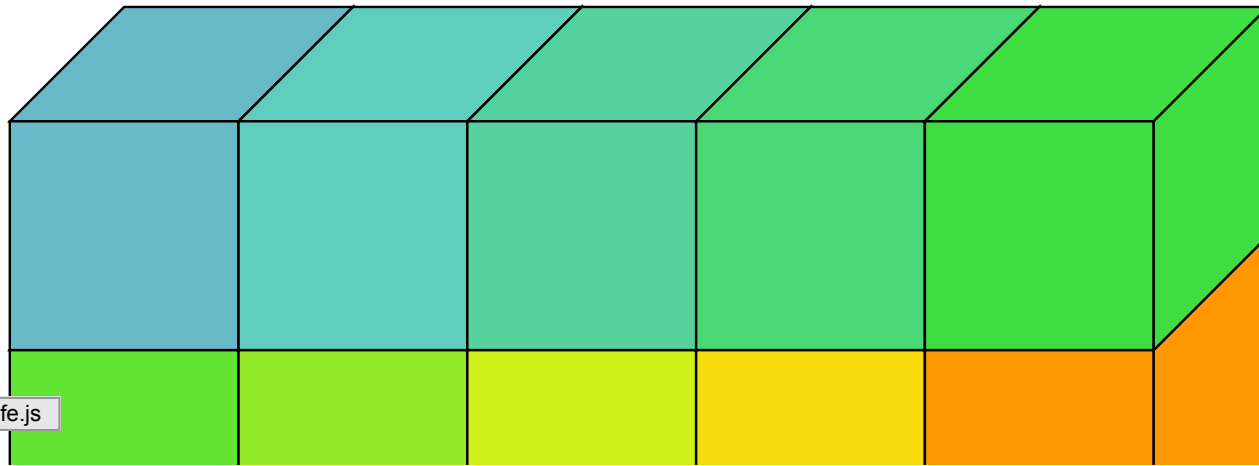
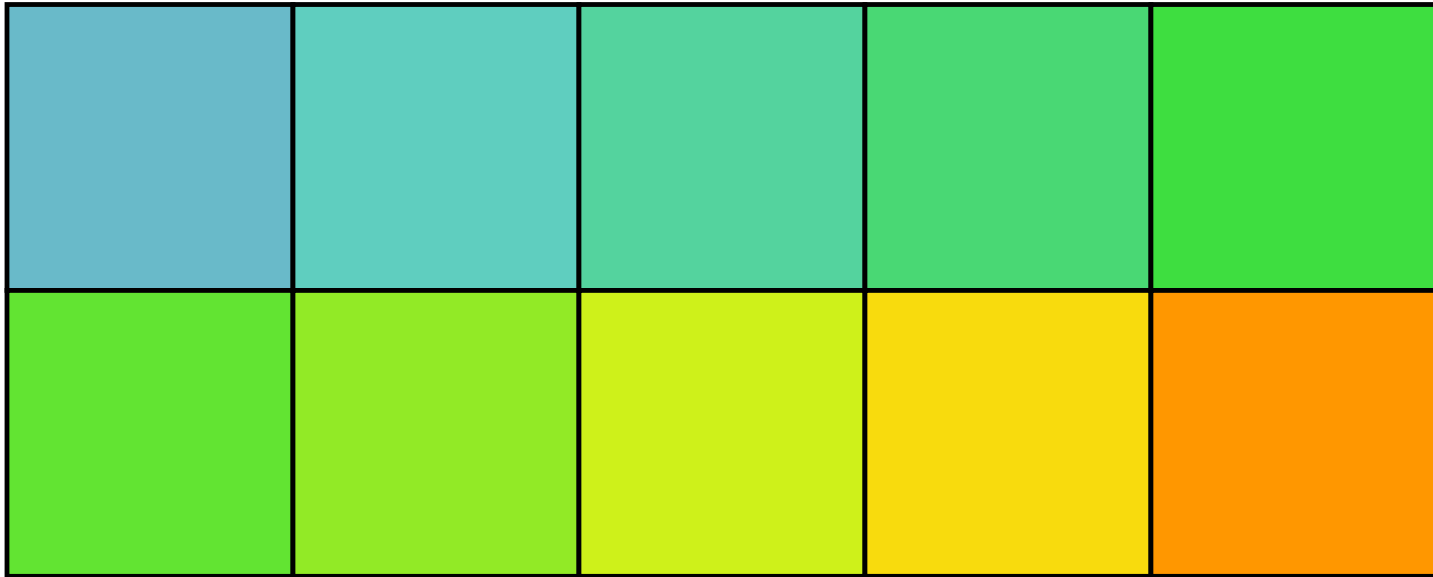


# Shape Manipulation: Permutation





# Shape Manipulation: View





# Why not just use lists?

- Functions to manipulate shape
- Mathematical notation
- Can act as Variables / Parameters
- Efficient control of memory (Module-3)





# Why not lists?

## Matrix (5, 2)

```
ls = [[1, 2], [3, 4], [5, 7], [2, 3], [2, 4]]
```

## View (1, 5, 2)

```
new_ls = [[ls[j][i] for i in range(2)] for j in range(5)]
```

## Transpose (2, 5)

```
matrix_trans = [[ls[i][j] for i in range(5)] for j in range(2)]
```



# Issues

- Operators requires copying

```
matrix_trans = [[ls[i][j] for i in range(5)] for j in range(2)]
```

- Storage shaped based on usage

```
new_tensor = [[[ls[j][i] for i in range(2)] for j in range(5)]]
```



## Idea 2: Views

- Separate storing information from user view
- Keep a mapping from users version to storage



# What's bad about tensors?

- Hard to grow or shrink
- Only numerical values
- Lose comprehensions / python built-ins
- Shapes are easy to mess up





# Next Couple Lectures

- No autodifferentiation for now
- Only consider forward tensor operations
- Add autodiff afterwards



# Tensor Internals

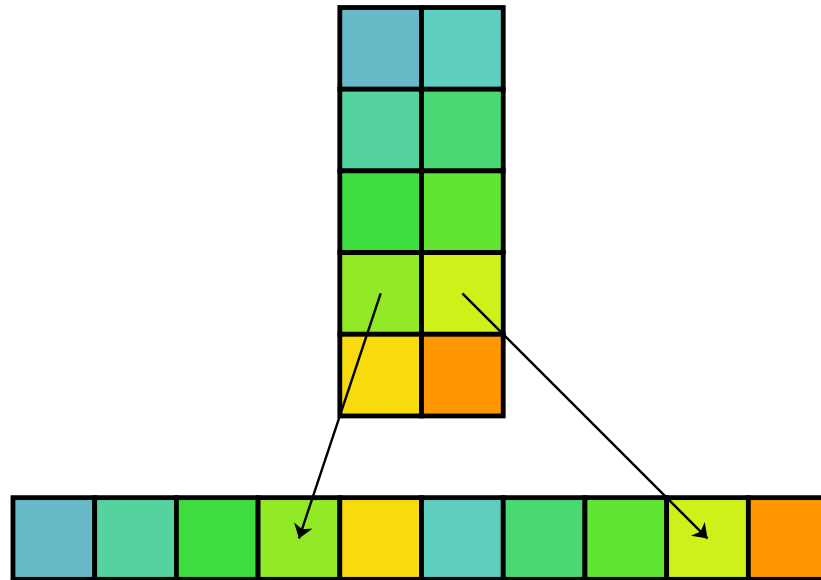


# How does this work

- **Storage** : 1-D array of numbers of length `size`
- **Strides** : tuple that provides the mapping from user `indexing` to the `position` in the 1-D `storage`.



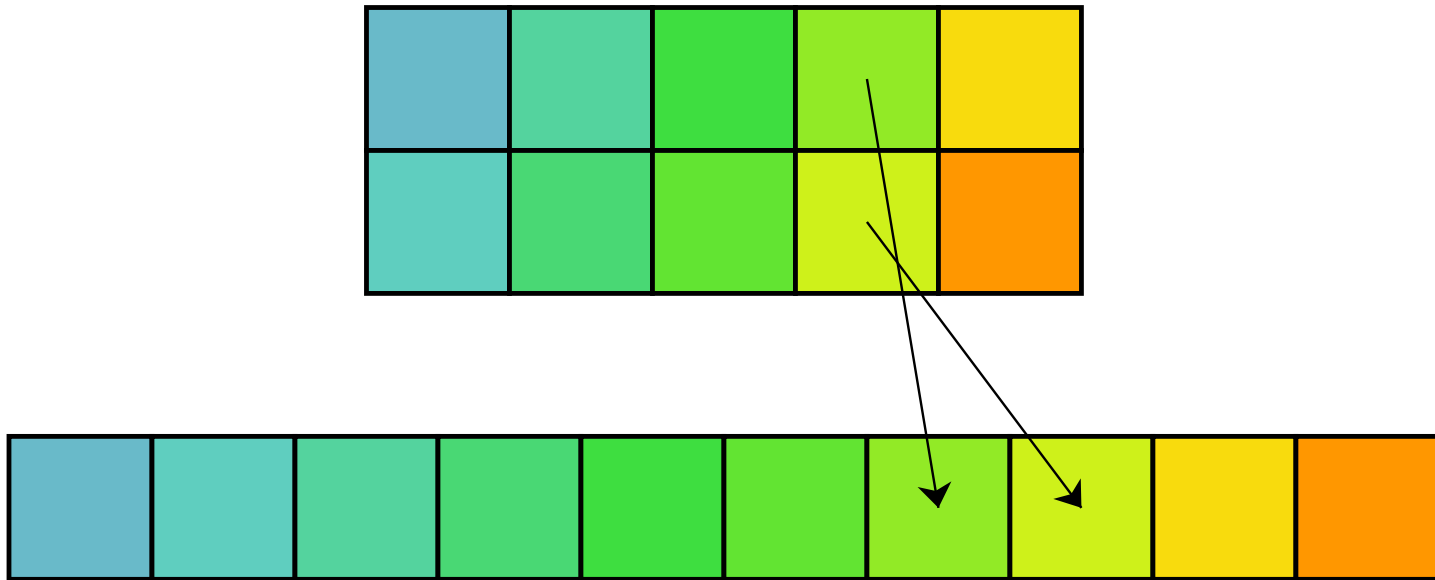
# Strides





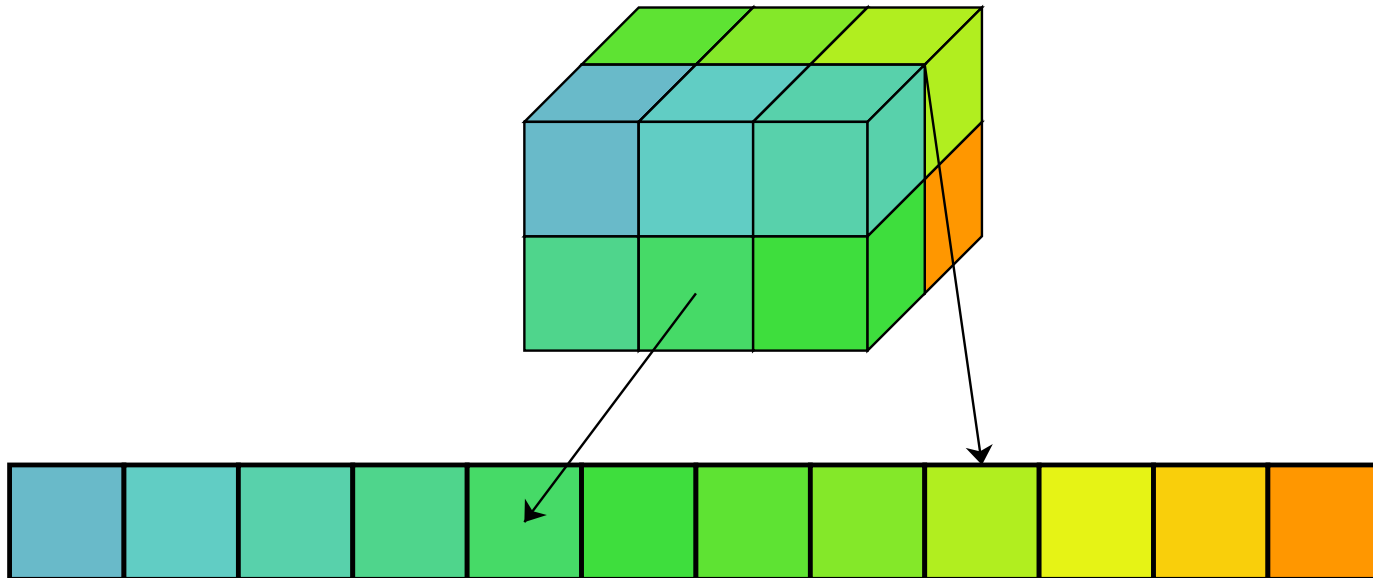


# Strides





# Strides





# Which is best?

- Can be useful when it is contiguous
- Our Convention: Bigger strides left
- $(s_1, s_2, s_3)$



# Operation 1: Indexing

- $v[i, j, k]$

How to find data point?





# Operation 2: Movement

How do I move to the next in the row? Column?



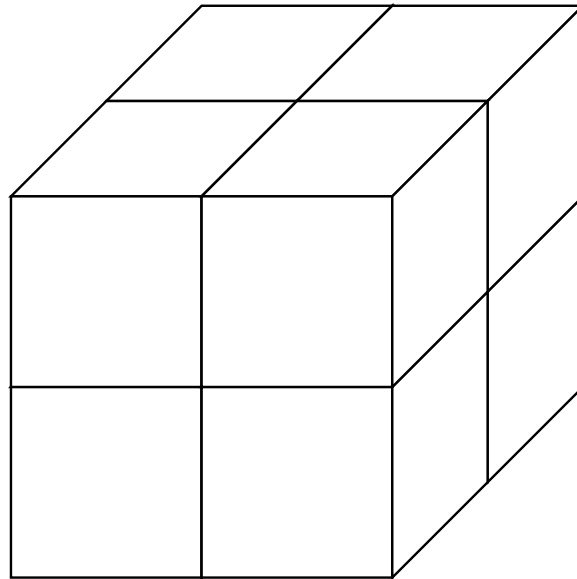
# Operation 3: Reverse Indexing

How do I find the index for data?



# Stride Intuition

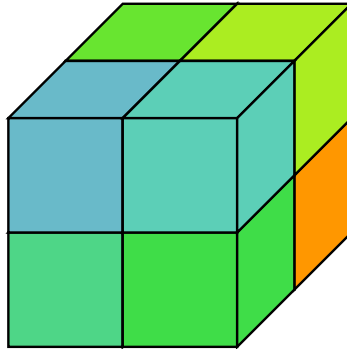
- Numerical bases,
- Index for position 0? Position 1? Position 2?





# Stride Intuition

- Index for position 0? Position 1? Position 2?
- $[0, 0, 0]$ ,  $[0, 0, 1]$ ,  $[0, 1, 0]$







# Conversion Formula

- Divide and mod
- $k = p$
- $j = (p // s_2)$
- ...



# Key Operations

- Map from index to position
- Map from position to index



# Implementation

- `TensorData` : Manager of strides and storage



# Module-2





# Overview

- `tensor.py` - Tensor Variable
- `tensor_functions.py` - Tensor Functions
- `tensor_data.py` - Storage and Indexing
- `tensor_ops.py` - Low-level tensor operations



# Q&A

