# C?

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# Overview

- Background
- Implementation
  - Syntax
  - Program Structure
  - Features
- Libraries
  - Math
  - DEEP
- Demo

## **Design Goals**

- Languages are made or broken by their libraries
  - Python: Numpy, Pandas, Theano, Tensorflow
  - Ruby: Rails
  - Prolog: ...?

- What does a library need?
  - Easy to use, hard to break: strong typing, yet familiar syntax
  - Custom types for extensibility: structs
  - Abstracting calls from definitions: function pointers
  - Heavy data crunching: matrices
  - Links to other languages with better libraries

## Implementation: Syntax

Basically C

- {} for scoping
- Lines end with ;
- Variables declared as typ NAME
- Requires int main() {} as execution entry point

There's some Go. Andrew wanted it.

```
int main() {
    int i;
    i = 0;
    while ( i < 10 ) {
        i = i + 1;
        print( i );
    7
    for (i = 0; i < 10; i = i + 1) {
        print( i );
    }
    return 0;
```

## Implementation: Program Structure

- Statically Scoped
- Declarations for structs/functions/variables must come before use
- Standard Control Flow
  - If...else...
  - While, For
  - Return
- Didn't stray from MicroC was not our area of interest

# Features: Arrays

- Every array has 8 bytes overhead
  - Total size in bytes
  - Length
- Array literals
- Dynamic array resizing
- Concatenation and Append

```
int[] a;
int[] b;
a = (int[]) {1,2,3};
b = (int[]) \{4,5,6\};
a = concat(a,b); // {1,2,3,4,5,6}
a = append(a,7); // {1,2,3,4,5,6,7}
print(len(a)); // 7
```

## Features: Structs

- Arbitrary collection of custom types
  - Nested structs
  - Arrays
- Method Dispatch
- Allocated on Heap, pass by reference

```
struct rectangle {
  float: width, length;
}
[struct rectangle r] area() float {
  return r.width * r.length;
int main() {
  struct rectangle rectangle;
  rectangle = make(struct rectangle); // malloc space
  rectangle.width = 3.0;
  rectangle.length = 4.0;
  print_float(rectangle.area());
                                       // 12.0
```

#### **Features: Function Pointers**

- Abstract function calling from function definition
- Allow for creation of modular plug and play components

```
int add(int x, int y) {
   return x + y;
int mult(int x, int y) {
   return x * y;
/* In the function pointer type below, the last value type is the return */
void print_bin(fp (int, int, int) f, int x, int y) {
   print(f(x, y));
    return;
int main() {
                                 /* 42 */
   print_bin(add, 7, 35);
                                 /* 42 */
   print_bin(mult, 7, 6);
   return 0;
```

## Features: C Links

- Link to any C code with extern keyword
- Provide C code in /lib/ folder
- Compiler combines C LLVM with generated LLVM for single executable

```
extern void printbig(int c);
int main() {
    printbig(72); /* H */
    return 0;
}
```

#### **Features: Matrices**

- Matrix implementation through eigen library
- Large number of eigen operators available, built-in

```
int main(){
    fmatrix fm1;
    fmatrix fm2;
    fmatrix fm3;
```

```
/* Create a 5 by 5 matrix of zeros */
fm1 = init_fmat_zero(5, 5);
/* Create a 5 by 5 matrix of 2.5's */
fm2 = init_fmat_const(2.5, 5, 5);
```

```
/* Matrix literal */
fm3 = [[1.0, 2.0, 3.0], [4.0, 5.0, 60], [7.0, 8.0, 9.0]];
```

```
print_mat((fm1 + 1.0) + fm2);
fm1 = fm1 + 1.0;
print_mat((fm1 + 12.0) .. fm2);  /* Matrix multiplication */
print_mat(fm1 * fm2);  /* Hadamard product */
```

```
return 0;
```

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#### Libraries: Math

- Goal: Build generic library that uses externed code mixed with self built code

- Implementation:
  - Extended a significant portion of C standard math library, including trig, exp, log functions
  - Built basic number manipulation extensions
    - e.g. max, min
    - e.g. sqrt, square
  - Combined eigen math library with own code to build useful distributions
    - e.g. rand\_norm() pulls a random number from an input normal distribution
    - e.g. sigmoid() returns a defined value from the sigmoid distribution

# Libraries: DEEP

A basic machine learning library for easily fully-connected, feedforward models

- Arbitrary layer architecture
- Arbitrary cost and activation functions
- User-defined hyperparameters
- Uses every single feature!!

struct fc_model {	
<pre>fmatrix[] train_x;</pre>	
<pre>fmatrix[] train_y;</pre>	
<pre>fmatrix[] test_x;</pre>	
<pre>fmatrix[] test_y;</pre>	
<pre>fmatrix[] biases;</pre>	
<pre>fmatrix[] weights;</pre>	
<pre>int[] layer_sizes;</pre>	
int epochs;	
<pre>int mini_batch_size;</pre>	
float learning_rate;	
<pre>fp (float) weight_init;</pre>	
<pre>fp (float, float) activate;</pre>	
<pre>fp (float, float) activate_prime;</pre>	
<pre>fp (fmatrix, fmatrix, float) cost;</pre>	
<pre>fp (fmatrix, fmatrix, fmatrix, fmatrix) cost_pr.</pre>	ime;

# Demo: MNIST

- Benchmark machine learning problem
- 28x28 grayscale images of handwritten digits
- 60,000 training
- 10,000 test



#### Demo: MNIST

- 97.2% classification accuracy

epochs = 20; learning\_rate = .1; mini\_batch\_size = 10; layer\_sizes = (int[]) {784, 50, 10};

/\* allocate memory \*/
fc = make(struct fc\_model);

fc.train\_x = train\_fm\_images; fc.train\_y = train\_fm\_labels; fc.test\_x = test\_fm\_images; fc.test\_y = test\_fm\_labels; fc.layer\_sizes = layer\_sizes; fc.epochs = epochs; fc.mini batch size = mini batch size; fc.learning\_rate = learning\_rate; fc.weight init = norm init; fc.activate = sigmoid; fc.activate\_prime = sigmoid\_prime; fc.cost = cross\_entropy\_cost; fc.cost prime = cross entropy cost prime;

fc.train();
fc.demo(5);