

Introduction of Functional Programming for Scientists

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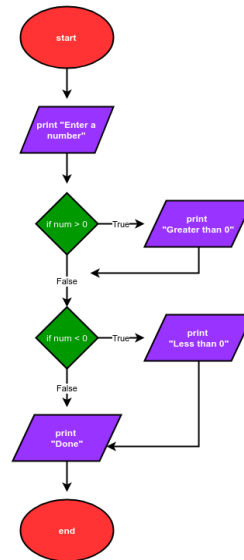
Programming Language Paradigms

- Imperative
 - Procedure Programming
 - Object Oriented Programming
- Declarative
 - Functional Programming
 - ...
- ...

Imperative Programming

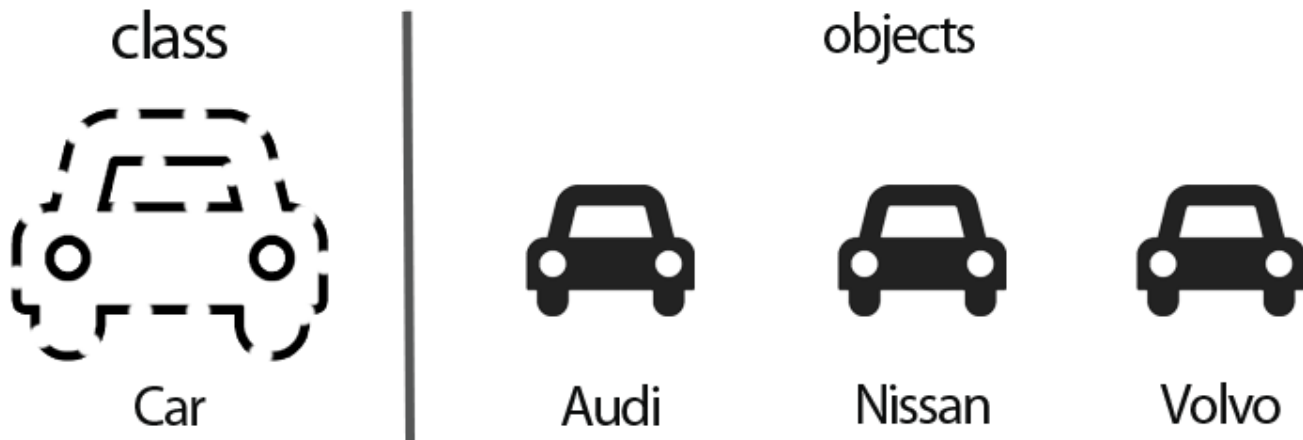
Procedual Programming

- Fortran
- C
- ...



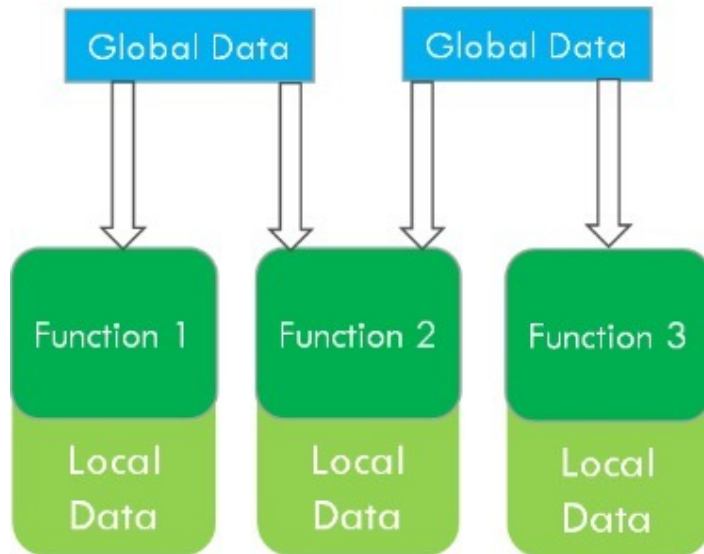
Object Oriented Programming

- C++
- Java
- Python (?)
- ...

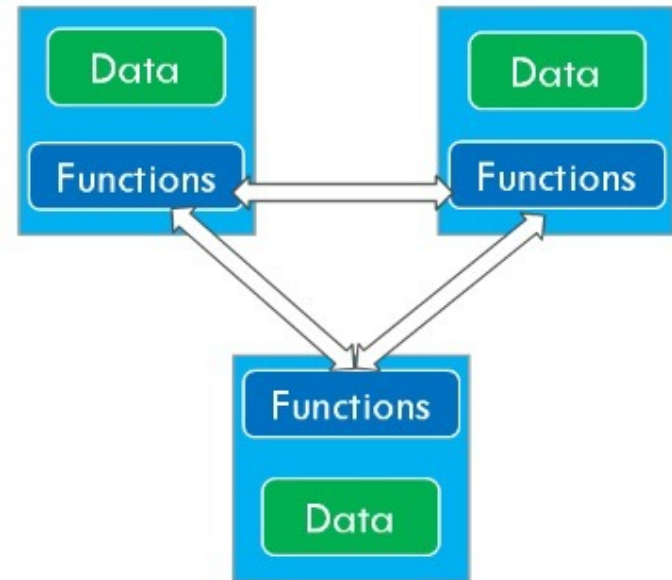


Procedure vs OOP

Procedural Oriented Programming



Object Oriented Programming



OOP in Tensorflow

```
model = tf.keras.models.Sequential([
    tf.keras.layers.Flatten(input_shape=(28, 28)),
    tf.keras.layers.Dense(128, activation='relu'),
    tf.keras.layers.Dropout(0.2),
    tf.keras.layers.Dense(10, activation='softmax')
])

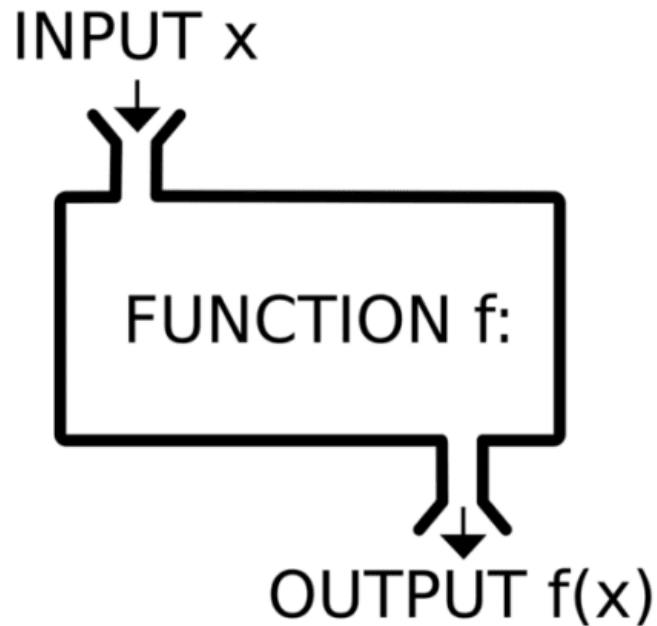
model.compile(optimizer='adam',
              loss='sparse_categorical_crossentropy',
              metrics=['accuracy'])

model.fit(x_train, y_train, epochs=5)
model.evaluate(x_test, y_test, verbose=2)
```

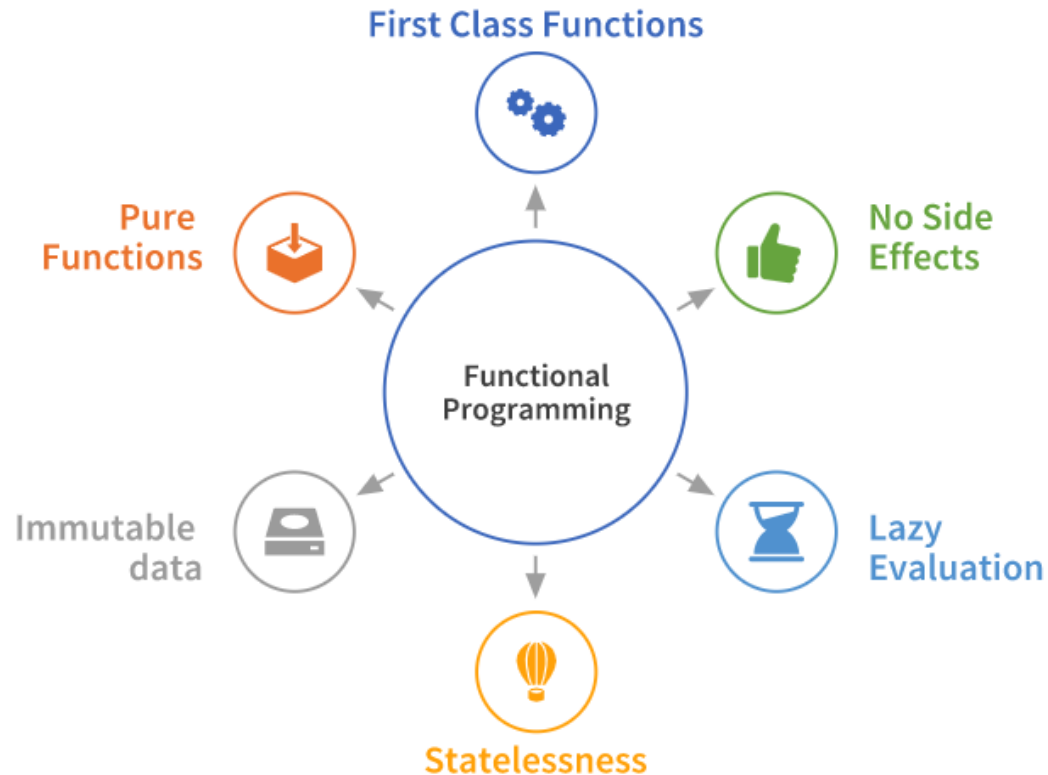
Declarative Programming

Functional Programming

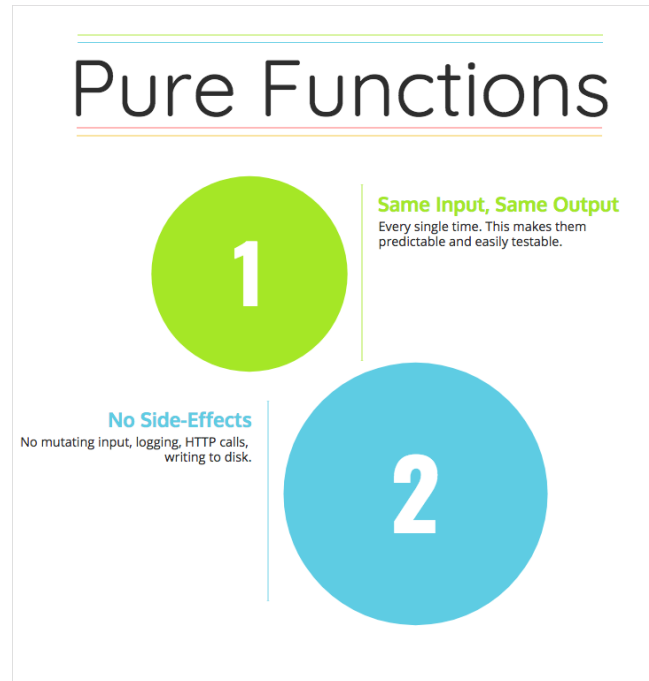
- Lisp
- Haskell
- Clojure
- ...



What is Functional Programming?

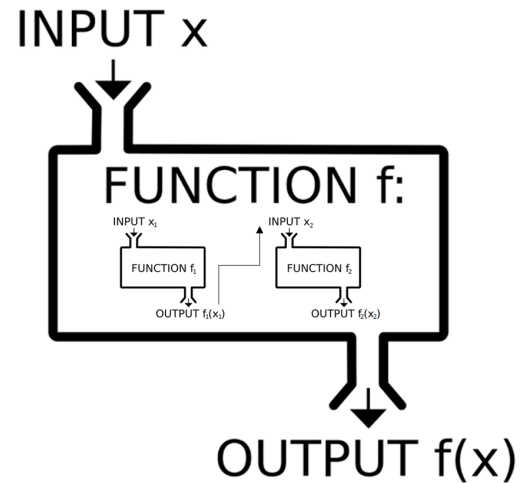


Pure Function



First Class Function

- Higher order function: Input/Output is a function
- Anonymous and nested function
- Non-local variables and closures



Closure - 1

```
def outer_func():  
    x = 5  
    def inner_func(y = 3):  
        return (x + y)  
    return inner_func  
  
a = outer_func()  
print(a())
```

Results?

Closure - 1

```
def outer_func():  
    x = 5  
    def inner_func(y = 3):  
        return (x + y)  
    return inner_func  
  
a = outer_func()  
print(a())
```

Results?

8

Closure - 2

```
text = "global text"
def outer_func():
    text = "enclosing text"
    def inner_func():
        text = "inner text"
        print('inner_func:', text) # inner_func: global text
    print('outer_func:', text) # outer_func: enclosing text
    inner_func()
    print('outer_func:', text) # outer_func: enclosing text

print('global:', text) # global: global text
outer_func()
print('global:', text) # global: global text
```

Results?

Closure - 2

```
text = "global text"
def outer_func():
    text = "enclosing text"
    def inner_func():
        text = "inner text"
        print('inner_func:', text) # inner_func: global text
    print('outer_func:', text) # outer_func: enclosing text
    inner_func()
    print('outer_func:', text) # outer_func: enclosing text

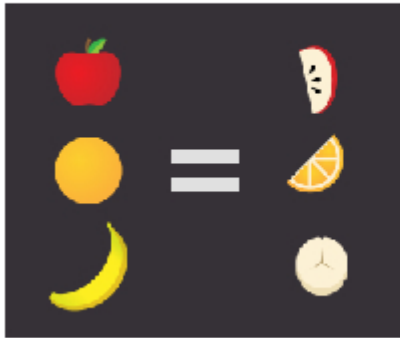
print('global:', text) # global: global text
outer_func()
print('global:', text) # global: global text
```

Results?

```
global: global text
outer_func: enclosing text
inner_func: inner text
outer_func: enclosing text
global: global text
```


Loops => map/reduce/filter

Array.map()



Array.filter()



Array.reduce()



Why FP? Why now?

Concurrency & Parallelism

Distributed Systems

Optimization

Simplicity

Ceding Control to the Language/Runtime

Focusing on Abstraction

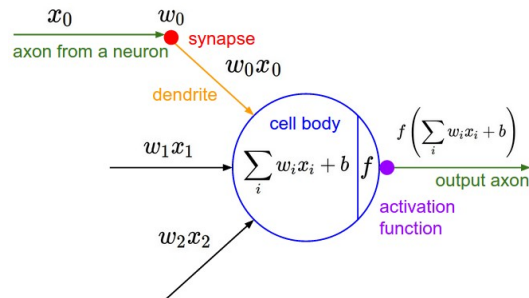
Model becomes Functional

Hide model details in function

```
function compile_model(input_size, output_size)
  unit_size = 16
  model = Flux.Chain(
    Flux.Dense(input_size, unit_size),
    Flux.Dense(unit_size, unit_size),
    Flux.Dense(unit_size, output_size))

  model
end

model = compile_model(input_size, output_size)
y = model(x)
```



More details

```
function compile_model(input_size, output_size, statval)
    unit_size = 16
    model = Flux.Chain(
        Flux.Dense(input_size, unit_size),
        Flux.Dense(unit_size, unit_size),
        Flux.Dense(unit_size, output_size))

    loss(x, y) = Flux.mse(model(x), y)
    # statval : collection of stats such as  $\mu$  and  $\sigma$ 
    # :RMSE : symbol for evaluation metric indicator
    accuracy(data) = evaluation(data, model, statval, :RMSE)
    opt = Flux.ADAM()

    model, loss, accuracy, opt
end

model, loss, accuracy, opt = compile_model(input_size, output_size)
train!(loss, Flux.params(model), train_set, opt)
@show loss(x, y)
@show accuracy(test_set)
```

How evaluation works

Elementary Functions

```
# column-wise sum
_RMSE(y::AbstractVector, ŷ::AbstractVector) =
  sqrt.(sum((y .- ŷ).^2, dims=[1]) / length(y))
_RMSE(y::AbstractMatrix, ŷ::AbstractMatrix) =
  sqrt.(sum((y .- ŷ).^2, dims=[1]) / size(y, 1))
```

Abstraction for any dataset type

```
function RMSE(dataset, model, statvals::AbstractNDSparse)
  _μ = statvals["total", "μ"][:value]
  _σ = statvals["total", "σ"][:value]

  # sum(f, itr) + do block
  # no allocation + mapreduce => faster!
  let cnt = 0
    # column sum for batches
    _rmseval = sum(dataset) do xy
      _rmseval = _RMSE(xy[2], model(xy[1]))

      # replace Inf, NaN to zero, no impact on sum
      replace!(_rmseval, Inf => 0, -Inf => 0, NaN => 0)
      # number of columns which is not Inf and not NaN
      cnt += count(x -> !(isnan(x) || isinf(x)), _rmseval)
    end

    _rmseval
  end

  # rmse mean
  _rmseval / cnt
end
end
```


Abstraction for any single metric

```
function evaluation(dataset, model, statvals::AbstractNDSparse, met
eval(quote
  let _cnt = 0
    # column-wise sum for batches
    _sum = sum($(dataset)) do xy
      # Float or (1 x batch) Array
      _val = $(metric)(xy[2], $(model)(xy[1]))

      # number of columns which is not Inf and not NaN
      # if _val is Float, cnt must be 1
      _cnt += count(x -> !(isnan(x) || isinf(x)), _val)

      # If _val is Array -> use `replace!` to replace Inf, NaN to
      # If _val is Number -> just assign zero
      typeof(_val) <: AbstractArray ? replace!(_val, Inf=> 0, -Inf=>
        zero(_val))

      sum(_val)
    end # do - end

    _cnt = _cnt == 0 && isapprox(_sum, 0.0) ? 1 : _cnt

    # mean
    _sum / _cnt
  end # let - end
```

Evaluate over multiple metrics

```
function evaluations(dataset, model, statvals::AbstractNDSparse, metrics)
    metric_vals = map(metrics) do metric
        evaluation(dataset, model, statvals, metric)
    end # do - end

    Tuple(metric_vals)
end
```

Evaluate over multiple metrics

```
function evaluations(dataset, model, statvals::AbstractNDSparse, metrics)
    metric_vals = map(metrics) do metric
        evaluation(dataset, model, statvals, metric)
    end # do - end

    Tuple(metric_vals)
end
```

Usage

```
eval_metrics = [:RMSE, :MAE, :MSPE, :MAPE]
# test set
for metric in eval_metrics
    _eval = evaluation(test_set, model, statval, metric)
    @info " $(string(ycol)) $(string(metric)) for test : ", _eval
end
```