

Outline

- ① Hyperparameter Tuning
- ② Batch Normalization
- ③ Multi-class classification
- ④ TensorFlow

Hyperparameter Tuning

Importance

- ① Learning Rate α
- ② Momentum β , minibatch size, number of hidden units
- ③ number of layers learning rate decay
- ④ others, like the parameters in Adam

Scale of Learning Rate

Learning rate should be sampled on different scales

For example, $\alpha \in [0.001, 0.01, 0.1, 1]$

Intuitive example: learn w_0 in the problem $y = w_0x$.

Loss function $l(w) = (y - wx)^2$

Derivative $l'(w) = 2x(wx - y)$

Update by $w^{(t+1)} = w^{(t)} - 2\alpha x(w^{(t)}x - y)$

Assume $w_0 = 2$ and we have train model twice on two different datasets

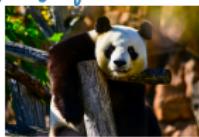
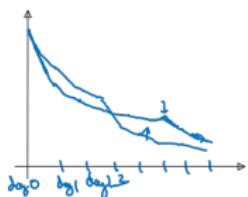
Dataset1: $(x, y) = \{(1, 2); (2, 4)\}$

Dataset2: $(x, y) = \{(10, 20); (20, 40)\}$

The learning rate in dataset 2 should be 100 times smaller than that in dataset 1!

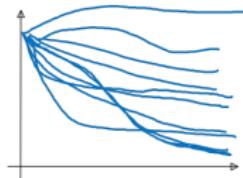
Tuning Can be Slow

Babysitting one
model



Panda ↵

Training many models
in parallel

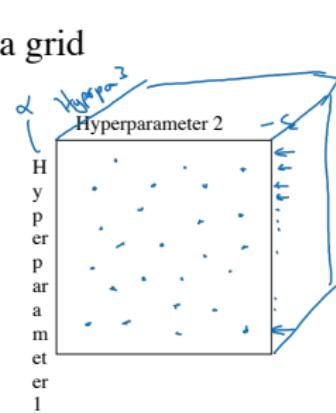
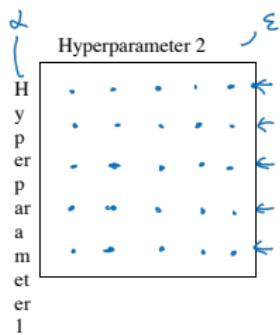


Caviar ↵ Andrew

Other Parameters

Sample momentum $\beta \in [0.9, 0.99, 0.999]$ Computing the "moving average" of $1/(1 - \beta) = [10, 100, 1000]$ minibatches Sample randomly

Try random values: Don't use a grid

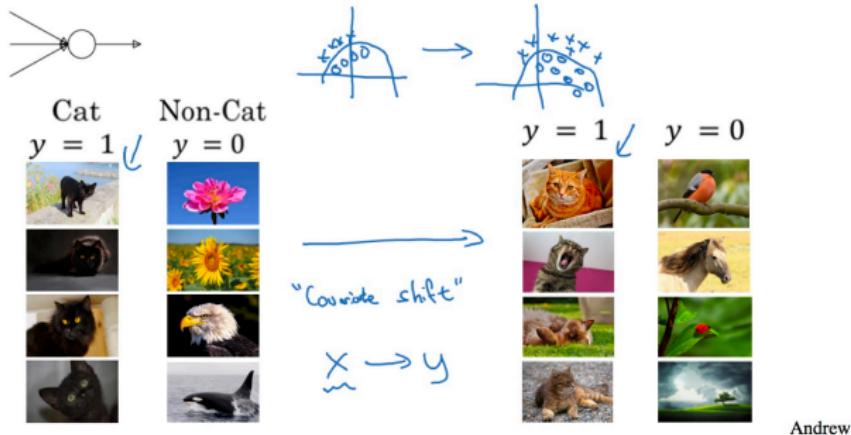


Andrew

Batch Normalization

Problem: Minibatch has small size and not i.i.d. between them
Different colors of cats in different mini-batches

Learning on shifting input distribution



ALL neuron activities will be affected and need to be adjusted
Not only input ones

Batch Normalization

Input: Values of x over a mini-batch: $\mathcal{B} = \{x_{1\dots m}\}$;

Parameters to be learned: γ, β

Output: $\{y_i = \text{BN}_{\gamma, \beta}(x_i)\}$

$$\mu_B \leftarrow \frac{1}{m} \sum_{i=1}^m x_i \quad // \text{mini-batch mean}$$

$$\sigma_B^2 \leftarrow \frac{1}{m} \sum_{i=1}^m (x_i - \mu_B)^2 \quad // \text{mini-batch variance}$$

$$\hat{x}_i \leftarrow \frac{x_i - \mu_B}{\sqrt{\sigma_B^2 + \epsilon}} \quad // \text{normalize}$$

$$y_i \leftarrow \gamma \hat{x}_i + \beta \equiv \text{BN}_{\gamma, \beta}(x_i) \quad // \text{scale and shift}$$

Subtract mean and variance information from minibatch

If any of these are necessary, then learn it through the network.

Ideally, in the cat example, $\hat{\beta} = 0$ if we don't want to distinguish cats with different colors, and $\hat{\beta}! = 0$ otherwise.

Graph

Value calculations

$$c = a^2 + 2b - 2a$$

$$\text{temp1} = a^2$$

$$\text{temp2} = 2 * b$$

$$\text{temp3} = \text{temp1} + \text{temp2}$$

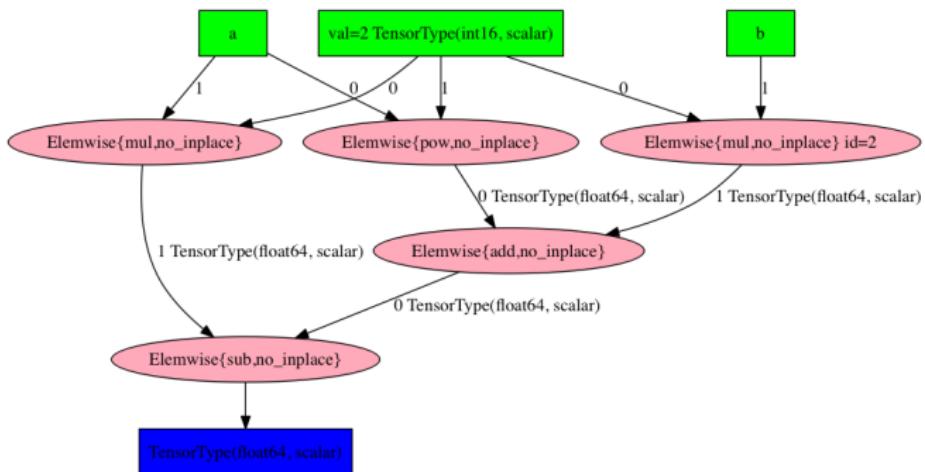
$$\text{temp4} = 2 * a$$

$$c = \text{temp3} - \text{temp4}$$

No connection. How to compute gradient?

Graph

$$c = a^2 + 2b - 2a$$



Session, Operation and Tensor

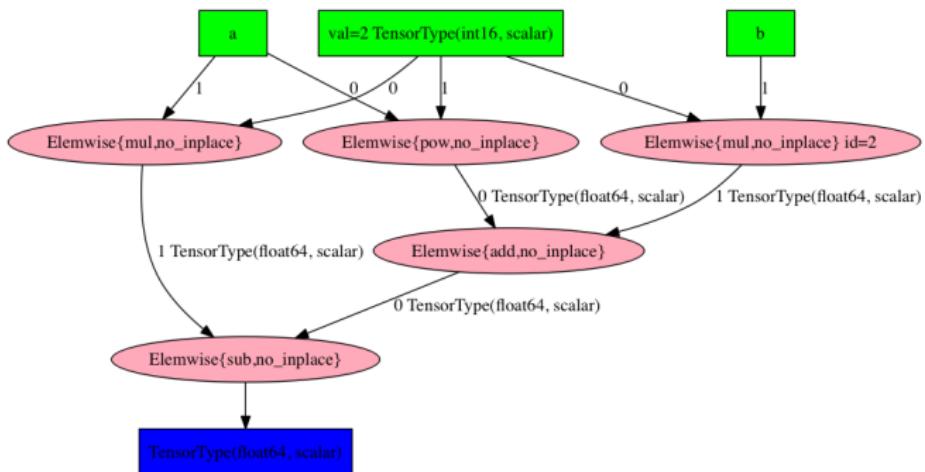
Operation: a graph node that performs operations on tensors

Tensors: represents one of the outputs of operations

Session: where operations are executed and tensors are evaluated

Graph

$$c = a^2 + 2b - 2a$$



Handle

Imagine you want to read a book in a library. You have two choices:

- 1) obtain the **content** of the book.
- 2) obtain the **address** of the book.

Handle: use method 2.

- + Gradient calculation is automatic
- + It consumes no time to obtain an address in memory
- + Copying a large object is slow
- + Higher level of abstraction. Same code can run on CPU and GPU
- Unexpected changes in its value.

Ex: $a = 3; b = a; a = 4; b = ?$

Tensor Types

Constant: constant tensors. Widely used implicitly by TensorFlow.

e.g., $b = 2a$, 2 will be converted to a constant tensor.

Variable: a variable

- 1) has a fixed shape;
- 2) need to be initialized before calling operations on it.
- 3) is automatically collected in the graph.

examples of variables?

Placeholder: used as a handle to feed data.

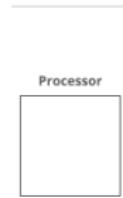
(Go to exercise part1)

Handle Revisited: Datatype

Knowing address is not enough.

"int32" and 'float32' are datatypes that specifies

- 1) how long the data is
- 2) how to decode the "random" 0-1 strings.



Processor	0 00000000
	1 00000001
	2 00000011
	3 00000011
	4 00000100
	5 00000100
	6 00000101
	7 00000110
	8 00000110
	9 00000111
	10 00001000
	11 00001010
	12 00001010

Tensor types revisited

Constant:

- ① usage: implicit on constant numbers mostly
- ② datatype: implicit
- ③ shape: implicit

Variable:

- ① usage: parameters and cost function
- ② datatype: user-specified
- ③ shape: user-specified

Placeholder

- ① usage: data input
- ② datatype: user-specified
- ③ shape: given by real input data