# GlobalDataLoader in Multi DeepLearning Task

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## Table of Contents

Introduction

② Global DataLoader

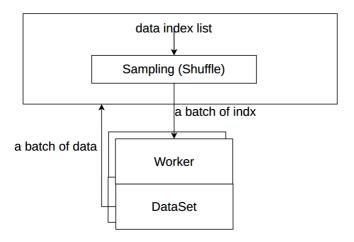
## Table of Contents

Introduction

② Global DataLoader



# DataLoader in Pytorch





4/29

# Problem: Repeated Reading and Processing

#### Situation

To compare the performance of different algorithms, Many DeepLearning tasks are training in the same Dataset.

#### **Problem**

Every task has its own DataLoader. So the data will be repeatedly read and processed by different tasks.

#### Result

As the number of tasks increases, so does the training time. And what increases is the time to load the data

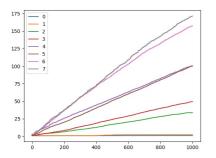


Figure: data loading time

Figure: data training time

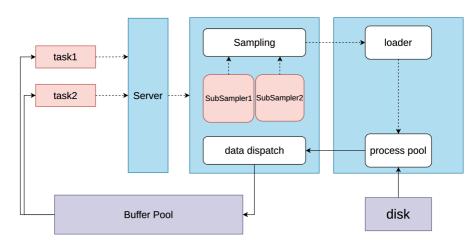
## Table of Contents

Introduction

② Global DataLoader



## Architecture



## Sampling: problem description

#### Defination

For a single task, the sampler needs to select an element from the index set S.

Similarly, for multiple tasks, the sampler needs to select some elements  $\{s_1, s_2, ...\}$  from multiple sets  $\{S_1, S_2, ...\}$ 

### Requirments

- The index in the set S should be randomly sampled.  $p(s_i) = \frac{1}{|S_i|}$
- Duplicate indexes need to be merged. Maximize the probability  $p(s_i = s_j)$
- There can be no problem of "starvation"



# Independently Sampling Algorithm

### Assumption

There are two sets:  $S_1$ ,  $S_2$ , and their length is  $n_1$ ,  $n_2$ . The intersection set of them is  $S_i$ , whose length is  $n_i$ . We divide the set  $S_1$  into  $S_i$  and  $S_{d1} = S_1 - S_i$ . We divide the set  $S_2$  into  $S_i$  and  $S_{d2} = S_2 - S_i$ .

### Example

$$S_1 = \{1, 2, 3, 4, 5\} = \{1, 2, 3\} \cup \{4, 5\}$$
  
 $S_2 = \{1, 2, 3, 6, 7\} = \{1, 2, 3\} \cup \{6, 7\}$ 



## Independently Sampling Algorithm

### Algorithm

- *S*<sub>1</sub>:
  - $step_{11}$ : randomly select a set from  $S_i$  and  $S_{d1}$
  - $step_{12}$ : if the set is  $S_{d1}$ , randomly select a element1 from  $S_{d1}$
  - $step_{13}$ : if the set is  $S_i$ , randomly select a element1 from  $S_i$
- *S*<sub>2</sub>:
  - $step_{21}$ : randomly select a set from  $S_i$  and  $S_{d2}$
  - $step_{22}$ : if the set is  $S_{d2}$ , randomly select a element2 from  $S_{d2}$
  - $step_{23}$ : if the set is  $S_i$ , randomly select a element2 from  $S_i$

### Probability

$$p(element1) = \frac{1}{n_1}, p(element2) = \frac{1}{n_2}$$

$$p(element1 = element2) = \frac{n_i}{n_1 * n_2}$$
(1)

# Dependently Sampling Algorithm I

#### Idea

The  $step_{13}$  is same as  $step_{23}$ . We can merge them.

### Algorithm

- *S*<sub>1</sub>:
  - $step_{11}$ : randomly select a set from  $S_i$  and  $S_{d1}$
  - $step_{12}$ : if the set is  $S_{d1}$ , randomly select a element1 from  $S_{d1}$
  - $step_{13}$ : if the set is  $S_i$ , go to  $step_{33}$
- *S*<sub>2</sub>:
  - $step_{21}$ : randomly select a set from  $S_i$  and  $S_{d2}$
  - $step_{22}$ : if the set is  $S_{d2}$ , randomly select a element2 from  $S_{d2}$
  - $step_{23}$ : if the set is  $S_i$ , go to  $step_{33}$
- step<sub>33</sub>: randomly select a element from S<sub>i</sub>



# Dependently Sampling Algorithm I

### Probability

$$p(element1) = \frac{1}{n_1}$$

$$p(element2) = \frac{1}{n_2}$$

$$p(element1 = element2) = \frac{n_i}{n_1} * \frac{n_i}{n_1} = \frac{n_i^2}{n_1 * n_2}$$
(2)

### Example

$$S_1 = \{1, 2, 3\} \cup \{4, 5\}; S_2 = \{1, 2, 3\} \cup \{6, 7\}$$

- As for  $S_1$ , select set  $\{1, 2, 3\}$  with probability 0.6
- As for  $S_2$ , select set  $\{1, 2, 3\}$  with probability 0.6
- Finally, randomly sampling element in  $\{1, 2, 3\}$
- p(element1 = element2) = 0.6 \* 0.6 = 0.36

# Dependently Sampling Algorithm I

#### Idea

The  $step_{11}$  and  $step_{21}$  are similar. We can merge them.

### Algorithm

- $step_1$ : randomly select a set from  $S_i$  and  $S_{d1}$
- *S*<sub>1</sub>:
  - $step_{12}$ : if the set is  $S_{d1}$ , randomly select a element1 from  $S_{d1}$
  - $step_{13}$ : if the set is  $S_i$ , go to  $step_{33}$
- *S*<sub>2</sub>:
  - $step_{22}$ : if the set is not  $S_{d1}$ , randomly select a element2 from  $S_{d2}$
  - $step_{23}$ : if the set is  $S_i$ , go to  $step_{33}$
- $step_{33}$ : randomly select a element from  $S_i$



## **Problem**

### Probability

As for  $S_1$ :

$$p_1(S_i) = \frac{n_i}{n_1}$$

$$p_1(S_{d1}) = 1 - \frac{n_i}{n_1}$$
(3)

As for  $S_2$ :

$$p_2(S_i) = \frac{n_i}{n_2}$$

$$p_2(S_{d2}) = 1 - \frac{n_i}{n_2}$$
(4)

#### Problem

if  $n_1 \neq n_2$ , then  $p_1(S_i) \neq p_2(S_i)$  and  $p_1(S_{d1}) \neq p_2(S_{d2})$ 

## Case1: n1 < n2

#### Problem

$$(p_1(S_i) = \frac{n_i}{n_1}) > (p_2(S_i) = \frac{n_i}{n_2})$$
 (5)

### **Approach**

So in  $step_1$ , when select  $S_i$ , it should be changed  $S_{d2}$  in probability of p. The equation is

$$p_2(S_i) = p_1(S_i) * (1 - p) = \frac{n_i}{n_2}$$

$$p_2(S_{d2}) = p_2(S_{d1}) + p_1(S_i) * p = 1 - \frac{n_i}{n_2}$$
(6)

then

$$p=1-\frac{n_1}{n_2}$$



# Dependently Sampling Algorithm II

### Algorithm

- $step_1$ : randomly select a set S from  $S_i$  and  $S_{d1}$
- *S*<sub>1</sub>:
  - $step_{12}$ : if the S is  $S_{d1}$ , randomly select a element1 from  $S_{d1}$
  - $step_{13}$ : if the S is  $S_i$ , go to  $step_3$
- *S*<sub>2</sub>:
  - $step_{22}$ : if the S is  $S_{d2}$ , randomly select a element2 from  $S_{d2}$
  - $step_{23}$ : if the S is  $S_i$ 
    - randomly let  $S=S_{d1}$  in probability of  $1-\frac{n_1}{n_2}$
    - if S is  $S_{d2}$ , randomly select a element2 from  $S_{d2}$
    - if S is  $S_i$ , go to  $step_3$
- $step_3$ : randomly select a element from  $S_i$



# Dependently Sampling Algorithm II

### Probability

$$p(element1 = element2) = p_1(S_i) * (\frac{n_1}{n_2}) = \frac{n_i}{n_2}$$

### Example

$$S_1 = \{1, 2, 3\} \cup \{4\}; \ S_2 = \{1, 2, 3\} \cup \{6, 7\}$$

- In  $step_1$ , select set  $\{1,2,3\}$  with probability  $\frac{3}{5}=0.75$
- Then as for  $S_2$ , select set  $\{1,2,3\}$  with probability  $\frac{4}{5}=0.8$
- Finally, randomly sampling element in  $\{1, 2, 3\}$
- p(element1 = element2) = 0.75 \* 0.8 = 0.6



## Case1: n1 > n2

#### **Problem**

$$(p_1(S_{d1}) = 1 - \frac{n_i}{n_1}) > (p_2(S_{d_2}) = 1 - \frac{n_i}{n_2})$$
(7)

#### Approach

So in  $step_1$ , when select  $S_{d1}$ , it should be changed  $S_i$  in probability of p. The equation is

$$p_2(S_i) = p_1(S_i) + p_2(S_{d1}) * p = \frac{n_i}{n_2}$$

$$p_2(S_{d2}) = p_2(S_{d1}) * (1 - p) = 1 - \frac{n_i}{n_2}$$
(8)

then

$$p = 1 - \frac{n1 * (n2 - n_c)}{n2 * (n1 - n_c)}$$

# Dependently Sampling Algorithm II

### $\mathsf{Algorithm}$

- $step_1$ : randomly select a set S from  $S_i$  and  $S_{d1}$
- *S*<sub>1</sub>:
  - $step_{12}$ : if the S is  $S_{d1}$ , randomly select a element1 from  $S_{d1}$
  - $step_{13}$ : if the S is  $S_i$ , go to  $step_3$
- *S*<sub>2</sub>:
  - $step_{22}$ : if the S is  $S_{d2}$ , randomly select a element2 from  $S_{d2}$ 
    - randomly let  $S = S_i$  in probability of  $1 \frac{n1*(n2-n_c)}{n2*(n1-n_c)}$
    - if S is  $S_{d2}$ , randomly select a element2 from  $S_{d2}$
    - if S is  $S_i$ , randomly select a element2 from  $S_i$
  - $step_{23}$ : if the S is  $S_i$ , go to  $step_3$
- $step_3$ : randomly select a element from  $S_i$



# Dependently Sampling Algorithm II

## Probability

$$p(element1 = element2) = p_1(S_i)) = \frac{n_i}{n_1}$$

### Example

$$S_1 = \{1, 2, 3\} \cup \{4, 5\}; \ S_2 = \{1, 2, 3\} \cup \{6\}$$

- In  $step_1$ , select set  $\{1, 2, 3\}$  with probability  $\frac{3}{4} = 0.6$
- Then  $S_2$  must select  $\{1, 2, 3\}$
- Finally, randomly sampling element in  $\{1, 2, 3\}$
- p(element1 = element2) = 0.6



# Why not shuffle $S_1 \cup S_2$

### Probability

$$p(element1 = element2) = \frac{|S_1 \cap S_2|}{|S_1 \cup S_2|} < \frac{|S_1 \cap S_2|}{\max(|S_1|, |S_2|)}$$

### randomly select in $S_1 \cup S_2$ : "starvation"

if 
$$|S_1|=99, |S_2|=1$$
, then  $p(element \in S_1)=0.99, p(element \in S_2)=0.01$  So task2 may be starving

### shuffle $S_1 \cup S_2$ : offset

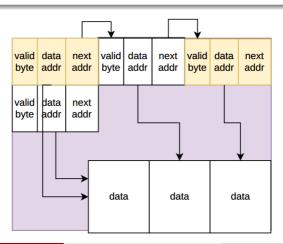
$$S_1 = \{3,4\}, S_2 = \{1,3,4\}$$
  
 $\{3,4\}$   
 $\{1,2,3,4\}$   
 $\{1, 3,4\}$ 



## Buffer Pool: Data Structure

#### data

- There are two kinds of nodes: inode and datanode
- Every task has a head inode address



## Buffer Pool: Valid Byte

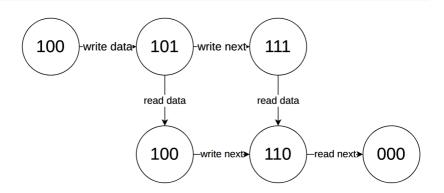
### valid byte

- data bit: If the data bit is equal to 1, the data addr is valid.
   Otherwise invalid
- next bit: If the next bit is equal to 1, the next addr is valid.
   Otherwise invalid
- used bit: If the used bit is equal to 1, this inode is used by some tasks

## Buffer Pool: Automata

### valid byte

used bit | next bit | data bit |



## Buffer Pool: allocate inode

#### $\mathsf{case}1$

There is enough free space to allocate

```
if inode_tail + inode_size > data_head:
    return inode_tail
```

#### case2

#### Free Some unused inode

```
for head in all_heads:
    if check_free(head) is True:
    return head
```

## Buffer Pool: allocate data node

#### case1

There is enough free space to allocate

```
1 | if inode_tail + inode_size > data_head:
2 | return inode_tail
```

#### case2

Free Some unused datanode

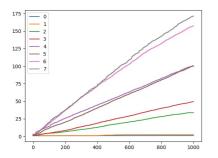
```
free = True
for datanode in all_datanodes:
    for ref in refs of datanode:
        if databit(ref) == 0 && dataaddr(ref) == datanode:
            free = False
            break
if free is True:
    return datanode
```

## Table of Contents

Introduction

② Global DataLoader





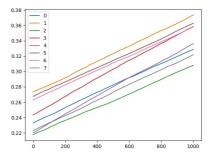


Figure: time

Figure: time with GlobalDataLoader