MindFormers实验手册-v1.0

1. 运行环境及模型信息

目标:以LLaMA2-7B模型为例,熟悉Mindformers大模型套件的微调和推理流程

配置: Atlas800 A2 单节点8卡

标准镜像: http://mirrors.cn-central-221.ovaijisuan.com/detail/129.html

运行环境:

名称	版本
MindFormers	r1.1
MindPet	1.0.3
MindSpore	2.3.rc2
CANN	8.0.RC1.beta1
驱动和固件	8.0.RC1.beta1
Python	3.9

2. 相关材料与准备工作

实操模型: LLaMA2-7B

模型权重和相关文件放置在指定目录下,按照以下结构形式组织文件内容:

```
mindformers

└── checkpoint_download

└── 11ama2

├── 11ama2_7b.ckpt # 模型权重

├── 11ama2_7b.yaml # 模型配置文件

└── tokenizer.model # 模型tokenizer文件
```

数据集路径: /data01/datasets/

```
datasets
├── alpaca_data.json
└── belle_chat_ramdon_10k.json
```

3. 微调

以下所有操作均在容器内进行

3.1 数据预处理

Step1. 执行 alpaca_converter.py ,使用fastchat工具添加prompts模板,将原始数据集转换为多轮对话格式

```
cd ./mindformers/tools/dataset_preprocess/llama/

python alpaca_converter.py \
   --data_path /data01/datasets/alpaca_data.json \
   --output_path /{path}/alpaca-data-conversation.json
```

```
# 参数说明
data_path: 存放alpaca数据的路径
output_path: 输出转换后对话格式的数据路径
```

Step2. 执行 11ama_preprocess.py ,进行数据预处理、Mindrecord数据生成,将带有prompt模板的数据转换为mindrecord格式

```
python llama_preprocess.py \
--dataset_type qa \
--input_glob /{path}/alpaca-data-conversation.json \
--model_file /{path}/tokenizer.model \
--seq_length 4096 \
--output_file /{path}/alpaca-fastchat4096.mindrecord
```

转换成功日志:

```
/root/miniconda3/envs/ryj_ms2.3/lib/python3.9/site-packages/numpy/core/getlimits.py:549: UserWarning: The value of the smallest subnormal for <class 'numpy.float64'> type is zero setatr(self, word, getlett/machar, word).flation/journal.float/self.py:69: UserWarning: The value of the smallest subnormal for <class 'numpy.float64'> type is zero. return self. float to striself.smallest subnormal for <class 'numpy.float64'> type is zero. return self. float to striself.smallest subnormal for <class 'numpy.float32'> type is zero. return self.float to striself.smallest subnormal for <class 'numpy.float32'> type is zero. return self.float to strit/machar, word).flation/journal.float/self.py:float32'> type is zero. return self.float to striself.smallest.py:float32'> type is zero. return self.float to striself.smallest.gupy.float32'> type is zero. return self.float to striself.gupy.float32'> type is zero. return self.float to striself.gupy.float32'> type is zero. return self.float to striself.gupy.float32'> type is zero.
```

3.2 全参微调

Step1. 打开 config/llama2/finetune_llama2_7b.yaml, 训练数据集路径设置为3.1中预处理好的微调数据集路径,并在 input_columns 中添加 labels

```
train_dataset: &train_dataset
  data_loader:
    type: MindDataset
    dataset_dir: "/{path}/alpaca-fastchat4096.mindrecord"
    shuffle: True
  input_columns: ["input_ids", "labels"]
```

Step2. 修改运行相关配置

```
load_checkpoint: '/{path}/llama2_7b.ckpt'# 使用权重的绝对路径auto_trans_ckpt: True# 打开权重自动转换use_parallel: True# 开启并行设置run_mode: 'finetune'# 设置微调模式
```

Step3. 修改并行配置,并行策略可以进行小范围修改进行尝试

```
parallel_config:
  data_parallel: 8
  model_parallel: 1
  pipeline_stage: 1
  micro_batch_num: 1
```

Step4. 使用msrun启动分布式微调任务

```
bash scripts/msrun_launcher.sh "run_mindformer.py \
--config configs/llama2/finetune_llama2_7b.yaml \
--run_mode finetune" 8
```

查看微调执行情况

```
tail -f output/msrun_log/worker_0.log
```

构建模型后,权重切分和网络编译需要等待10分钟左右。待出现loss则训练拉起成功

实时查看NPU使用情况

```
watch -n 1 npu-smi info
```

Every 1.0s:	npu-smi info					
+ npu-smi 23	.0.3	Version: 2	23.0.3			+
+ NPU Name Chip		Health Bus-Id	Power(W) AICore(%)	Temp(C) Memory-	Usage(MB)	+ Hugepages-Usage(page) HBM-Usage(MB)
+======== 0 910B 0	======================================	OK 0000:C1:00.0	284.3 73	58 0 / (====== 0	0 / 0 63056/ 65536
1 910B 0	3	OK 0000:C2:00.0	270.5 73	55 0 / (====== 0	0 / 0 63059/ 65536
2 910B 0	3	OK 0000:81:00.0	252.2 74	54 0 /	====== 0	0 / 0 63056/ 65536
3 910B 0	3 	0K 0000:82:00.0	252.2 73	56 0 /	====== 0	0 / 0 63055/ 65536
4 910B 0	3	OK 0000:01:00.0	271.7 73	68 0 /	9 0	0 / 0 63062/ 65536
5 910B 0	3	0K 0000:02:00.0	280.0 74	67 0 /	====== 0	0 / 0 63064/ 65536
6 910B 0	3	OK 0000:41:00.0	240.1 6	66 0 /	====== 0	0 / 0 63109/ 65536
7 910B 0	3	OK 0000:42:00.0	203.6 0	65 0 / 0	0	0 / 0 63061/ 65536
+	ip	Process id	Process nam	e	Pr	ocess memory(MB)
0 0		2445996	python		59	710
1 0		2446011	python		59	710
+=====================================		2446048	python		 59	======================================
+=====================================		·======+ 2446053	python		 59	======================================
+=====================================		-======+ 2446064	python	=====	59	:======+++++++++++++++++++++++++++++++
†=====================================	+ 	2446107	python	======	59	:======= 9710
+=====================================		========++++++++++++++++++++++++++++++	python	======	======= 59	:=======++++++++++++++++++++++++++++++
+=====================================	=====================================	2446119	python		59	======++++++++++++++++++++++++++++++++
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3.3 LoRA微调 (课后练习)

参考MindFormers开源仓中的: LLaMA2微调文档

4. 推理

以下所有操作均在容器内进行

4.1 自回归推理

启动python终端,输入以下内容以完成自回归推理:

```
# 设置MindSpore图模式并指定使用的device_id
import mindspore as ms
ms.set_context(mode=0, device_target="Ascend", device_id=0)
from mindformers import AutoModel, AutoTokenizer

# 通过AutoModel接口实例化模型
model = AutoModel.from_pretrained("llama2_7b", use_past=False, seq_length=512)
# 通过AutoTokenizer接口实例化tokenizer
tokenizer = AutoTokenizer.from_pretrained("llama2_7b")
```

```
# 生成输入
input_ids = tokenizer("I love Beijing, because")["input_ids"]
# 调用model.generate()接口执行文本生成推理,多次执行推理,规避首次编图耗时
for i in range(5):
    output = model.generate(input_ids, do_sample=True, top_k=3)
    # 解码并打印输出
    print(tokenizer.decode(output))
```

推理结果:

```
2024-05-23 14:07:50,779 - mindformers/generation/text_generation/text_generation/text_generation/text_generation/text_generation-text_generation/text_generation-text_generation-text_generation-text_generation-text_generation-text_generation-text_generation-text_generation-text_generation-text_generation-text_generation-text_generation-text_generation-text_generation-text_generation-text_generation-text_generation-text_generation-text_generation-text_generation-text_generation-text_generation-text_generation-text_generation-text_generation-text_generation-text_generation-text_generation-text_generation-text_generation-text_generation-text_generation-text_generation-text_generation-text_generation-text_generation-text_generation-text_generation-text_generation-text_generation-text_generation-text_generation-text_generation-text_generation-text_generation-text_generation-text_generation-text_generation-text_generation-text_generation-text_generation-text_generation-text_generation-text_generation-text_generation-text_generation-text_generation-text_generation-text_generation-text_generation-text_generation-text_generation-text_generation-text_generation-text_generation-text_generation-text_generation-text_generation-text_generation-text_generation-text_generation-text_generation-text_generation-text_generation-text_generation-text_generation-text_generation-text_generation-text_generation-text_generation-text_generation-text_generation-text_generation-text_generation-text_generation-text_generation-text_generation-text_generation-text_generation-text_generation-text_generation-text_generation-text_generation-text_generation-text_generation-text_generation-text_generation-text_generation-text_generation-text_generation-text_generation-text_generation-text_generation-text_generation-text_generation-text_generation-text_generation-text_generation-text_generation-text_generation-text_generation-text_generation-text_generation-text_generation-text_generation-text_generation-text_generation-text_generation-text_generati
```

4.2 增量推理

启动python终端,输入以下内容以完成增量推理:

```
# 设置MindSpore图模式并指定使用的device_id
import mindspore as ms
ms.set_context(mode=0, device_target="Ascend", device_id=0)
from mindformers import AutoModel, AutoTokenizer

# 通过AutoModel接口实例化模型
model = AutoModel.from_pretrained("llama2_7b", use_past=True, seq_length=512)
# 通过AutoTokenizer接口实例化tokenizer
tokenizer = AutoTokenizer.from_pretrained("llama2_7b")

# 生成输入
input_ids = tokenizer("I love Beijing, because")["input_ids"]

# 调用model.generate()接口执行文本生成推理,多次执行推理,规避首次编图耗时
for i in range(5):
    output = model.generate(input_ids, do_sample=True, top_k=3)
    # 解码并打印输出
    print(tokenizer.decode(output))
```

推理结果:

4.3 流式推理

启动python终端,输入以下内容以完成流式推理:

```
# 设置MindSpore图模式并指定使用的device_id
import mindspore as ms
ms.set_context(mode=0, device_target="Ascend", device_id=0)
from mindformers import AutoModel, AutoTokenizer

# 通过AutoModel接口实例化模型
model = AutoModel.from_pretrained("llama2_7b", use_past=True, seq_length=512)
# 通过AutoTokenizer接口实例化tokenizer
```

```
tokenizer = AutoTokenizer.from_pretrained("llama2_7b")

# 生成输入
input_ids = tokenizer("I love Beijing, because")["input_ids"]

# 标准输出流
from mindformers import TextStreamer
streamer = TextStreamer(tokenizer)

# 调用model.generate()接口执行文本生成推理,多次执行推理,规避首次编图耗时
for i in range(5):
    output = model.generate(input_ids, do_sample=True, top_k=3,
    streamer=streamer)

# 解码并打印输出
    print(tokenizer.decode(output))
```

推理结果:

4.4 Batch推理

启动python终端,输入以下内容以完成batch推理:

```
# 设置MindSpore图模式并指定使用的device_id
import mindspore as ms
ms.set_context(mode=0, device_target="Ascend", device_id=0)
from mindformers import AutoModel, AutoTokenizer
# 通过AutoModel接口实例化模型
# 多batch推理时模型实例化时batch_size设置为对应值
model = AutoModel.from_pretrained("llama2_7b", batch_size=4, use_past=True,
seq_length=512)
# 通过AutoTokenizer接口实例化tokenizer
tokenizer = AutoTokenizer.from_pretrained("llama2_7b")
# 生成多batch输入
input_list = ["Hey how are you doing today?",
             "I love Beijing, because",
             "LLaMA is a",
             "Huawei is a company that"]
input_ids = tokenizer(input_list, max_length=64, padding="max_length")
["input_ids"]
# 调用model.generate()接口执行文本生成推理,多次执行推理,规避首次编图耗时
for i in range(5):
   output = model.generate(input_ids, do_sample=True, top_k=3)
   # 解码并打印输出
   print(tokenizer.decode(output))
```

2024-08-23 1472-188, 165 - mindromers/generation/text_generator.py882]. 1000 - total time. 12,20066209011475 s; generated tokens, 1962 tokens, generated speed, 143.4000906112722 tokens/s
2424-08-23 1472-188, 165 - mindromers/geninders/generation/text_gen

4.5 分布式推理 (课后练习)

请参照<u>分布式推理教程</u>和LLaMA2多卡推理文档)进行学习