

ONNX 格式介绍

Mingzhu Yan

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1. Pytorch 手写数字识别

1.1 神经网络

```
class Net(nn.Module):
    def __init__(self):
        super(Net, self).__init__()
        self.conv1 = nn.Conv2d(1, 10, kernel_size=5)
        self.conv2 = nn.Conv2d(10, 20, kernel_size=5)
        self.conv2_drop = nn.Dropout2d()
        self.fc1 = nn.Linear(320, 50)
        self.fc2 = nn.Linear(50, 10)

    def forward(self, x):
        x = F.relu(F.max_pool2d(self.conv1(x), 2))
        x = F.relu(F.max_pool2d(self.conv2_drop(self.conv2(x)), 2))
        x = x.view(-1, 320)
        x = F.relu(self.fc1(x))
        x = F.dropout(x, training=self.training)
        x = self.fc2(x)
        return F.log_softmax(x, dim=1)
```

1.2 训练并保存快照

```
def train(epoch):
    network.train()
    for batch_idx, (data, target) in enumerate(train_loader):
        optimizer.zero_grad()

        output = network(data)

        loss = F.nll_loss(output, target)
        loss.backward()
        optimizer.step()

        if batch_idx % log_interval == 0:
            torch.save(network.state_dict(), "./model.pth")
            torch.save(optimizer.state_dict(), "./optimizer.pth")
```

1.3 Pytorch 默认模型格式 .pth

```
{  
    "conv1.weight": {  
        "shape": [  
            10,  
            1,  
            5,  
            5  
        ],  
        "values_preview": [  
            -0.04187455028295517,  
            0.33693239092826843,  
            0.22457480430603027,  
            0.2785506844520569,  
            0.24370838701725006  
        ]  
    },  
    "conv1.bias": {  
        "shape": [  
            10  
        ],  
        "values_preview": [  
            -0.01581696793437004,  
            -0.008713895455002785,  
            0.19034163653850555,  
            0.10842573642730713,  
            0.15954284369945526  
        ]  
    },  
    "conv2.weight": {  
        "shape": [  
            20,  
            10,  
            5,  
            5  
        ],  
        "values_preview": [  
            -0.008292394690215588,  
            -0.08871737867593765,  
            -0.04187455028295517,  
            0.33693239092826843,  
            0.22457480430603027,  
            0.2785506844520569,  
            0.24370838701725006  
        ]  
    }  
}
```

model.pth

- 是个压缩包, 里面是序列化后的 python 字典
- 可以将其转换为 json 可读格式

1.4 使用 Pytorch 运行 .pth 模型

```
img = Image.open(image_path)
# 转换维度 [1, 1, 28, 28]
img_tensor = transform(img).unsqueeze(0) # type: ignore

# 执行推理
with torch.no_grad():
    output = network(img_tensor)
    # 获取概率最大的索引
    prediction = output.data.max(1, keepdim=True)[1]
```

- pth 不存储实际执行顺序
- pth 不存储每一个层该如何执行

想要执行, 必须使用 pytorch, 并构建相同的网络结构 (minst-demo/pytorch_run.py)

2. 导出为 onnx 并使用 onnxruntime 运行

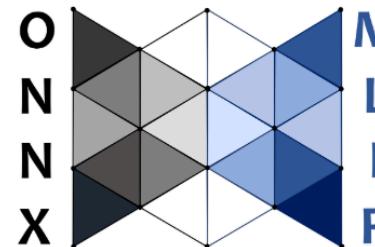
2.1 通用的模型格式

- ONNX 不依赖于具体的深度学习框架
- 基于 protobuf, 支持多种编程语言
- 存储执行顺序
- 定义了算子集, 和算子的行为, 需要由具体的 runtime 实现

任何实现了 ONNX 的运行时均可以运行
ONNX 模型



onnx-mlir



Representation and Reference
Lowering of ONNX Models in
MLIR Compiler Infrastructure

[View the Project on GitHub](#)
[onnx/onnx-mlir](#)

How-Tos

- [Inference Using Python](#)
- [Inference Using C/C++](#)
- [Inference Using Java](#)

2.2 pth 转换为 onnx 并用 onnxruntime 运行

```
network = Net()
network.load_state_dict(torch.load("./model.pth", map_location=torch.device("cpu")))
network.eval()

# 2. 创建一个“伪输入” (Dummy Input)
# ONNX 导出需要通过一次“干跑”来追踪计算图
dummy_input = torch.randn(1, 1, 28, 28)

# 3. 导出模型
torch.onnx.export(
    network, # 要导出的模型
    dummy_input, # 伪输入 # type: ignore
    "model.onnx", # 导出文件名
    do_constant_folding=False, # 是否执行常量折叠优化
    input_names=["input"], # 输入节点名称
    output_names=["output"], # 输出节点名称
)
```

3. ONNX 格式

3.1 protobuf 定义

ONNX 的格式使用 protobuf 定义

<https://github.com/onnx/onnx/blob/main/onnx/onnx.proto>

3.1 protobuf 定义

```
message ModelProto {  
    // The version of the IR this model targets. See Version enum above.  
    // This field MUST be present.  
    optional int64 ir_version = 1;  
  
    // The OperatorSets this model relies on.  
    repeated OperatorSetIdProto opset_import = 8;  
  
    // The name of the framework or tool used to generate this model.  
    optional string producer_name = 2;  
  
    // The version of the framework or tool used to generate this model.  
    optional string producer_version = 3;  
  
    // Domain name of the model.  
    // We use reverse domain names as name space indicators. For example:  
    // `com.facebook.fair` or `com.microsoft.cognitiveservices`  
    optional string domain = 4;  
  
    // The parameterized graph that is evaluated to execute the model.  
    optional GraphProto graph = 7;
```

4. Python 手动构建 onnx 模型

4.1 Python 手动构建 onnx 模型

```
input_tensor_info = helper.make_tensor_value_info(  
    name="X", elem_type=TensorProto.FLOAT, shape=[1, 4]  
)  
  
output_tensor_info = helper.make_tensor_value_info(  
    name="Y", elem_type=TensorProto.FLOAT, shape=[1, 4]  
)  
  
node_def = helper.make_node(  
    op_type="Softmax", inputs=["X"], outputs=["Y"], name="node_softmax_1"  
)  
  
graph_def = helper.make_graph(  
    name="test_softmax_graph",  
    inputs=[input_tensor_info],  
    outputs=[output_tensor_info],  
    nodes=[node_def],  
    initializer=[],  
)  
  
model_def = helper.make_model(  
    graph_def,  
    producer_name="my_human_hand",  
    opset_imports=[helper.make_opsetid(domain="", version=11)],  
    ir_version=11,  
)  
  
onnx.save(model_def, "softmax.onnx")
```

4.2 onnx 对算子的描述

<https://onnx.ai/onnx/operators/>