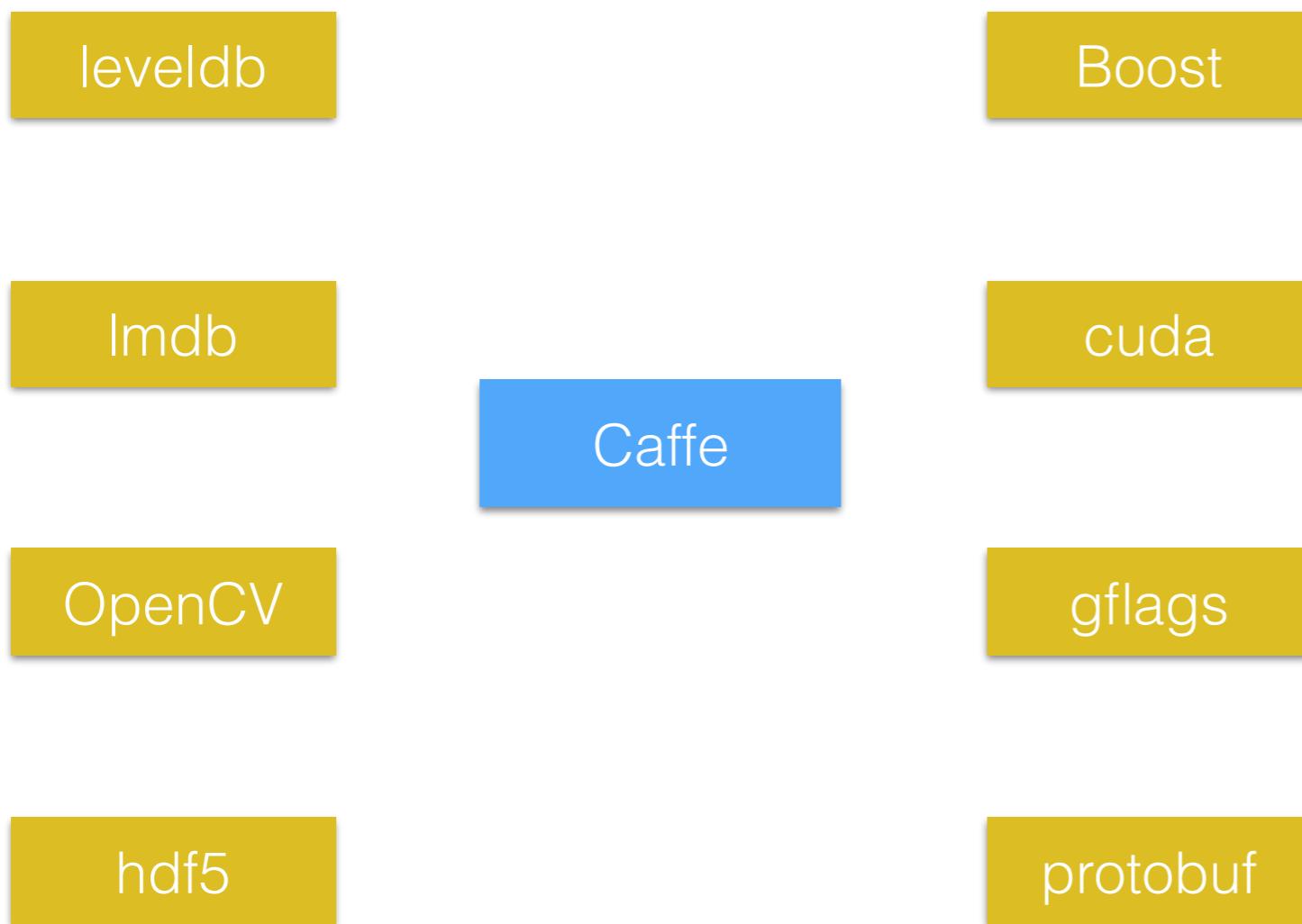


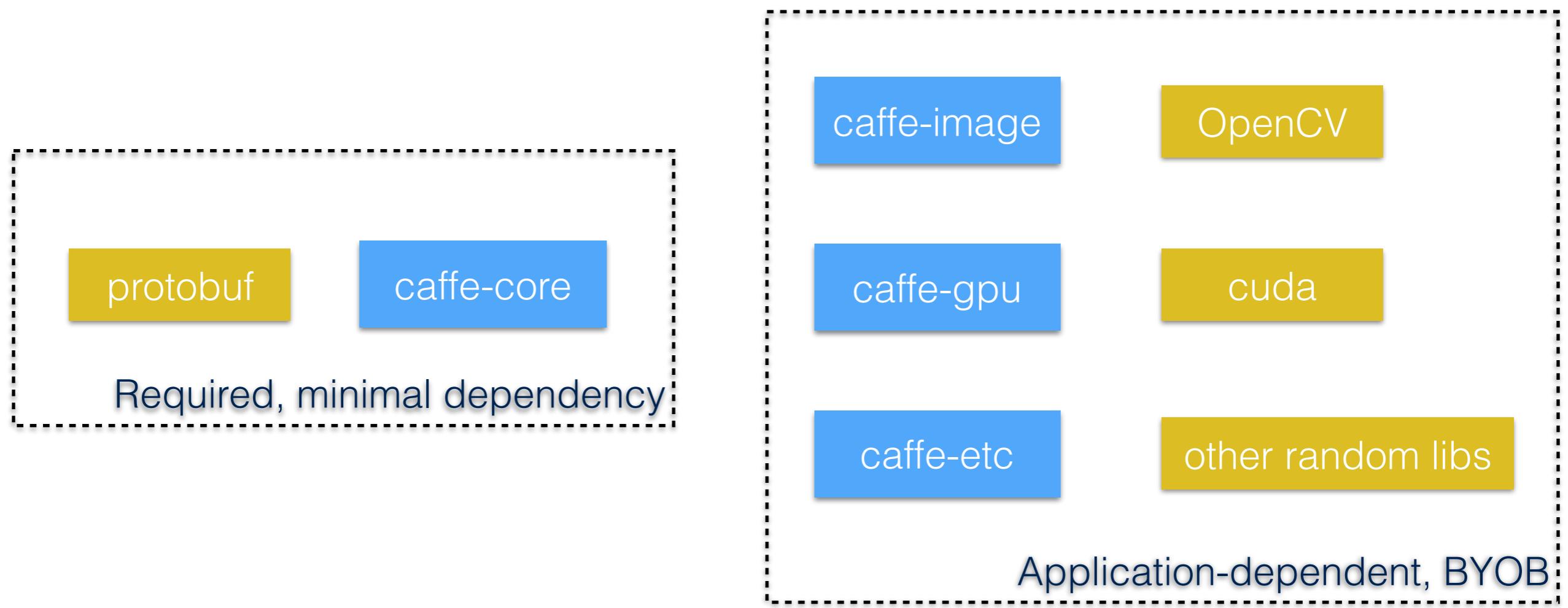
Improving Caffe: Some Refactoring

Yangqing Jia

Dependencies?



Dependency!



Devices

- Current Caffe

```
class Layer {  
    void Forward_cpu();  
    void Forward_gpu();  
    void Backward_cpu();  
    void Backward_gpu();  
}
```

- Simplified Interface

```
template <typename Device>  
class Layer {  
    void Forward(); void Backward()  
}
```

Blob is Anything

- Current Caffe

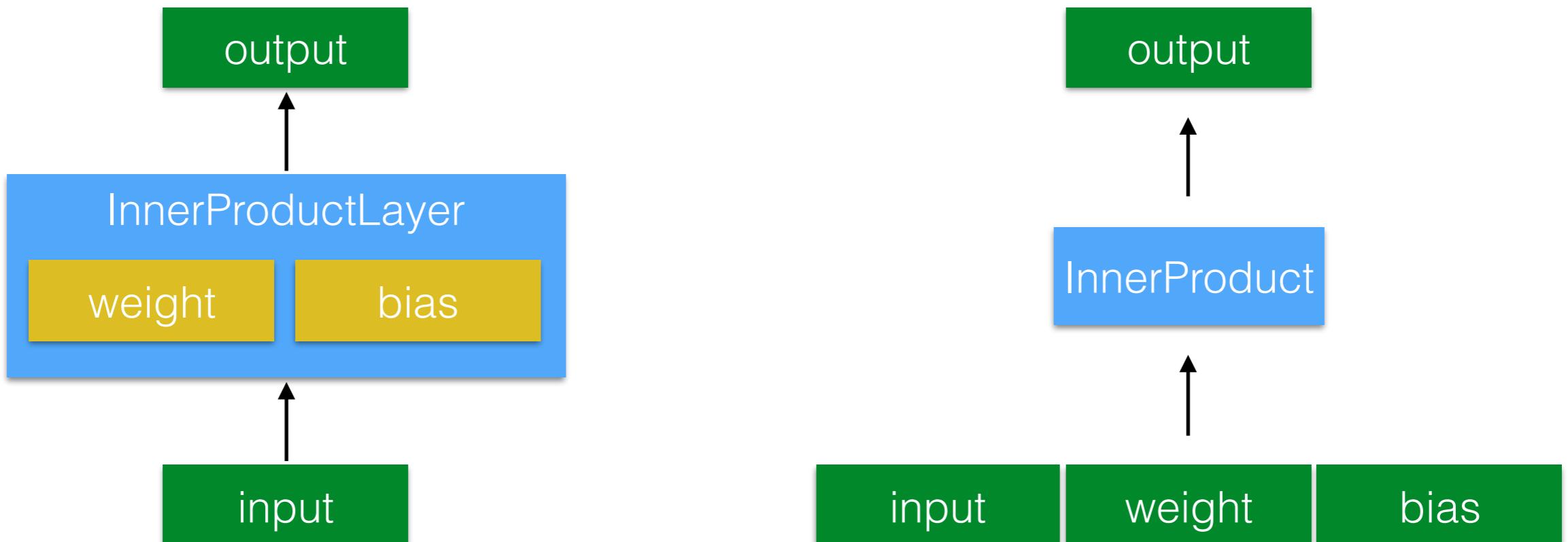
```
template <typename Dtype>
class Blob {
    Dtype* data; ...
};
```
- Simplified Version

```
class Blob {
    AnyPointer data;
    DataType type;
};
```
- Keep the relaxed Blob is N-dimensional generality.
(N-D Blobs in Caffe master)

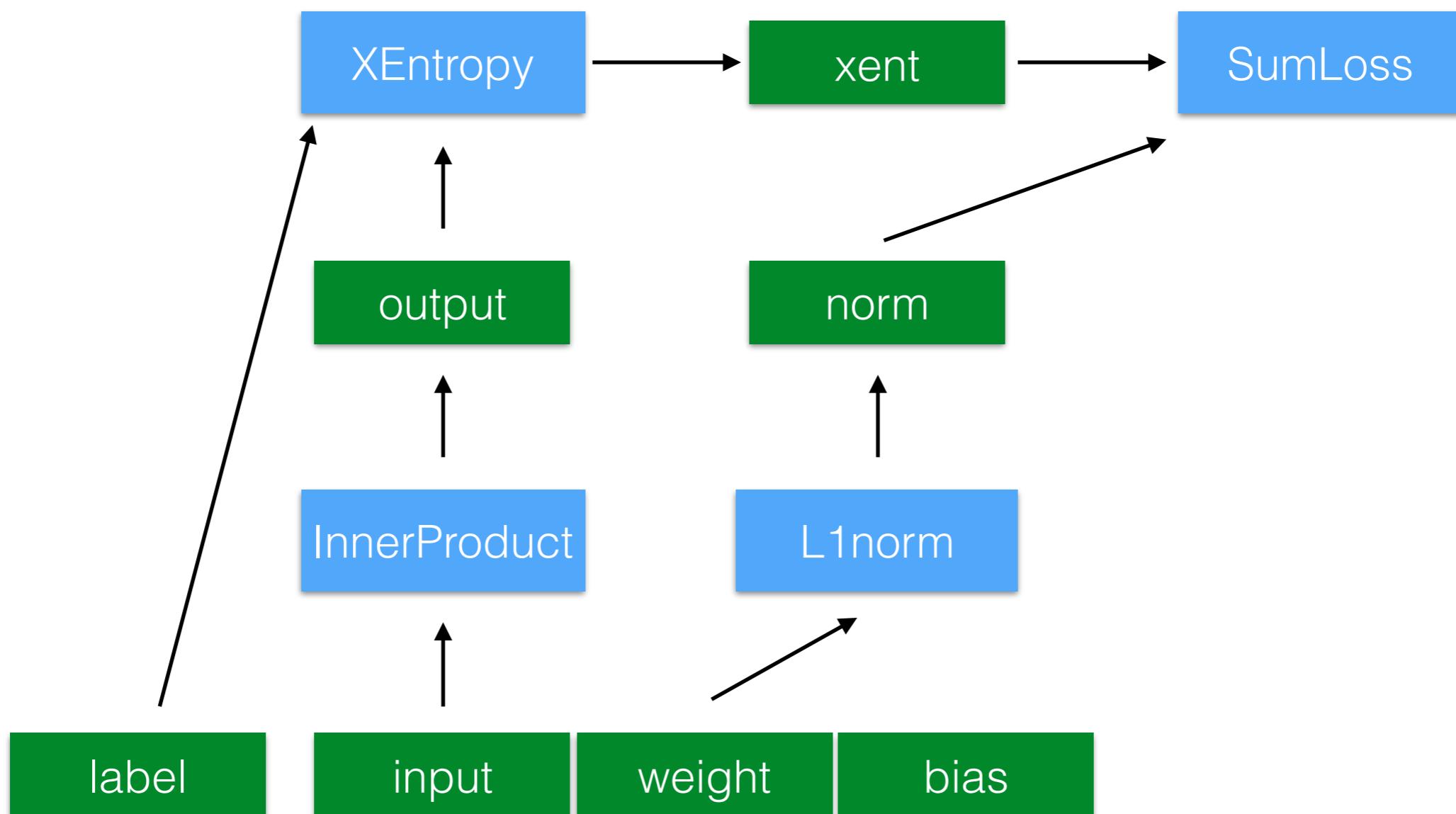
Operators

- Current Caffe
 - class Layer {
 void Forward();
 void Backward();
};
- Simplified Interface
 - class Operator {
 void Run();
};

Which means...



Easier Algorithms



Explicit Gradients

- Current Caffe

```
class Layer {  
    void Forward();  
    void Backward();  
};
```
- I'm playing with...

```
@GradientRegistry.RegisterGradient("Relu")  
def ReluGradient(op):  
    return CreateOperator("ReluGradient")
```

Solvers are Ops Too

```
for param in [filter1, bias1, filter2, bias2]:  
    net.WeightedSum([param, ONE, param.Grad(), LR],  
                    param)
```

Ops for SGD

```
# Add weight decay.  
for param in [filter1, filter2]:  
    net.WeightedSum([param.Grad(), ONE, param, WEIGHT_DECAY],  
                    param.Grad())  
  
# do momentum  
for param in [filter1, bias1, filter2, bias2]:  
    net.Accumulate([param.Grad()], param.Momentum(), gamma=0.9)  
# update  
for param in [filter1, bias1, filter2, bias2]:  
    net.WeightedSum([param, ONE, param.Momentum(), LR], param)
```

Use Cases

- Accumulate gradients
multiple forward and backward + single update
- Multi-GPU?
computation with communication on the side
- Quick research runs
a custom solver, etc.

Purpose...

- Faster Experimenting and Prototyping
- Simpler Compilation
- More Portability