```
1. for iter in range ( a r g s . epochs ) :
     loss locals = [ ] , w locals = [ ]
     m = max(int (args. frac *args. num users), 1)
      idxs users = np . random . choice (range( args . num users ) ,
      m, replace=False ) # random activation number of users
      for id x in idxs users :
           local = LocalUpdate ( args=args , dataset=dataset_
7.
  train ,
8 -
            idxs=dict users [ idx ] )
9.
          w, loss = local . train
                   ( net=copy . deepcopy ( net glob ) . to ( args
10.
  . device ) )
11.
                if args . all clients :
12.
                    w locals [ idx ] = copy . deepcopy (w)
13.
               else :
                     w locals . append ( copy . deepcopy (w) )
14.
15.
                 loss_locals.append ( copy . deepcopy ( loss ) )
16.
        # update global weights
17.
        w 	ext{ glob} = 	ext{FedAvg (} w 	ext{ locals )}
18.
        # copy weight to ne t glob
19.
        net glob . load state dict ( w glob )
20.
21.
22.
        print( Create Model Training )
23.
        print( Waiting for { } clients . . . format( args
  . num users ) )
24.
        while True :
25.
                connection , address = soc . accep t ( )
26.
                 clients . append ( {
27.
                       " connection " : connection ,
                       " address " : address
28.
29.
                  } )
30.
                  print( Connected to : + address [0] + ' : ' +
31.
          str (address [1]))
32.
                  print( Thread Number + str (len( clients ) ) )
33.
                   if (len( clients ) >= args . num_users ) :
34.
                          print( " Client sready " )
35.
                  else :
36.
            continue
37.
                   sleep (3)
38.
39.
        print (FedAVG and Local Update Process )
40.
        print( P rocess Completed )
41.
        soc . close ( )
42.
        print( Socket is closed . )
43.
44.
45.
        print( \ nInitializing weights , please wait . . . \ n )
```

```
46. self . model = ini t model
        self . model path = " models / server / save /model
  . keras "
48.
        self . saveModel ( )
49.
        def loadModel ( self , model path ) :
50.
            self . model = keras
  . models . load model ( model path )
             return self . model
52.
53.
54.
        def compile( self , model ) :
           model . compile( optimizer= ' adam ' ,
55.
              loss=keras . losses . SparseCategoricalCrossentropy
56.
57.
                (from logits=True),
              metrics =[ ' accuracy ' ] )
58.
59.
        return model
60.
61.
        import tensor flow model optimization as tfmot
62.
63.
        print( Initializing weights , please wait . . . )
        if args . quantize :
          init_model , _ = self . train ( init_model )
65.
          self . model = tfmot
 . quantization . keras . quantize model
67.
             ( ini t model )
        self . model path = " models / server / save /model
68.
  . keras "
69.
        self . saveModel ( )
70.
     def saveModel ( self ) :
72.
          if self . args . quantize :
73.
             self . model = self . compile( self . model )
74.
            self . model . save ( self . model path )
          with zipfile . Zip File ( " { } . zip "
  .format( self . model path ) ,
76.
            'w ' , compression= zip file . ZIP DEFLATED) as f :
77.
               f . write ( self . model path )
78.
           return None
79.
80.
        def loadModel ( s e l f , model path ) :
81.
        try :
82.
          with tfmot . quantization . keras . quantize scope ( ) :
             self . model = keras
83.
  . models . load model ( model path )
84.
        return self . model
85.
86.
       def compile( self , model ) :
          model . compile( optimizer= ' adam ' ,
87.
```

```
88.
             loss=keras . losses . SparseCategoricalCrossentropy
89.
             ( from logits=True ) , metrics =[ ' accuracy ' ] )
90.
        return model
91.
92.
        import tensor flow model optimization as tfmot
93.
94.
        print( Initializing weights , please wait . . . )
95.
        if args . prune :
          num images = self . train images . shape [0]*
96.
97.
              (1 - self . args . validation     split )
98.
          end step = np . ceil ( num images
  / self . args . bs ) . astype
              ( np . int32 )*args . local ep
99.
100.
          pruning params = {
101.
           ' pruning schedule ' : tfmot
  . sparsity . keras . PolynomialDecay
102.
            ( initial _ sparsity =0.50 , final _ sparsity =0.80 ,
103.
           begin step =0, end step=end step )
104.
            }
105.
          Self . model = tfmot
   . sparsity . keras . prune low magnitude
106.
              ( ini t_model , **pruning_params )
107.
          callbacks = [
                tfmot . sparsity . keras . UpdatePruningStep ( ) ,
108.
109.
                tfmot . sparsity . keras . PruningSummaries
110.
                     ( log dir= tempfile . mkdtemp ( ) ) ,
111.
                 ]
112.
       self . model , = self . train ( self . model , callbacks )
        self . model_path = " models / server / save /model
  . keras "
114. self . saveModel ( )
```