```python
import numpy as np

"""
This file implements various first-order update rules that are commonly used for
training neural networks. Each update rule accepts current weights and the
gradient of the loss with respect to those weights and produces the next set of
weights. Each update rule has the same interface:

def update(w, dw, config=None):

Inputs:
  - w: A numpy array giving the current weights.
  - dw: A numpy array of the same shape as w giving the gradient of the
    loss with respect to w.
  - config: A dictionary containing hyperparameter values such as learning rate,
    momentum, etc. If the update rule requires caching values over many
    iterations, then config will also hold these cached values.

Returns:
  - next_w: The next point after the update.
  - config: The config dictionary to be passed to the next iteration of the
    update rule.

NOTE: For most update rules, the default learning rate will probably not perform
well; however the default values of the other hyperparameters should work well
for a variety of different problems.

For efficiency, update rules may perform in-place updates, mutating w and
setting next_w equal to w.
"""


def sgd(w, dw, config=None):
  """
  Performs vanilla stochastic gradient descent.

  config format:
  - learning_rate: Scalar learning rate.
  """
  if config is None: config = {}
  config.setdefault('learning_rate', 1e-2)

  w -= config['learning_rate'] * dw
  return w, config


def sgd_momentum(w, dw, config=None):
  """
  Performs stochastic gradient descent with momentum.

  config format:
  - learning_rate: Scalar learning rate.
  - momentum: Scalar between 0 and 1 giving the momentum value.
    Setting momentum = 0 reduces to sgd.
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  - velocity: A numpy array of the same shape as w and dw used to store a moving
    average of the gradients.
  """
  if config is None: config = {}
  config.setdefault('learning_rate', 1e-2)
  config.setdefault('momentum', 0.9) # set momentum to 0.9 if it wasn't there
  v = config.get('velocity', np.zeros_like(w))   # gets velocity, else sets it to zero.

  # ================================================================ #
  # YOUR CODE HERE:
  #   Implement the momentum update formula.  Return the updated weights
  #   as next_w, and the updated velocity as v.
  # ================================================================ #

  alpha = config['momentum']
  eps = config['learning_rate']

  v = alpha * v - eps * dw
  w += v

  next_w = w

  # ================================================================ #
  # END YOUR CODE HERE
  # ================================================================ #

  config['velocity'] = v

  return next_w, config

def sgd_nesterov_momentum(w, dw, config=None):
  """
  Performs stochastic gradient descent with Nesterov momentum.

  config format:
  - learning_rate: Scalar learning rate.
  - momentum: Scalar between 0 and 1 giving the momentum value.
    Setting momentum = 0 reduces to sgd.
  - velocity: A numpy array of the same shape as w and dw used to store a moving
    average of the gradients.
  """
  if config is None: config = {}
  config.setdefault('learning_rate', 1e-2)
  config.setdefault('momentum', 0.9) # set momentum to 0.9 if it wasn't there
  v = config.get('velocity', np.zeros_like(w))   # gets velocity, else sets it to zero.

  # ================================================================ #
  # YOUR CODE HERE:
  #   Implement the momentum update formula.  Return the updated weights
  #   as next_w, and the updated velocity as v.
  # ================================================================ #

  alpha = config['momentum']
  eps = config['learning_rate']

  v_old = v
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111   v = alpha * v - eps * dw
112   w += (v + alpha * (v - v_old))
113
114   next_w = w
115
116   # ================================================================ #
117   # END YOUR CODE HERE
118   # ================================================================ #
119
120   config['velocity'] = v
121
122   return next_w, config
123
124 def rmsprop(w, dw, config=None):
125   """
126   Uses the RMSProp update rule, which uses a moving average of squared
    gradient
127   values to set adaptive per-parameter learning rates.
128
129   config format:
130   - learning_rate: Scalar learning rate.
131   - decay_rate: Scalar between 0 and 1 giving the decay rate for the squared
132     gradient cache.
133   - epsilon: Small scalar used for smoothing to avoid dividing by zero.
134   - beta: Moving average of second moments of gradients.
135   """
136   if config is None: config = {}
137   config.setdefault('learning_rate', 1e-2)
138   config.setdefault('decay_rate', 0.99)
139   config.setdefault('epsilon', 1e-8)
140   config.setdefault('a', np.zeros_like(w))
141
142   next_w = None
143
144   # ================================================================ #
145   # YOUR CODE HERE:
146   #   Implement RMSProp.  Store the next value of w as next_w.  You need
147   #   to also store in config['a'] the moving average of the second
148   #   moment gradients, so they can be used for future gradients. Concretely,
149   #   config['a'] corresponds to "a" in the lecture notes.
150   # ================================================================ #
151
152   a = config['a']
153   beta = config['decay_rate']
154   eps = config['learning_rate']
155   nu = config['epsilon']
156
157   a = beta * a + (1 - beta) * dw * dw
158   w -= (eps * dw) / (np.sqrt(a) + nu)
159
160   config['a'] = a
161   next_w = w
162
163   # ================================================================ #
164   # END YOUR CODE HERE
165   # ================================================================ #
166
167   return next_w, config
168
169
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```python
def adam(w, dw, config=None):
  """
  Uses the Adam update rule, which incorporates moving averages of both the
  gradient and its square and a bias correction term.

  config format:
  - learning_rate: Scalar learning rate.
  - beta1: Decay rate for moving average of first moment of gradient.
  - beta2: Decay rate for moving average of second moment of gradient.
  - epsilon: Small scalar used for smoothing to avoid dividing by zero.
  - m: Moving average of gradient.
  - v: Moving average of squared gradient.
  - t: Iteration number.
  """
  if config is None: config = {}
  config.setdefault('learning_rate', 1e-3)
  config.setdefault('beta1', 0.9)
  config.setdefault('beta2', 0.999)
  config.setdefault('epsilon', 1e-8)
  config.setdefault('v', np.zeros_like(w))
  config.setdefault('a', np.zeros_like(w))
  config.setdefault('t', 0)

  next_w = None

  # ================================================================ #
  # YOUR CODE HERE:
  #   Implement Adam.  Store the next value of w as next_w.  You need
  #   to also store in config['a'] the moving average of the second
  #   moment gradients, and in config['v'] the moving average of the
  #   first moments.  Finally, store in config['t'] the increasing time.
  # ================================================================ #

  t = config['t']
  v = config['v']
  a = config['a']
  eps = config['learning_rate']
  nu = config['epsilon']
  beta1 = config['beta1']
  beta2 = config['beta2']

  t += 1
  v = beta1 * v + (1 - beta1) * dw
  a = beta2 * a + (1 - beta2) * dw * dw
  v_u = v / (1 - beta1**t)
  a_u = a / (1 - beta2**t)
  w -= (eps * v_u) / (np.sqrt(a_u) + nu)

  config['t'] = t
  config['v'] = v
  config['a'] = a
  next_w = w

  # ================================================================ #
  # END YOUR CODE HERE
  # ================================================================ #

  return next_w, config


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