Experiment - Implementation of Transfer Learning

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In [1]:
         import numpy as np
         import matplotlib.pyplot as plt
         from torchvision import transforms
         from torchvision import models
         import torch
         from torch.autograd import Variable
         import torch.nn as nn
         import torch.nn.functional as F
         from torch.optim import lr_scheduler
         from torch import optim
         from torchvision.datasets import ImageFolder
         from torchvision.utils import make_grid
         from torch.utils.data import Dataset,DataLoader
         import time
In [3]: | is_cuda = False
         if torch.cuda.is_available():
             is_cuda = True
In [4]:
         def show(inp) :
             inp = inp.numpy().transpose((1,2,0))
             plt.imshow(inp)
In [6]: transform = transforms.Compose([transforms.Resize((224,224)),
                                         transforms.ToTensor()
         ])
In [7]:
         train = ImageFolder('/kaggle/input/animals141/dataset/dataset',transform)
         train_data_loader = torch.utils.data.DataLoader(train,batch_size=32,num_workers=3,shuff1
In [8]:
         e=True)
In [9]:
         train[0]
         train.classes
In [10]:
In [11]:
         print(train.class_to_idx)
         show(train[50][0])
In [12]:
In [14]:
         vgg = models.vgg16(pretrained = True)
         vgg = vgg.cuda()
In [15]:
In [16]:
         vgg.classifier[6].out_features = 151
         for param in vgg.features.parameters():
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param.requires_grad = False

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def fit(epoch,model,data_loader,phase='training',volatile=False):
In [18]:
             if phase == 'training':
                 model.train()
             if phase == 'validation':
                 model.eval()
                 volatile=True
             running loss = 0.0
             running_correct = 0
             for batch_idx , (data,target) in enumerate(data_loader):
                 if is cuda:
                     data,target = data.cuda(),target.cuda()
                 data , target = Variable(data,volatile),Variable(target)
                 if phase == 'training':
                     optimizer.zero_grad
                 output = model(data)
                 loss = F.cross_entropy(output, target)
                 running_loss += F.cross_entropy(output,target,size_average=False).item()
                 preds = output.data.max(dim=1,keepdim=True)[1]
                 running_correct += preds.eq(target.data.view_as(preds)).cpu().sum()
                 if phase == 'training':
                     loss.backward()
                     optimizer.step()
             loss = running loss/len(data loader.dataset)
             accuracy = 100. * running_correct/len(data_loader.dataset)
             print(f'{phase} loss is {loss:{5}.{2}} and {phase} accuracy is {running_correct}/{le
         n(data_loader.dataset)}{accuracy:{10}.{4}}')
             return loss,accuracy
         train_losses , train_accuracy = [],[]
In [19]:
         val_losses , val_accuracy = [],[]
         for epoch in range(1,15):
             epoch_loss, epoch_accuracy = fit(epoch,vgg,train_data_loader,phase='training')
             train_losses.append(epoch_loss)
             train_accuracy.append(epoch_accuracy)
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In [17]: optimizer = optim.SGD(vgg.classifier.parameters(),lr=0.01)