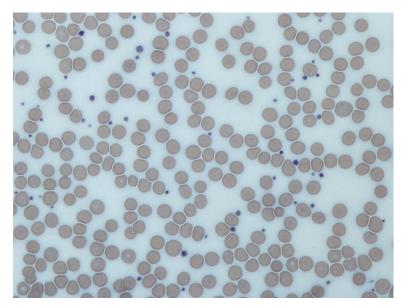
# **CSET340- Advanced Computer Vision and Video analytics**

# Task-1:- Blob detection and Image Quality Enhancement.

**Blob detection**:- Blob detection is a basic method in computer vision used to locate areas of interest in a picture. Apply the **three** different blob detection techniques (LoG, DoG, HoG) **separately** on the same image.



**Image:-** Image from a <u>light microscope</u> (500 ×) showing platelets (small **purple** dots) surrounded by <u>red blood cells</u> (large **gray** circular structures)

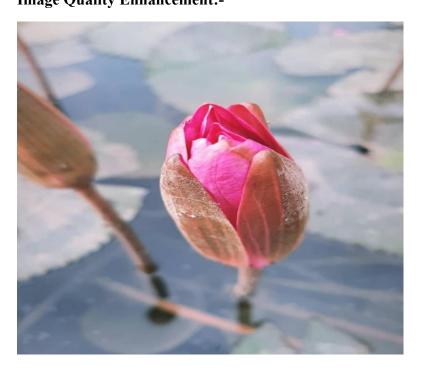


Image:- Candy image. How many candies are present? (Optional task)





Images:- Satellite images
Image Quality Enhancement:-



- Perform this using the following techniques
  - o Adjusting brightness and contrast
  - Sharpening images
  - o Removing noise from images
  - Enhancing color in images
  - o Image resizing and scaling
  - o Inverse Transform
  - Equalizing histograms
  - Super-resolution
  - o Color correction

Task-2:- Image Classification using AlexNet and VGG16 on Cifar-100 Objective:

- Compare the performance of AlexNet and VGG16 on an image classification task.
- Use CIFAR-100, a common dataset for image classification.
- Analyse model accuracy, loss, and inference time on a dataset.
- Step 1: Installation of necessary libraries
- Step 2: Load the dataset.
- Step 3: Load the pretrained Model (AlexNet and VGG16)
- Step 4: Train the Models
- Step 5: Evaluate the Performances
- Step 6: Compare the results.

#### **Datasets-**

1. CIFAR-100 dataset:- The CIFAR100 (Canadian Institute For Advanced Research) dataset consists of 100 classes with 600 color images of 32×32 resolution for each class.

This dataset is just like the CIFAR-10, except it has 100 classes containing 600 images each. There are 500 training images and 100 testing images per class. The 100 classes in the CIFAR-100 are grouped into 20 super classes. Each image comes with a "fine" label (the class to which it belongs) and a "coarse" label (the superclass to which it belongs).



- 2. Load CIFAR 100 Dataset Training Subset in Python
  - a. import deeplake
  - b. ds = deeplake.load("hub://activeloop/cifar100-train")
- 3. Load CIFAR 100 Dataset Testing Subset in Python

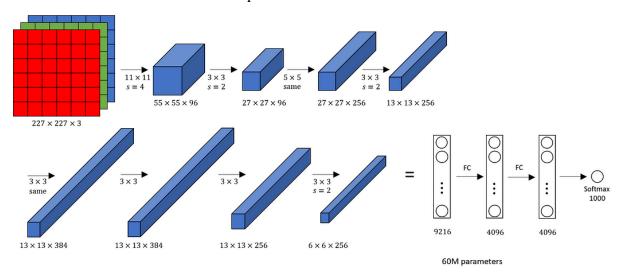
- a. import deeplake
- b. ds = deeplake.load("hub://activeloop/cifar100-test")
- 4. CIFAR 100 Dataset Structure

### a. CIFAR 100 Data Fields

- i. images: tensor containing images of the dataset.
- ii. labels: tensor containing labels for their respective image.
- iii. coarse labels: tensor containing superclass for their respective image.

#### Alexnet:-

- AlexNet consists of 5 convolution layers, 3 max-pooling layers, 2 Normalized layers, 2 fully connected layers and 1 SoftMax layer.
- Each convolution layer consists of a convolution filter and a non-linear activation function called "ReLU".
- The pooling layers are used to perform **the max-pooling** function and the input size is fixed due to the presence of fully connected layers.
- The input size is mentioned at most of the places as 224x224x3 but due to some padding which happens it works out to be 227x227x3.
- AlexNet has over 60 million parameters.



## VGG-16:-

- The 16 in VGG16 refers to 16 layers that have weights.
- In VGG16 there are 13 convolutional layers, 5 Max Pooling layers, and 3 Dense layers which sum up to 21 layers but it has only sixteen weight layers i.e., learnable parameters layer.
- VGG16 takes input tensor size as 224, 244 with 3 RGB channel
- Convolution layers of 3x3 filter with stride 1 and always used the same padding and maxpool layer of 2x2 filter of stride 2.
- Conv-1 Layer has 64 number of filters, Conv-2 has 128 filters, Conv-3 has 256 filters, Conv 4 and Conv 5 has 512 filters.