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# Generation of RBC, WBC Subtype, and Platelet Count Report Using YOLO

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Abstract: Artificial intelligence introduced a way to combine machines' computing ability with human intelligence. Machine learning is a sub-branch of AI that consists of different algorithms to implement concepts of AI in practical terms. But when a machine has to process numerous amounts of data, deep learning algorithms come into the picture. It is observed that when a computing system has to deal with image data then neural network algorithms give efficient ways to process them and draw unique patterns from them. Object detection is the task of identifying required objects from an image. This type of technology plays a crucial role in medical image processing. Some algorithms can efficiently identify and classify objects from an image. It is observed that Fast R-CNN and Faster R-CNN, mask R-CNN, have given pretty good accuracy while performing such tasks. But when time is a concern for a system, such methods put an obstacle of training time and architectural complexity. The You Only Look Once (YOLO) object detection method has introduced a new way of processing images in a single pass. This algorithm is famous because of its speed and correctness. There are numerous models of YOLO developed now which include YOLOv1 to YOLOV8. This paper gives the performance of the latest version of YOLO on the blood cell dataset. Keywords: YOLO, CBC, YOLOv5, YOLOv7, YOLOv8, Deep Learning

# I. INTRODUCTION

Blood is a fluid connective tissue. It is an important parameter of a patient's health check-up. Human blood is composed of 55% of liquid and 45% of solid components. The liquid part is called plasma which consists of proteins, dissolved salts, and nutrients whereas the solid part consists of cellular structures which include White Blood Cells, Red Blood Cells Platelets, etc. White blood cells are also called attacking cells as they invade unwanted or foreign bodies that entered the bloodstream. Several types of WBCs attack differently on foreign bodies. Their count helps doctors to understand the overall disease spread in the body. RBCs play an important role in supplying oxygen to all body organs. Platelets are also known as thrombocytes which generate clots and stop bleeding. They are important to surviving and fighting cancer, chronic diseases, and traumatic injuries.



In today's busy lifestyle people are quite unaware of their health and it causes serious health problems. Most of the time many diseases are concerned with blood, so doctors often ask patients to do their blood tests. In this test, we get the report which consists of the exact count of blood cells present in the blood smear. This count helps physicians in diagnosing diseases, deciding therapy plans, and predicting the stage of a particular disease like cancer. To perform these kinds of tests generally patients visit pathological labs for blood tests, where nurses or medical specialists take a drop of the blood, and then pathologists analyze and perform counting of blood cells by placing slides of blood under a microscope. This process is quite time-consuming and needs 4 to 5 hours for final report generation as pathologists manually perform counting. So, to make this process easier and more efficient we have developed a model which will take the image of a blood smear, process it automatically and give the report of blood cell count in just a few minutes. There are several models which worked on this idea but their architectural and implementational complexity made it time-consuming and tedious. Previously developed models have worked individually on specialized blood cells.



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But in this paper, we are introducing a way of detecting and counting all WBCs subtypes along with platelets and RBCs using the latest version of the You Only Look Once, Model

# II. METHODOLOGY

This section will give a detailed explanation of the flow of the system and its implementation



Let us now understand each section in detail:

#### A. Front End

This is nothing but a user interface that hides the complexity of the model. This is implemented in the python flask framework. This section will also accept peripheral blood smear images in the intended format.

The image should be in JPG, JPEG, and PNG format. In this system, the user is nothing but a lab technician. This user interface will accept some more details like the patient's full name, reference doctor name, age gender, etc. Input from this stage will serve as input to the YOLO framework.

#### B. YOLO Model

You Only Look Once (YOLO) is one of the widely used computer vision models. The YOLO family of models has continued to evolve since its initial release in 2016. The latest version of YOLO is version 8. The original YOLO model was the first object detection network to combine the problem of drawing bounding boxes and identifying class labels in one end-to-end differentiable network. YOLO is a single-stage detector, handling both object identification and classification in a single pass of the network. YOLO is not the only single-stage detection model but it is generally more preferment in terms of speed and accuracy. Before YOLO, R-CNNs were among the most commonly used method for detecting objects, but when accuracy and speed is a things of concern then YOLO is the winner of this competition. Each updated version of YOLO has some unique improvements to the previous one. This upgradation may be in terms of the backend neural network or way of training the model. Let us now have a look at different versions of YOLO models

#### 1) YOLOv1

It was first established in the year 2015 by Joseph Redmon et al. Its main characteristic was its Single stage unified detection. It was the first single-stage object detector approach that treated detection as a regression problem. The detection architecture only looked once at the image to predict the location of the objects and their class labels. it uses a single neural network that predicts class probabilities and bounding box coordinates from an entire image in one pass. YOLOv1 has achieved 63.4 mAP



#### 2) YOLOv2

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It was established in the year 2016 by Joseph Redmon and Ali Farhadi. YOLOv2 was trained on detection datasets like Pascal VOC and MS COCO. YOLOv2 has achieved 76.8 mAP. Its characteristic was the introduction of anchor boxes and multi-scale training.

## 3) YOLOv3

It was established in the year 2018 by Joseph Redmon and Ali Farhadi. They introduced a new network architecture called Darknet-53. This was a more efficient and fast neural network architecture. YOLOv3 has achieved 28.2 mAP. Its characteristic was a feature pyramid network (FPN)

### 4) YOLOv4

It was established in the year 2022 by Meituan. This kind of model consists of a backbone, neck, and head. The backbone can be a pre-trained convolutional neural network such as VGG16 or CSPDarkNet53 trained on COCO or ImageNet data sets. The backbone of the YOLO v4 network acts as the feature extraction network that computes feature maps from the input images. The neck connects the backbone and the head. The neck concatenates the feature maps from different layers of the backbone network and sends them as inputs to the head. The head processes the aggregated features and predicts the bounding boxes, objectness scores, and classification scores.

#### 5) YOLOv5

It was established in the year 2020 by Ultralytics. YOLOv5 offers a family of object detection architectures pre-trained on the MS COCO dataset. YOLOv5 was implemented in the PyTorch environment which eliminated the Darknet framework's limitations. YOLOv5 was one of the official state-of-the-art models in the Torch Hub. There were different sub-models of YOLOv5 like YOLOv51, YOLOv5s, YOLOv5x, and YOLOv5n which can be used depending on the size of the dataset.

#### 6) YOLOv6

It was established in the year 2022 by Meituan. YOLO v6 differed from YOLOV5 in terms of the CNN architecture used. YOLO v6 used a variant of the EfficientNet architecture called EfficientNet-L2. In YOLO v6 new methods of anchor boxes called Dense Anchor boxes were introduced.

#### 7) YOLOv7

It was introduced in the year 2022 by Chien-Yao Wang et all. Improvement in YOLO v7 was the use of an efficient loss function called focal loss. It has been observed that higher resolution than the previous version. YOLO v7 was efficient because of its speed as it can process images at a rate of 155 frames per second

#### 8) YOLOv8

This model is the latest version of YOL introduced in the year 2023 by ultralytics. Its characteristic is that the head portion of the model is free from anchor boxes. It is observed that it takes less time for training. One of the remarkable changes in this version is that while training if the model founds that there is no change in the weights then early stopping is carried out by the system itself. This is the key benefit of using

YOLOv8. Such improvement helps in preventing the model from overfitting.

The YOLOv8-based system can be broken down into major four steps:

- *a)* Creating and annotating the dataset
- b) Model Training
- c) Saving the best weights
- d) Validation and deployment

Let us understand these steps in brief:

• Creating and annotating dataset-

In this, we first created a dataset from a set of images by adding some noise and rotating the images to different angles. As initially we have a lesser number of images to use as the dataset, we have done this step.



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Images from Dataset



Monocyte Neutrophil

To ease the entire process of training and understanding and optimizing the performance we have created annotations in YOLO format only instead of going with the standard XML format. In the case of XML or CSV format, it first needs to be converted into a YOLO-like structure which was quite a tedious process. There is a library called LabelIMG in Python itself to ease the process of the labeling image dataset. But to save time we have used an online available tool for image annotation tasks.

YOLO Data Annotation format is:

<object-class> <x> <y> <width> <height

Here object classes are referred to as classes that are needed to be identified e.g. dog, cat, etc.

x is referred to as the center x-coordinate of the bounding box. It is calculated as actual\_center\_x\_coordinate\_of\_bounding\_box /width\_ of\_the\_image.

y is referred to as the center y-coordinate of the bounding box. It is calculated as actual\_center\_y\_coordinate\_of\_bounding\_box /height\_ of\_the\_image.

width is referred as width\_of\_the\_bounding\_box/width\_of\_the\_image height is referred as height\_of\_the\_bounding\_box/height\_of\_the\_image



360PX

SO HERE

x=200/360 y=250/363 width=45/360 height=35/363

#### • Model Training

There are two modern techniques are introduced to improve object detection and classification: Mosaic data augmentation and Class-specific anchor boxes. Mosaic data augmentation is a data augmentation technique that mixes up images to create new training images. This ultimately helps to increase the diversity of the training data resulting in the prevention of overfitting and improving generalization. Mosaic data augmentation combines 4 training images into one in random proportions. The algorithms are the following:



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- ➤ Take 4 images from the train set;
- Resize them to the same size;
- Integrate the images into a 4x4 grid;
- > Crop a random image patch from the center.

Class-specific anchor boxes are anchor boxes that are specific to each class, allowing the model to better match the shape and aspect ratio of objects in the target class. In YOLO v8 new model architecture called SPP-YOLO Spatial Pyramid Pooling YOLO is introduced which helps to handle objects at different scales.

#### • Saving Best Weights

After training the pre-trained YOLO model for 93 epochs weights are automatically get saved in best.pt file.

#### • Validation And Deployment

In YOLOV8 we have performed validation using images saved in the val directory. For that model .val() method have used. Here we have used best weights for further classification.

• User Interface and Final Result

Age Sex 66 Male
66 Male
Referenece Doctor Date
Dr.Bardapurkar 05-04-2023
Submit
ME: Meena Mohit Malpure AGE: 66 SEX: M.
ME: Meena Mohit Malpure AGE: 66 SEX: M SFFERED BY: Dr.Bardapurkar DATE: 2023-04-05
ME: Meena Mohit Malpure AGE: 66 SEX: M. EFFERED BV: Dr.Bardapurkar DATE: 2023-04-05 COMPONENT COUNT
AGE: 66 DATE: 2023-04-05 SEX: M. DATE: 2023-04-05   COMPONENT COUNT   TOTAL WBC : 1   Eosinophil : 0
ME: Meena Mohit Malpure FFERED BY: Dr.Bardapurkar COMPONENT TOTAL WBC : 1 Eosinophil : 0 Neutrophil : 0
ME: Meena Mohit Malpure EFFERED BY: Dr.Bardapurkar AGE: 66 DATE: 2023-04-05 SEX: M. DATE: 2023-04-05   COMPONENT COUNT   TOTAL WBC : 1   Eosinophil : 0   Neutrophil : 0   Basophil : 1
ME: Meena Mohit Malpure FFERED BY: Dr.Bardapurkar COMPONENT TOTAL WBC : 1 Eosinophil : 0 Neutrophil : 1 Lymphocyte : 0
ME: Meena Mohit Malpure EFFERED BY: Dr.Bardapurkar AGE: 66 DATE: 2023-04-05 SEX: M DATE: 2023-04-05   COMPONENT COUNT   TOTAL WBC : 1   Eosinophil : 0   Neutrophil : 0   Basophil : 1   Lymphocyte : 0
ME: Meena Mohit Malpure EFFERED BY: Dr.Bardapurkar AGE: 66 DATE: 2023-04-05 SEX: M DATE: 2023-04-05   COMPONENT COUNT   TOTAL WBC : 1   Eosinophil : 0   Neutrophil : 0   Basophil : 1   Lymphocyte : 0   Monocyte : 0



Comparison of YOLOv5, YOLOv7 and YOLOv8 Comparison based on training time and accuracy

- Training time for 250 epochs
- YOLOv5: 0.915 hours, It Stopped training early because no improvement was observed in the last 100 epochs. Hence although we have run it for 250 epochs, it completed the training at 171 epochs.

	Epoch	GPU mem	box loss	obj loss	cls loss	Instances	
	169/249	4.66G	0.01765	0.04955	0.0006861	287	
F→		Class	Images	Instances	P	R	n
		all	280	3084	0.831	0.904	e
	Epoch	GPU_mem	box_loss	obj_loss	cls_loss	Instances	
	170/249	4.66G	0.01687	0.04925	0.0006639	204	
		Class	Images	Instances	P	R	m
		all	280	3084	0.848	0.909	6
	171 epochs co Optimizer str Optimizer str	mpleted in ipped from ipped from	0.915 hour runs/train runs/train	s. /exp8/weigh /exp8/weigh	ts/last.pt, ts/best.pt,	14.5MB 14.5MB	
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	171 epochs co Optimizer str Optimizer str Validating ru Fusing layers Model summary e 1	<pre>mpleted in ipped from ipped from ms/train/ex :: 157 layer Class all basophil osinophil ymphocyte</pre>	0.915 hour runs/train runs/train cp8/weights cs, 7029004 Images 280 280 280 280 280	s. /exp8/weigh /exp8/weigh /best.pt parameters Instances 3084 40 40	ts/last.pt, ts/best.pt, , 0 gradien P 0.914 1 0.99 1	14.5MB 14.5MB ts, 15.8 GFLC R 0.904 0.83 1 0.834	DPs n e e e
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> *YOLOv7*: It took 2.732 hours for training the model.

+	- Code	+ Text						
		241/242	12.20	0.01401	0.01023 0.000	cocci	0.00020	512
~	O		Class	Images	Labels		P	R
2h	_		all	280	3084		0.88	0.91
	C≁	_						
		Epoch	gpu_mem	box	obj	cls	total	labels
		248/249	12.2G	0.01485	0.01741 0.000	3277	0.03259	290
			Class	Images	Labels		P	R
			all	280	3084		0.857	0.937
		Epoch	gpu mem	box	obj	cls	total	labels
		249/249	12.2G	0.01487	0.01787 0.000	3446	0.03308	221
			Class	Images	Labels		Р	R
			all	280	3084		0.865	0.922
			basophil	280	40		0.996	1
		e	osinophil	280	40		0.848	1
		1	vmphocvte	280	40		0.928	0.875
		-	monocyte	280	40		0.565	0.9
		n	eutrophil	280	40		0.824	1
			nlatelet	200	234		0.053	0 053
			pracerec	200	254		0.955	0.335
		250 epochs	completed	in 2.732 ho	urs.		0.942	0.725
		Optimizer s	tripped fr	om runs/tra:	in/exp/weights	/last	.pt, 74.8M	В
	(	Optimizer s	tripped fr	om runs/tra	in/exp/weights	/best	.pt, 74.8M	В
		4						

> *YOLOv8*: It took 2.699 hours for training the model.

	241/249	12.00	0.0140	0.0100/	0.000504	0.00000	212
O		Class	Images	Lab	els	P	R
-		all	280	з	084	0.885	0.912
C⇒							
	Epoch	gpu_mem	box	obj	CIS	total	labels
	248/249	12.3G	0.01479	0.01721	0.000344	0.03235	290
		Class	Images	Lab	els	P	R
		all	280	З	084	0.881	0.918
	Epoch	gpu mem	box	obj	cls	total	labels
	249/249	12.3G	0.01484	0.01753	0.0003477	0.03272	221
		Class	Images	Lab	els	P	R
		all	280	з	084	0.879	0.921
		basophil	280		40	0.997	1
	e	osinophil	280		40	0.924	1
	1	vmphocyte	280		40	b.897	0.875
		monocyte	280		40	0.585	0.9
	n	eutrophil	280		40	0.854	0.975
		nlatelet	280		234	0.961	0.948
		rhc	280	2	650	0.934	0 747
	250 epochs	completed i	n 2.699 ho	urs.			
	Ontimizon c	tripped fro	m nunc/tra	in/ovn/wa	ights/lass	t nt 74 9M	P
	Optimizer s	tripped fro	m runs/tra.	in/exp/we	ights/las	L.pt, 74.0M	



From this, we can observe that as we go with the higher versions of YOLO the training time decreases.

- Accuracy for 250 epochs
- > YOLOV5: From the confusion matrix we have calculated its accuracy as 98.01% at 171 epoch early stop.



> YOLOV7: From the confusion matrix we have calculated its accuracy as 98.96%.



> *YOLOV8:* From the confusion matrix, we have calculated its accuracy as 98.52%.





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#### III. CONCLUSIONS

This paper gives us brief views of the YOLO algorithm and its different versions. Here we have proposed a model for generating the blood cell count by using the YOLOv8 version which is the fastest and latest version of the YOLO algorithm. Here we have mentioned the step-wise description of each process of the model. The project is divided into two parts front end and the YOLO algorithm model. Here we have discussed the results of the YOLOv5, YOLOv7, and YOLOv8 versions which are compatible with the torch framework of Python. There is still room for future improvement in terms of the speed and accuracy of the YOLO model.

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