


YOLO MODEL PERFORMANCE METRICS – combined dataset

Yolov8s

val: Fast image access  (ping: 2.1±3.2 ms, read: 213.6±186.8 MB/s, size: 1996.2 KB)

val: Scanning /content/drive/MyDrive/Colab

Notebooks/combined_dataset/val/labels.cache... 82 images, 2 backgrounds, 0 corrupt:

100%

 82/82 91.2Kit/s 0.0s

	Class	Images	Instances	Box (P)	R	mAP50
mAP50-95): 100%			3/3 0.5it/s	6.1s		
0.594	all	82	253	0.886	0.81	0.845
0.726	books	9	153	0.992	0.841	0.973
0.891	monitor	46	50	0.902	0.923	0.936
0.565	office-chair	23	36	0.907	0.667	0.774
0.536	whiteboard	5	5	0.971	1	0.995
0.196	table	6	7	0.619	0.429	0.393
0.652	tv	1	2	0.926	1	0.995

Speed: 0.4ms preprocess, 10.5ms inference, 0.0ms loss, 2.5ms postprocess per image

Results saved to **/content/runs/detect/val**

```
{'mAP50-95': np.float64(0.5943079250671102), 'mAP50':  
np.float64(0.8445052179463065), 'mAP75': np.float64(0.6452724627417927),  
'per_class_mAP': array([ 0.59431, 0.72618, 0.89145, 0.565,  
0.53572, 0.1959, 0.6516]), 'speed_ms': {'preprocess': 0.4079971463453126,  
'inference': 10.492500768289506, 'loss': 0.0003981219501062098, 'postprocess':  
2.4746130853744526}}
```

book: AP50-95=0.594

books: AP50-95=0.726

monitor: AP50-95=0.891


office-chair: AP50-95=0.565

whiteboard: AP50-95=0.536

table: AP50-95=0.196

tv: AP50-95=0.652

Yolov8s – heavy augmentation - train

val: Fast image access  (ping: 0.6±0.3 ms, read: 296.7±250.1 MB/s, size: 1996.2 KB)

val: Scanning /content/drive/MyDrive/Colab

Notebooks/combined_dataset/val/labels.cache... 82 images, 2 backgrounds, 0 corrupt:

100%

 82/82 122.9Kit/s 0.0s

	Class	Images	Instances	Box (P)	R	mAP50
mAP50-95): 100%			3/3 0.5it/s	6.0s		
0.597	all	82	253	0.858	0.791	0.869
0.64	books	9	153	0.959	0.92	0.967
0.842	monitor	46	50	0.911	0.9	0.945
0.525	office-chair	23	36	0.855	0.639	0.817
0.587	whiteboard	5	5	0.984	1	0.995

0.292	table	6	7	0.569	0.286	0.498
0.697	tv	1	2	0.871	1	0.995

Speed: 0.2ms preprocess, 10.0ms inference, 0.0ms loss, 4.0ms postprocess per image
 Results saved to **/content/runs/detect/val**
 {'mAP50-95': np.float64(0.597150691409977), 'mAP50':
 np.float64(0.8693538993610099), 'mAP75': np.float64(0.6690936377989775),
 'per_class_mAP': array([0.59715, 0.63977, 0.84172, 0.52469,
 0.58747, 0.2918, 0.69746]), 'speed_ms': {'preprocess':
 0.20990359755670374, 'inference': 10.042173817071937, 'loss':
 0.0007077317094648961, 'postprocess': 4.037841219514586}}

book: AP50-95=0.597
 books: AP50-95=0.640
 monitor: AP50-95=0.842
 office-chair: AP50-95=0.525
 whiteboard: AP50-95=0.587
 table: AP50-95=0.292
 tv: AP50-95=0.697

Yolo11n – train2

Ultralytics 8.3.199 🚀 Python-3.12.11 torch-2.8.0+cu126 CUDA:0 (Tesla T4, 15095MiB)
 YOLO11n summary (fused): 100 layers, 2,583,517 parameters, 0 gradients, 6.3 GFLOPs
val: Fast image access ✅ (ping: 0.3±0.1 ms, read: 238.1±216.1 MB/s, size: 1996.2 KB)

val: Scanning /content/drive/MyDrive/Colab
 Notebooks/combined_dataset/val/labels.cache... 82 images, 2 backgrounds, 0 corrupt:

100%	82/82 102.8Kit/s 0.0s					
	Class	Images	Instances	Box (P)	R	mAP50
mAP50-95): 100%			3/3 0.5it/s	5.7s		
0.591	all	82	253	0.869	0.8	0.869
0.695	books	9	153	0.992	0.813	0.964
0.866	monitor	46	50	0.957	0.889	0.932
0.559	office-chair	23	36	0.864	0.667	0.793
0.514	whiteboard	5	5	0.776	1	0.928
0.363	table	6	7	0.675	0.429	0.6
0.548	tv	1	2	0.951	1	0.995

Speed: 3.9ms preprocess, 6.5ms inference, 0.0ms loss, 1.9ms postprocess per image
 Results saved to **/content/runs/detect/val2**
 {'mAP50-95': np.float64(0.5906702983796589), 'mAP50':
 np.float64(0.8687523650864039), 'mAP75': np.float64(0.5741715201925097),
 'per_class_mAP': array([0.59067, 0.69502, 0.86559, 0.55877,
 0.51412, 0.3626, 0.54793]), 'speed_ms': {'preprocess': 3.868378317070981,
 'inference': 6.462925670714357, 'loss': 0.0005391707307397864, 'postprocess':
 1.9331505731670626}}

book: AP50-95=0.591
 books: AP50-95=0.695
 monitor: AP50-95=0.866
 office-chair: AP50-95=0.559
 whiteboard: AP50-95=0.514
 table: AP50-95=0.363
 tv: AP50-95=0.548

Yolov8n – train3

Ultralytics 8.3.199 🚀 Python-3.12.11 torch-2.8.0+cu126 CUDA:0 (Tesla T4, 15095MiB)
Model summary (fused): 72 layers, 3,007,013 parameters, 0 gradients, 8.1 GFLOPs
val: Fast image access ✅ (ping: 0.4±0.2 ms, read: 114.6±95.1 MB/s, size: 1996.2 KB)
val: Scanning /content/drive/MyDrive/Colab
Notebooks/combined_dataset/val/labels.cache... 82 images, 2 backgrounds, 0 corrupt:
100% ————— 82/82 162.8Kit/s 0.0s

	Class	Images	Instances	Box(P	R	mAP50
mAP50-95): 100%			3/3 0.6it/s	4.7s		
0.636	all	82	253	0.871	0.836	0.862
0.712	books	9	153	0.985	0.852	0.973
0.866	monitor	46	50	0.929	0.88	0.934
0.519	office-chair	23	36	0.866	0.715	0.774
0.72	whiteboard	5	5	0.946	1	0.995
0.353	table	6	7	0.556	0.571	0.502
0.648	tv	1	2	0.943	1	0.995

Speed: 0.4ms preprocess, 4.6ms inference, 0.0ms loss, 2.2ms postprocess per image
Results saved to **/content/runs/detect/val3**
{'mAP50-95': np.float64(0.6362101342658398), 'mAP50': np.float64(0.8623412351540852), 'mAP75': np.float64(0.7399145309415824), 'per_class_mAP': array([0.63621, 0.71206, 0.86567, 0.51872, 0.71984, 0.35313, 0.64784]), 'speed_ms': {'preprocess': 0.3985970853669065, 'inference': 4.621975573183768, 'loss': 0.0004041219476564992, 'postprocess': 2.154720914637526}}

book: AP50-95=0.636
books: AP50-95=0.712
monitor: AP50-95=0.866
office-chair: AP50-95=0.519
whiteboard: AP50-95=0.720
table: AP50-95=0.353
tv: AP50-95=0.648

Yolov5nu – train4

Ultralytics 8.3.199 🚀 Python-3.12.11 torch-2.8.0+cu126 CUDA:0 (Tesla T4, 15095MiB)
YOLOv5n summary (fused): 84 layers, 2,504,309 parameters, 0 gradients, 7.1 GFLOPs
val: Fast image access ✅ (ping: 0.4±0.1 ms, read: 104.5±89.4 MB/s, size: 1996.2 KB)
val: Scanning /content/drive/MyDrive/Colab
Notebooks/combined_dataset/val/labels.cache... 82 images, 2 backgrounds, 0 corrupt:
100% ————— 82/82 155.1Kit/s 0.0s

	Class	Images	Instances	Box(P	R	mAP50	mAP50-
95): 100%			3/3 0.5it/s	6.1s			
0.601	all	82	253	0.839	0.764	0.854	
0.663	books	9	153	0.89	0.908	0.944	
0.848	monitor	46	50	0.937	0.892	0.928	
0.545	office-chair	23	36	0.772	0.639	0.747	
0.575	whiteboard	5	5	0.756	1	0.995	

0.281	table	6	7	0.678	0.286	0.515
0.697	tv	1	2	1	0.858	0.995

Speed: 0.3ms preprocess, 4.5ms inference, 0.0ms loss, 2.5ms postprocess per image
 Results saved to **/content/runs/detect/val4**
 {'mAP50-95': np.float64(0.6014659873419974), 'mAP50': np.float64(0.8538749666010362),
 'mAP75': np.float64(0.6938772198109348), 'per_class_mAP': array([0.60147,
 0.66323, 0.84841, 0.54502, 0.57484, 0.2808, 0.6965]), 'speed_ms':
 {'preprocess': 0.2843894878047073, 'inference': 4.465770878041661, 'loss':
 0.000504219514834581, 'postprocess': 2.45230668293161}}

book: AP50-95=0.601
 books: AP50-95=0.663
 monitor: AP50-95=0.848
 office-chair: AP50-95=0.545
 whiteboard: AP50-95=0.575
 table: AP50-95=0.281
 tv: AP50-95=0.697

Yolov5su

Ultralytics 8.3.199 🚀 Python-3.12.11 torch-2.8.0+cu126 CUDA:0 (Tesla T4, 15095MiB)
 YOLOv5s summary (fused): 84 layers, 9,114,245 parameters, 0 gradients, 23.8 GFLOPs
val: Fast image access ✅ (ping: 0.3±0.1 ms, read: 389.7±342.4 MB/s, size: 1996.2 KB)
val: Scanning /content/drive/MyDrive/Colab
 Notebooks/combined_dataset/val/labels.cache... 82 images, 2 backgrounds, 0 corrupt:
 100%

 82/82 126.2Kit/s 0.0s

	Class	Images	Instances	Box(P	R	mAP50	mAP50-
95): 100%		3/3	0.5it/s	6.2s			
0.612	all	82	253	0.826	0.823	0.857	
0.732	books	9	153	0.993	0.912	0.974	
0.895	monitor	46	50	0.918	0.893	0.937	
0.537	office-chair	23	36	0.809	0.707	0.791	
0.505	whiteboard	5	5	0.823	1	0.995	
0.254	table	6	7	0.517	0.429	0.453	
0.748	tv	1	2	0.895	1	0.995	

Speed: 3.5ms preprocess, 8.9ms inference, 0.0ms loss, 1.9ms postprocess per image
 Saving /content/runs/detect/val/predictions.json...
 Results saved to **/content/runs/detect/val**
 {'mAP50-95': np.float64(0.6117038921214883), 'mAP50': np.float64(0.8573696452223936),
 'mAP75': np.float64(0.668004007313841), 'per_class_mAP': array([0.6117,
 0.73167, 0.89489, 0.53681, 0.50512, 0.25362, 0.74813]), 'speed_ms':
 {'preprocess': 3.5127960975609223, 'inference': 8.92631363414483, 'loss':
 0.000636085363156581, 'postprocess': 1.8564270365836486}}

book: AP50-95=0.612
 books: AP50-95=0.732
 monitor: AP50-95=0.895
 office-chair: AP50-95=0.537
 whiteboard: AP50-95=0.505
 table: AP50-95=0.254
 tv: AP50-95=0.748

Model Comparison

Rank	Model	mAP50–95	mAP50	Inference (ms)	Notable per-class highlights
1	YOLOv8n (train3)	0.636	0.862	4.6	Whiteboard 0.720 (best), Book 0.636, Monitor 0.866
2	YOLOv5su	0.612	0.857	8.9	Monitor 0.895 (best), TV 0.748 (best), Books 0.732 (best)
3	YOLOv5n (train4)	0.601	0.854	4.5	TV 0.697, Monitor 0.848, Books 0.663
4	YOLOv8s (aug)	0.597	0.869	10.0	TV 0.697, Monitor 0.842, Books 0.640
5	YOLOv8s	0.594	0.845	10.5	Monitor 0.891, Books 0.726, TV 0.652
6	YOLO11n (train2)	0.591	0.869	6.5	Table 0.363 (best among non-v8n), Monitor 0.866

Overall, yolov8n has the best performance.

Both training and validation losses (box, cls, dfl) consistently decrease across epochs showcasing good convergence, and no divergence.

Validation losses are a bit higher and noisier than training losses which is expected since dataset is relatively small.