YOLO MODEL

DATA PREPARATION INFERENCE IN DGX SERVER LIVE VIDEO IN CLIENT MODEL DEVELOPMENT SETUP INFERENCE SERVER **CREATE YAML FILE AND MANAGE SET UP WEBCAM KEEP SEG POINTS IN IMAGE BOUNDARIES** app = FastAPI() **DIRECTORIES** cap = cv2.VideoCapture(0) • Ensure all x and y coordinates in the segmentations stay within image boundaries (width and height). LOAD MODEL SEND FRAME FOR model = YOLO("models/run01.onnx") **EXPERIMENT TRACKING PROCESSING** KEEP BBOX POINTS (X, Y, W, H) IN IMAGE & SEG comet ml.login() • Encode each frame as a JPEG. **BOUNDARIES** PROCESS UPLOADED FRAME: • Send the frame to a remote server for • Constrain X and Y coordinates to the image dimensions. • Receive an image frame from the client. **MODEL TRAINING** segmentation via an HTTP POST • Adjust width W and height H to ensure the bounding box fits • Convert it to a format suitable for the model = YOLO("yolo11n-seg.pt") request. within the image. model. • If segmentation data exists, ensure bounding boxes align results = model.train(with the segmentation boundaries. **RECEIVE AND DISPLAY RUN INFERENCE:** data="dataset.yaml",

CONVERT TO YOLO FORMAT

convert_coco(labels_dir, save_dir, use_segments=True, use_keypoints=False, cls91to80=False, lvis=False)

NORMALIZE (0, 1) AND REMOVE DUPLICATES

project = "yolo-project-fin", epochs=200, imgsz=640, batch=32, patience=10, name="run01")

results = model(frame, conf=0.8) segmented frame = result[0].plot()

SEND RESULT BACK:

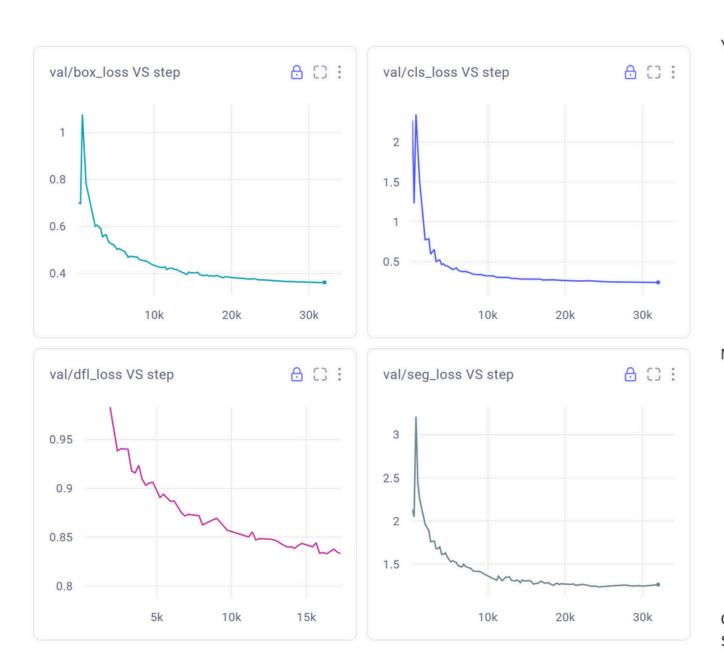
- Encode the segmented image as a JPEG.
- Return the processed image to the client.

RESULTS

- Decode the segmented frame returned by the server.
- Calculate and overlay FPS (frames per second) on the segmented frame.

Can be done with gradio

RESULTS



YOLO11n-seg summary (fuse	ed): 265 l	ayers, 2,839	,248 paramet	ers, 0 gradi	ients, 10.	2 GFLOPs				
Class		Instances	Box(P	R		mAP50-95)	Mask(P	R	mAP50	mAP50-95):
all	845	1863	0.968	0.935	0.953	0.899	0.968	0.931	0.946	0.866
bottled_soda	37	37	0.989	1	0.995	0.927	0.993	1	0.995	0.916
cheese	40	58	0.98	0.948	0.981	0.901	0.983	0.948	0.981	0.889
Kitkat	42	47	1	0.971	0.98	0.916	1	0.968	0.98	0.903
coffee	30	66	0.984	0.935	0.985	0.91	1	0.948	0.964	0.903
condensed_milk	37	52	0.968	1	0.989	0.963	0.97	1	0.989	0.952
Cooking_oil	40	40	0.993	1	0.995	0.955	0.995	1	0.995	0.932
corned_beef	41	247	0.972	0.986	0.987	0.935	0.972	0.978	0.987	0.843
garlic	33	33	0.968	0.939	0.985	0.905	0.976	0.939	0.985	0.888
instant_noodles	30	30	0.988	1	0.995	0.981	0.99	1	0.995	0.989
ketchup	35	35	0.99	1	0.995	0.995	0.991	1	0.995	0.991
lemon	37	46	0.982	1	0.995	0.982	0.985	1	0.995	0.988
Nestle All Purpose Cream	3	5 63	0.989	0.984	0.993	0.961	0.991	0.984	0.993	0.942
Nestle All Purpose Cream Mayonnaise	3 31	5 63 105	0.989 0.943	0.984 0.941	0.993 0.988	0.961 0.9	0.991 0.923	0.984 0.917	0.993 0.963	0.942 0.745
Mayonnaise	31	105	0.943	0.941	0.988	0.9	0.923	0.917	0.963	0.745
Mayonnaise peanut_butter	31 26	105 158	0.943 0.908	0.941 0.519	0.988 0.609	0.9 0.577	0.923 0.891	0.917 0.506	0.963 0.581	0.745 0.511
Mayonnaise peanut_butter pasta	31 26 37	105 158 39	0.943 0.908 0.972	0.941 0.519 0.974	0.988 0.609 0.983	0.9 0.577 0.913	0.923 0.891 0.978	0.917 0.506 0.974	0.963 0.581 0.983	0.745 0.511 0.922
Mayonnaise peanut_butter pasta pineapple_juice	31 26 37 35	105 158 39 116	0.943 0.908 0.972 0.941	0.941 0.519 0.974 0.948	0.988 0.609 0.983 0.972	0.9 0.577 0.913 0.942	0.923 0.891 0.978 0.946	0.917 0.506 0.974 0.948	0.963 0.581 0.983 0.972	0.745 0.511 0.922 0.921
Mayonnaise peanut_butter pasta pineapple_juice crackers	31 26 37 35 32	105 158 39 116 48	0.943 0.908 0.972 0.941 0.922	0.941 0.519 0.974 0.948 0.979	0.988 0.609 0.983 0.972 0.967	0.9 0.577 0.913 0.942 0.929	0.923 0.891 0.978 0.946 0.929	0.917 0.506 0.974 0.948 0.979	0.963 0.581 0.983 0.972 0.967	0.745 0.511 0.922 0.921 0.937
Mayonnaise peanut_butter pasta pineapple_juice crackers sardines	31 26 37 35 32 45	105 158 39 116 48 61	0.943 0.908 0.972 0.941 0.922 0.978	0.941 0.519 0.974 0.948 0.979 0.967	0.988 0.609 0.983 0.972 0.967 0.974	0.9 0.577 0.913 0.942 0.929 0.91	0.923 0.891 0.978 0.946 0.929 0.98	0.917 0.506 0.974 0.948 0.979 0.967	0.963 0.581 0.983 0.972 0.967 0.986	0.745 0.511 0.922 0.921 0.937 0.88
Mayonnaise peanut_butter pasta pineapple_juice crackers sardines pink_shampoo	31 26 37 35 32 45 36	105 158 39 116 48 61 51	0.943 0.908 0.972 0.941 0.922 0.978 0.993	0.941 0.519 0.974 0.948 0.979 0.967	0.988 0.609 0.983 0.972 0.967 0.974 0.995	0.9 0.577 0.913 0.942 0.929 0.91 0.993	0.923 0.891 0.978 0.946 0.929 0.98 0.994	0.917 0.506 0.974 0.948 0.979 0.967	0.963 0.581 0.983 0.972 0.967 0.986 0.995	0.745 0.511 0.922 0.921 0.937 0.88 0.99
Mayonnaise peanut_butter pasta pineapple_juice crackers sardines pink_shampoo	31 26 37 35 32 45 36 34	105 158 39 116 48 61 51	0.943 0.908 0.972 0.941 0.922 0.978 0.993 0.989	0.941 0.519 0.974 0.948 0.979 0.967 1	0.988 0.609 0.983 0.972 0.967 0.974 0.995	0.9 0.577 0.913 0.942 0.929 0.91 0.993 0.989	0.923 0.891 0.978 0.946 0.929 0.98 0.994	0.917 0.506 0.974 0.948 0.979 0.967	0.963 0.581 0.983 0.972 0.967 0.986 0.995	0.745 0.511 0.922 0.921 0.937 0.88 0.99 0.987
Mayonnaise peanut_butter pasta pineapple_juice crackers sardines pink_shampoo soap silverswan	31 26 37 35 32 45 36 34	105 158 39 116 48 61 51 34	0.943 0.908 0.972 0.941 0.922 0.978 0.993 0.989	0.941 0.519 0.974 0.948 0.979 0.967 1 1	0.988 0.609 0.983 0.972 0.967 0.974 0.995 0.995	0.9 0.577 0.913 0.942 0.929 0.91 0.993 0.989 0.743	0.923 0.891 0.978 0.946 0.929 0.98 0.994 0.991	0.917 0.506 0.974 0.948 0.979 0.967 1 1	0.963 0.581 0.983 0.972 0.967 0.986 0.995 0.995	0.745 0.511 0.922 0.921 0.937 0.88 0.99 0.987 0.581
Mayonnaise peanut_butter pasta pineapple_juice crackers sardines pink_shampoo soap silverswan toothpaste	31 26 37 35 32 45 36 34 34 25	105 158 39 116 48 61 51 34 95 42 327	0.943 0.908 0.972 0.941 0.922 0.978 0.993 0.989 0.907	0.941 0.519 0.974 0.948 0.979 0.967 1 1 0.821	0.988 0.609 0.983 0.972 0.967 0.974 0.995 0.995	0.9 0.577 0.913 0.942 0.929 0.91 0.993 0.989 0.743 0.944 0.527	0.923 0.891 0.978 0.946 0.929 0.98 0.994 0.991 0.868 0.993	0.917 0.506 0.974 0.948 0.979 0.967 1 1 0.763	0.963 0.581 0.983 0.972 0.967 0.986 0.995 0.995 0.855 0.995	0.745 0.511 0.922 0.921 0.937 0.88 0.99 0.987 0.581 0.925 0.416

RESULTS

