# Integrating storage manager (TASM) and query optimizer (FiGO) in Video-DBMS

Data Systems Project (Instructor: Prof. Kamal Karlapalem)

Name: Chirag Parikh Roll no: 2022900005

Email: <u>chirag.parikh@research.iiit.ac.in</u>

GitHub: <a href="https://github.com/chirag26495/FiGO\_TASMst">https://github.com/chirag26495/FiGO\_TASMst</a>

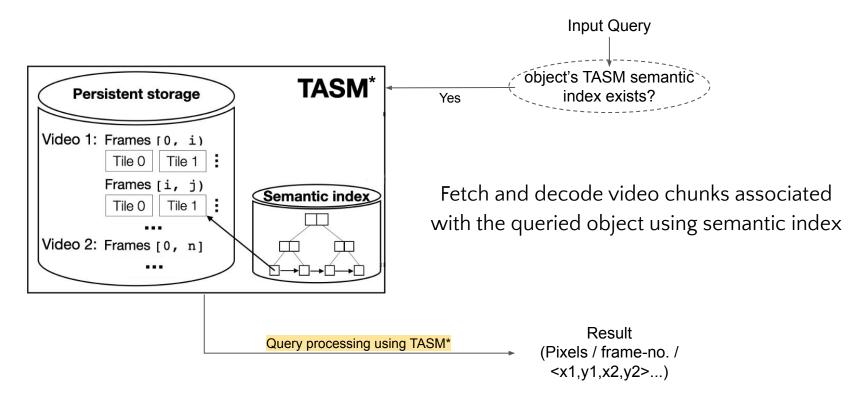
# Integration Method

# FiGO [2] and TASM [1] integration in VDBMS

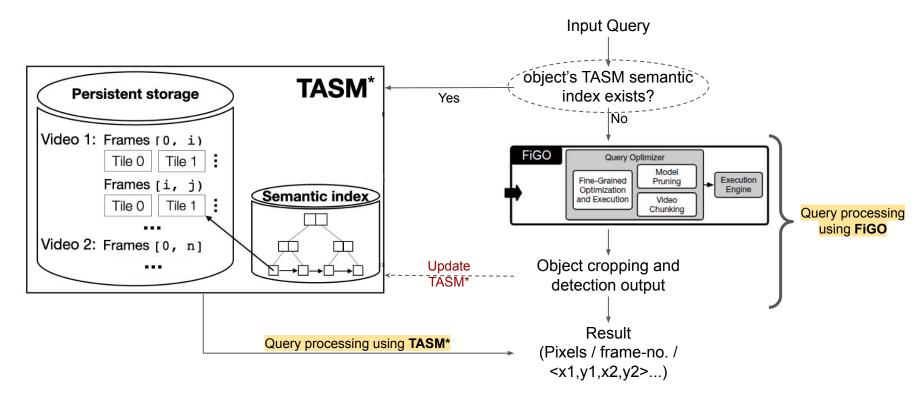
```
Input object query
            queried Video-ID
                                                                      object's TASM semantic
"v 1590672109537497088 uX7AXwvNLW-3oMdN hevcfixed": {
   "TASM video store": [
                                                                             index exists?
      "-untiled"
   "TASM indexed objects": [
                                                 check processed
      "person",
       "car",
                                               data stored in JSON
      "bicycle",
      "motorcycle",
      "traffic light",
       "bus"
   "FiGO scanned objects": [
      "person",
      "car",
      "bicycle",
      "motorcycle",
      "traffic light",
      "bus",
       "stop sign",
      "truck",
      "parking meter",
       "cow"
                                 processed videos stored inTASM.json
   "TASM video store": |
```

<sup>[1]</sup> Daum, Maureen, et al. "Tasm: A tile-based storage manager for video analytics." 2021 IEEE 37th International Conference on Data Engineering (ICDE). IEEE, 2021.

<sup>[2]</sup> Cao, Jiashen, et al. "Figo: Fine-grained query optimization in video analytics." Proceedings of the 2022 International conference on management of data. 2022.



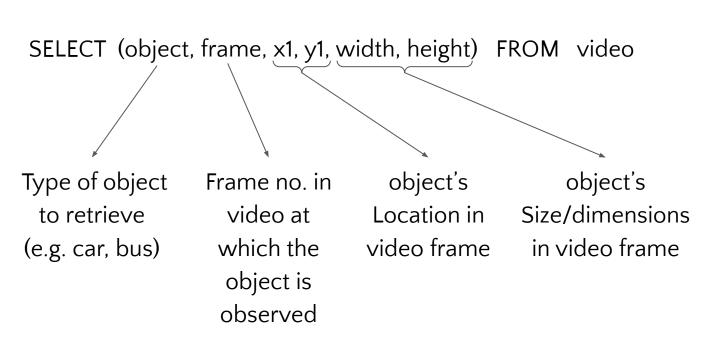
**Note**: As opposed to TASM, **TASM\*** stores videos in 8-frame chunks instead of full video in untitled form. This enables **faster retrieval of relevant frames** as per input object query.



Update TASM\* semantic index using detection output and store video in untiled 8-frame chunk form (unlike TASM) for faster retrieval of future queries

Input Query template

# Input Query template:



Sample Query Output

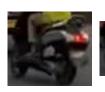


query Video

... **623** frames

rame563 motorcycle x198 y608 w28 h50.jpg

SELECT (**motorcycle**, frame, x1, y1, width, height) FROM video1\_hevcfixed.mp4







... 1392 objects retrieved

→from **504** frames

→stored as **cropped jpgs** 

152   153   154   155   155   155   155   155   155   155   155   155   155   155   155   155   155   155   155   155   155   155   155   155   155   155   155   155   155   155   155   155   155   155   155   155   155   155   155   155   155   155   155   155   155   155   155   155   155   155   155   155   155   155   155   155   155   155   155   155   155   155   155   155   155   155   155   155   155   155   155   155   155   155   155   155   155   155   155   155   155   155   155   155   155   155   155   155   155   155   155   155   155   155   155   155   155   155   155   155   155   155   155   155   155   155   155   155   155   155   155   155   155   155   155   155   155   155   155   155   155   155   155   155   155   155   155   155   155   155   155   155   155   155   155   155   155   155   155   155   155   155   155   155   155   155   155   155   155   155   155   155   155   155   155   155   155   155   155   155   155   155   155   155   155   155   155   155   155   155   155   155   155   155   155   155   155   155   155   155   155   155   155   155   155   155   155   155   155   155   155   155   155   155   155   155   155   155   155   155   155   155   155   155   155   155   155   155   155   155   155   155   155   155   155   155   155   155   155   155   155   155   155   155   155   155   155   155   155   155   155   155   155   155   155   155   155   155   155   155   155   155   155   155   155   155   155   155   155   155   155   155   155   155   155   155   155   155   155   155   155   155   155   155   155   155   155   155   155   155   155   155   155   155   155   155   155   155   155   155   155   155   155   155   155   155   155   155   155   155   155   155   155   155   155   155   155   155   155   155   155   155   155   155   155   155   155   155   155   155   155   155   155   155   155   155   155   155   155   155   155   155   155   155   155   155   155   155   155   155   155   155   155   155   155   155   155   155   155   155   155	height
trame561_motorcycle_x552_y596_w48_h50.jpg         153         motorcycle         398         522         78           trame562_motorcycle_x198_y606_w30_h48.jpg         153         motorcycle         132         546         200           frame562_motorcycle_x342_y574_w20_h34.jpg         154         motorcycle         372         524         74           frame562_motorcycle_x370_y600_w42_h68.jpg         154         motorcycle         124         548         208           frame562_motorcycle_x502_y580_w20_h30.jpg         155         motorcycle         346         524         72           frame563_motorcycle_x548_y596_w48_h52.jpg         155         motorcycle         116         550         218           frame563_motorcycle_x372_y602_w42_h68.jpg         156         motorcycle         120         544         230           frame563_motorcycle_x500_y582_w32_h30.jpg         156         motorcycle         292         526         70	16
ftrame562_motorcycle_x198_y606_w30_h48.jpg         153         motorcycle         132         546         200           ftrame562_motorcycle_x342_y574_w20_h34.jpg         154         motorcycle         372         524         74           ftrame562_motorcycle_x370_y600_w42_h68.jpg         154         motorcycle         124         548         208           ftrame562_motorcycle_x502_y580_w20_h30.jpg         155         motorcycle         346         524         72           ftrame562_motorcycle_x548_y596_w48_h52.jpg         155         motorcycle         116         550         218           ftrame563_motorcycle_x372_y602_w42_h68.jpg         156         motorcycle         120         544         230           ftrame563_motorcycle_x500_y582_w32_h30.jpg         156         motorcycle         292         526         70	82
frame562_motorcycle_x370_y600_w42_h68.jpg         154         motorcycle         124         548         208           frame562_motorcycle_x502_y580_w20_h30.jpg         155         motorcycle         346         524         72           frame562_motorcycle_x548_y596_w48_h52.jpg         155         motorcycle         116         550         218           frame563_motorcycle_x198_y608_w28_h50.jpg         156         motorcycle         120         544         230           frame563_motorcycle_x500_y582_w32_h30.jpg         156         motorcycle         292         526         70	332
frame562_motorcycle_x502_y580_w20_h30.jpg         155         motorcycle         346         524         72           frame562_motorcycle_x548_y596_w48_h52.jpg         155         motorcycle         116         550         218           frame563_motorcycle_x198_y608_w28_h50.jpg         156         motorcycle         120         544         230           frame563_motorcycle_x500_y582_w32_h30.jpg         156         motorcycle         292         526         70	78
Image: Second control of the following in the following	356 76 384 452 74
Image: Frame 1563 motorcycle x198 y608 w28 h50.jpg     155 motorcycle     116 motorcycle     116 motorcycle     120 motorcycle     218 motorcycle       Image: Frame 1563 motorcycle x372 y602 w42 h68.jpg     156 motorcycle     120 motorcycle     230 motorcycle       Image: Frame 1563 motorcycle x500 y582 w32 h30.jpg     156 motorcycle     292 motorcycle     292 motorcycle	
Image: Frame 163 motorcycle x372 y602 w42 h68.jpg     156 motorcycle     120 motorcycle     544 230       Image: Frame 163 motorcycle x500 y582 w32 h30.jpg     156 motorcycle     292 526 70	
Image:	
156 motorcycle v544 v596 w46 h52 ing 156 motorcycle 79	164
☐ frame562_motorcycle_x370_y600_w42_h68.jpg	
If rame562_motorcycle_x548_y596_w48_h52.jpg       156 motorcycle   120   544   230   452         158 motorcycle   202   528   70   74         158 motorcycle   202   528   70         158 motorcycle   202   528   70         158 motorcycle   202   528         158 motorcycle   202         158 motorcycle   202   202         158 motorcycle         158 mo	

query-output in csv



query Video

... **623** frames

SELECT (car, frame, x1, y1, width, height) FROM video1\_hevcfixed.mp4

### query Output



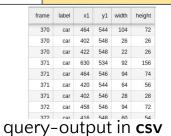






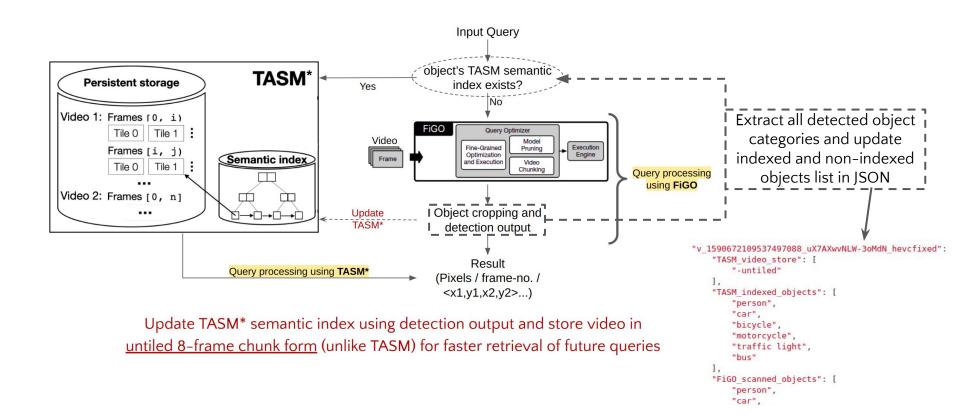
- ... 1649 objects retrieved →from 530 frames
  - $\rightarrow$ stored as **cropped jpgs**

Naı	ne
Ľ	frame105_car_x392_y442_w328_h326.jpg
Ľ	frame105_car_x496_y416_w78_h42.jpg
ı.	frame106_car_x214_y476_w68_h90.jpg
ď	frame106_car_x382_y474_w26_h22.jpg
ď	frame106_car_x394_y442_w326_h330.jpg
ď	frame106_car_x400_y472_w22_h24.jpg
ď	frame106_car_x496_y418_w78_h42.jpg
ď	frame106_car_x640_y436_w80_h86.jpg
M	frame107_car_x216_y476_w66_h90.jpg

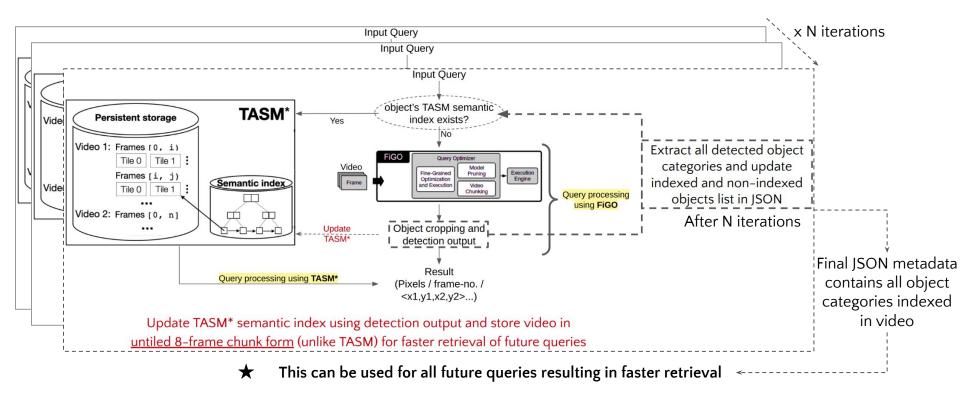


Automated retrieval and profiling of all object categories present in query video

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# Automated retrieval and profiling of all object categories present in query video

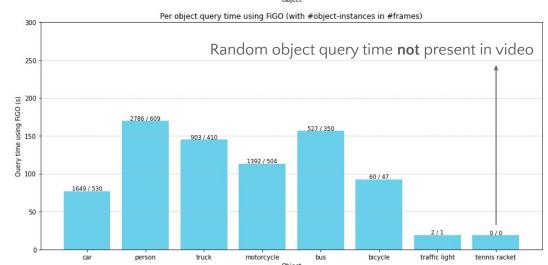


Note: Value of N is iteratively determined based on the number of object categories present in video

**Results** of

Automated retrieval and profiling

### Per object query time using modified TASM after FiGO integration (with #object-instances in #frames) 2.00 < 1.5s query time with FIGO & TASM\* integration, compared to $\star$ 1.75 -200s using FiGO only (\*Note: This includes cropped-jpg save time= fied TASM and FIGO (s) Total\_time - TASM\_query\_time) 1649 / 530 1.00 903/410 1392 / 504 527 / 350 0.25 60 / 47 2/1 0/0 0.00 car person truck motorcycle bus bicycle traffic light tennis racket Object



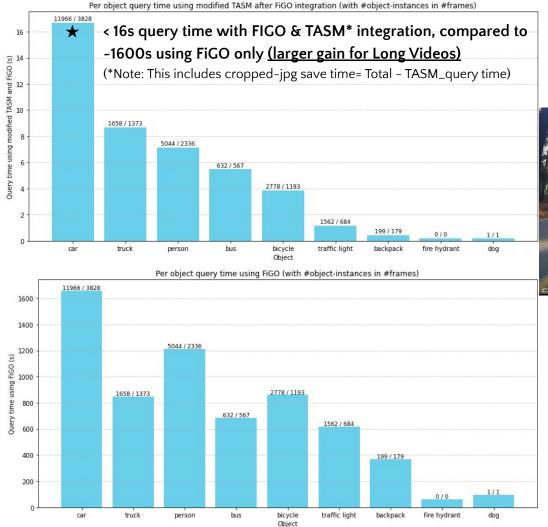
### **Query Video**



Total **623** frames (High Object Density)

**NOTE:** FiGO query times are **more** for <u>smaller objects</u> (e.g. person) in video as it requires **best model** having **high latency** 

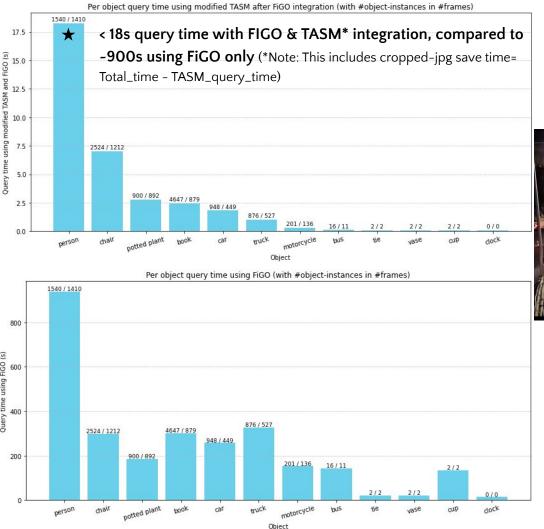
More Examples





Total **4645** frames (Long duration Video)



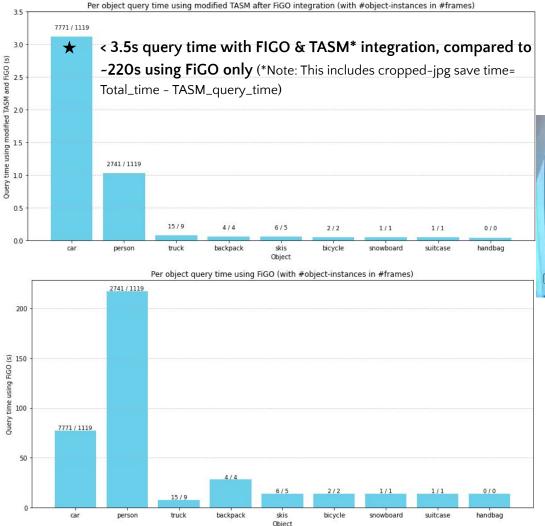




Total 1720 frames

**NOTE:** FiGO query times are **more** for smaller objects (e.g. person) in video as it requires **best model** having **high latency** 



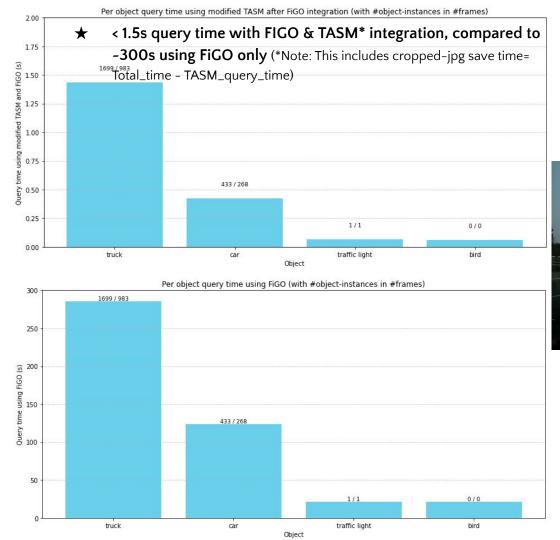




Total **1119** frames

**NOTE:** FiGO query times are **more** for <u>smaller objects</u> (e.g. person) in video as it requires **best model** having **high latency** 



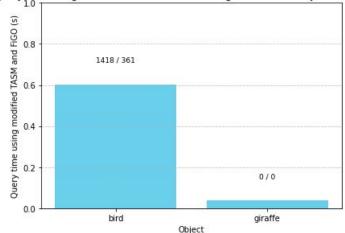


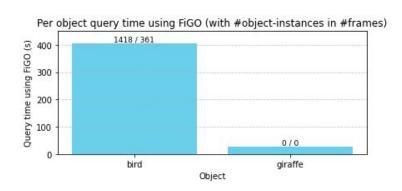


Total **1372** frames (Low Object Density)

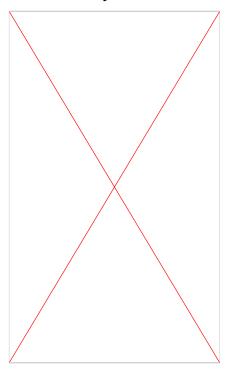
★ < 0.6s query time with FIGO & TASM\* integration, compared to -400s using FiGO only







### **Query Video**



Total **361** frames

# Challenges

- GPU differences (Quadro P5000 theirs vs RTX 30/20 series ours)
  - Uniform and non-uniform tile layout storage working now Huge thanks to Maureen Daum
    - Solution: Storing video in 8-frame chunks sequentially
  - **But** object retrieval from tiled video storage still fails:
    - Retrieving objects from **chunked + tiled** video storage differs greatly from just **tiled** video storage
    - This needs rethinking the TASM approach and significant code restructuring
    - Even after doing the above, The resultant performance gains to TASM+FiGO integration (proposed in this project) won't differ much in execution time since videos are now chunked so only relevant chunks of the video are read instead of reading it entirely as before
  - Thus for this project untiled video storage split in 8-frame chunks is used.

‡ BOARD	‡ FAMILY	‡ NVENC Generation	Desktop/ Mobile	CHIPS	Total # of NVENC	Max # of concurrent sessions	H.264 (AVCHD) YUV 4:2:0	H.264 (AVCHD) YUV 4:2:2	H.264 (AVCHD YUV 4:4			H.26: (HEV 4:2:0 YUV	C) (HE	65 H. VC) (H YUV 4:4:4 4
Quadro P5000	Pascal	6th Gen	D	1	2	Unrestricted	YES	NO	YES	YES	YES	NO	YES	S YE
‡ BOARD	‡ FA	MILY T NVE Gener		ktop/ # Di ile CHI	PS #	otal Max#o of concurr ZENC session	rent (AVCI	HD) (AVC	HD) (A	AVCHD) (/	AVCHD) (H	.265 HEVC) K YUV 4:2:0	H.265 (HEVC) YUV 4:2:2	H.265 (HEVC) 4K YUV 4:4:4
GeForce RTX 30	90 Amp	ere 7th Ge	en D	1	1	8	YES	NO	Y	ES Y	ES YE	ES	NO	YES

Source: https://developer.nvidia.com/video-encode-and-decode-gpu-support-matrix-new

# Contributions

- ★ FiGO and TASM\* integration: Significant query time reduction for repeated objects retrieval
  - Stores object-list metadata for each video →Supporting faster query execution for already indexed objects
- \* Automated extraction of all object categories present in a video
  - Obtained iteratively through multiple FiGO+TASM query runs →provide options for valid/useful object queries
- ★ Scalable for storage and **fast object retrieval** from **long duration videos** 
  - Due to 8-frame chunked video storage in modified TASM\* only relevant video chunks containing the queried object will be retrieved  $\rightarrow$  Suited for real-world applications like city-scale analysis on CCTV videos using this VDBMS
- ★ FiGO and TASM\* both support <u>multiple object attribute retrieval</u>:
  - object pixels, frame-level object location and object dimensions
- ★ TASM\* now works on consumer grade GPUs
  - Thanks Maureen: for 8-frame chunked sequential video storage to run on RTX 30/20 series GPUs

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