

2020
embedded
VISION
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Multi-modal Re-Identification: IOT + Vision for Residential Community Tracking

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- **Landscape** - Residential Communities in China
- **Problem** – Why tracking in communities?
- **Multi-modal tracking** - framework and results
 - Tracklets
 - Identity Stamping
 - Identity Conflation
- **Conclusions**

Residential Communities in China

Residential communities focus on the **needs of households**, providing secure housing integrated with recreation, groceries, retail and lifestyle services.



The Seedland Group

Craftmanship creates quality life products

For 14 years, the Seedland Real Estate Group Co., Ltd. has been committed to the exploration and innovation of human science and technology in all aspects of life, connecting science and technology with humanities, and redefining human understanding of the relationship between themselves and living space.

31 cities
45 projects
6 fastest city clusters



Guangzhou



Chongqing



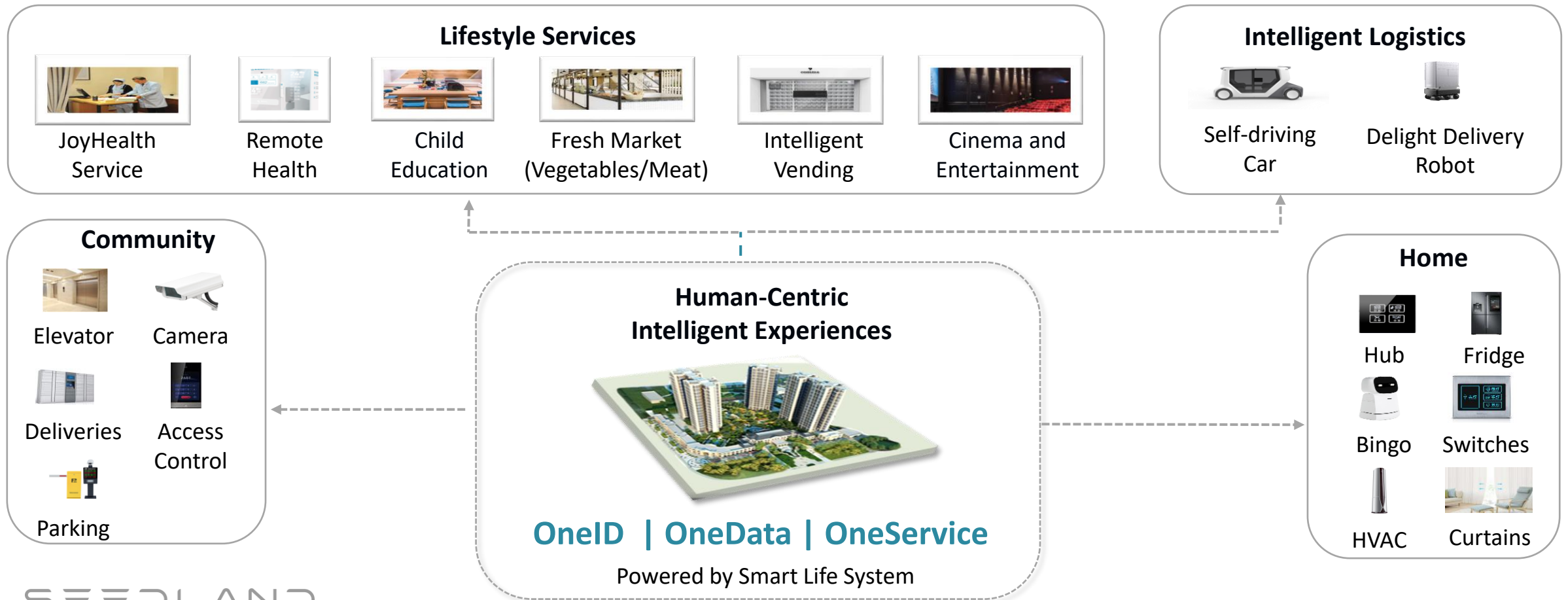
Qingdao



Shenzhen

Seedland Smart Community Ecosystem

Our goal is to pair smart **home**, smart **community**, signature smart premium **lifestyle services** and intelligent **logistics** to provide a **premier residential lifestyle experience**.



SLS Smart Community Solutions

Intelligent security, child safety, people flow analysis, intelligent delivery and contactless access control based on real-time IOT, cameras and AI.



Smart Life System

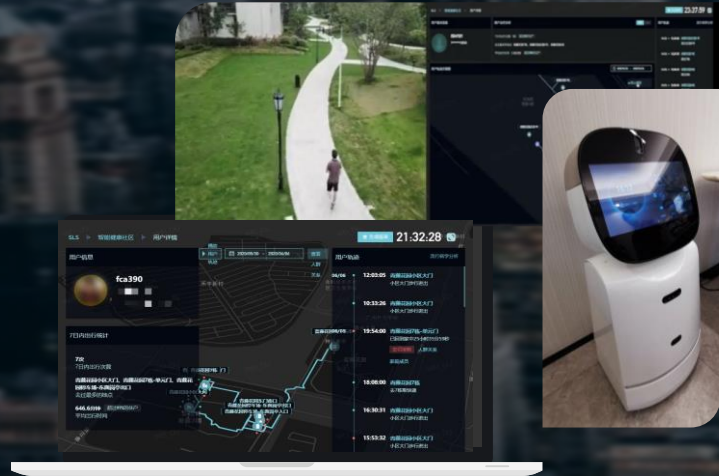
Community, Health, Business

THREE SLS SOLUTIONS

Operating in Guangzhou Ivy

SLS Intelligent Community Health

Intelligent health situational back-tracing for emergency contact tracing and multi-level risk alerts.



SLS Smart Business Services

Personalized and intelligent shopping through integrated community-wide ordering, standardized delivery, AI customer service assistance, intelligent business operations analysis.





Community Tracking

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Why Community Tracking?

Community tracking **delivers value** to residents and **property management** while strictly **preserving privacy** and allowing **user opt-out**.

Playground Tracing

Only residents/approved visitors enter child areas



Delivery Tracing

Delivery workers do not visit unnecessary areas.



Prohibited Tracing

Tracing rule breakers for property management



Contact Tracing

Contact tracing for non-phone users (kids/elderly)



Suspicious Tracing

Retracing paths of suspicious behavior

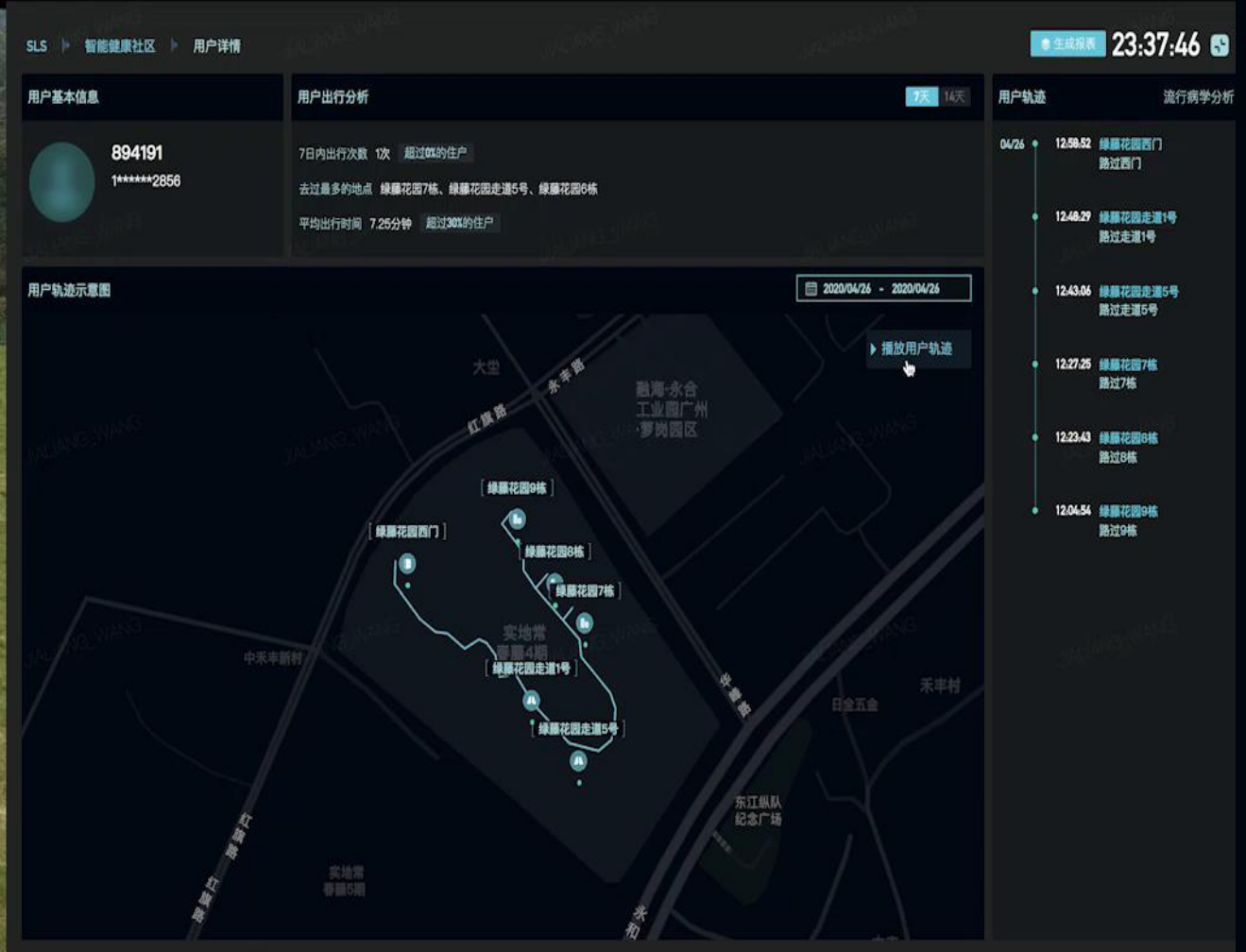


Lost Item Tracing

Retracing to find lost items, phones, toys, ...



Emergency Contact Back-Tracing



Combining Multiple Modalities

A user will interact with multiple devices and multiple modalities in their journey. Multi-modal allows us to see beyond cameras and generate semantic knowledge.



1. Family Structure



2. Travel Habits



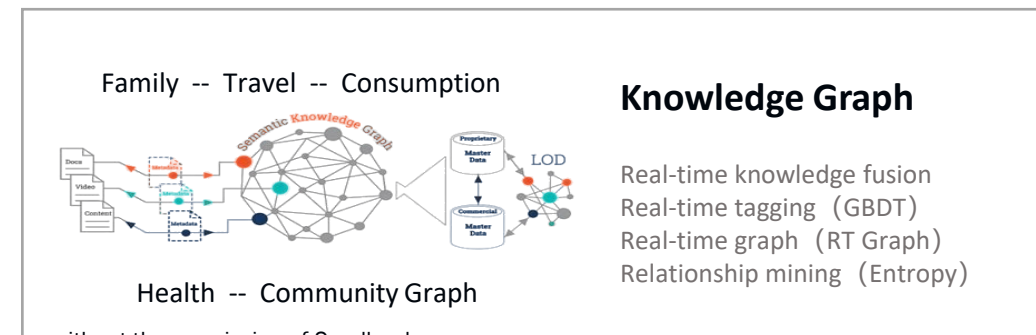
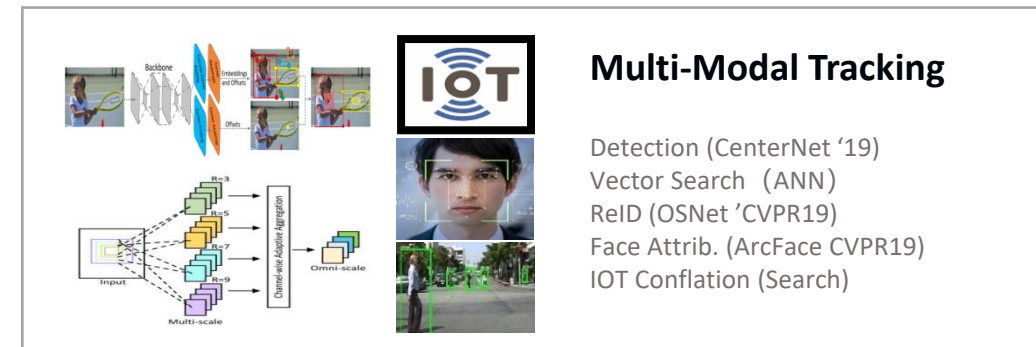
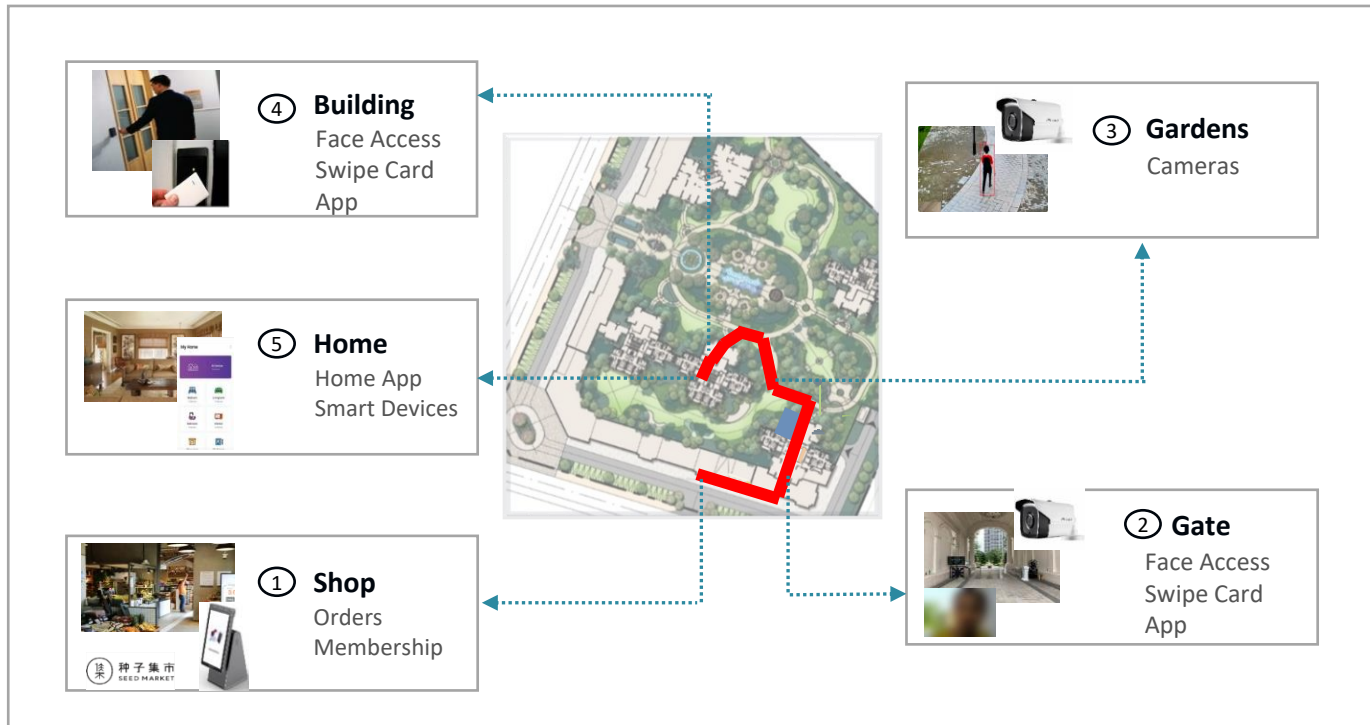
3. Consumption



4. Health



5. Community Graph



Face-Based Identity



- Bias - Face recognition is known to perform poorly on children and elderly.
- Angle - High-positioned security cameras make recognition difficult.

Device-Based Identity



- Mobile phones are NOT everywhere - children and elderly.
- Bluetooth-tracking low penetration in China.

Privacy



- Must support do-not-track.
- Must support usage for intended purposes only.

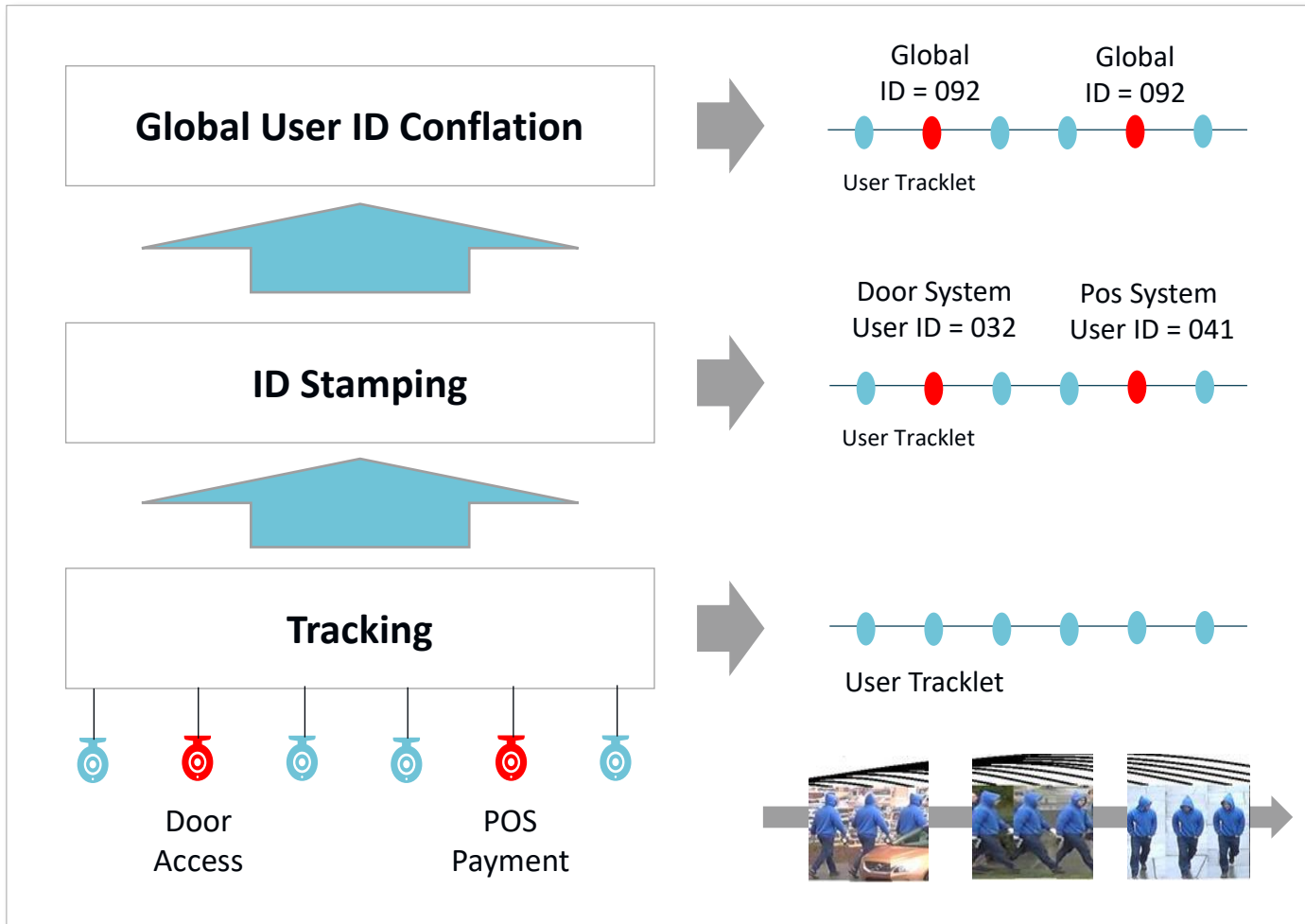
Obtain **explicit consent** during smart gate registration

Consent agreement clearly lists **specific user-value features** – no blanket permission for tracking

Non-ID associated short-term tracking (eg. suspicious person back-tracing) are **implicitly agreed** to via **security notice**



Cross-Day Stable User ID Tracking



Key Challenges

- ID stable across days
- Real-time
- Cost-efficient
- Visual invariance (clothes, bags, hats, ...)
- Population bias (children, elderly).

Leverage multiple modalities while real-time and cost efficient.



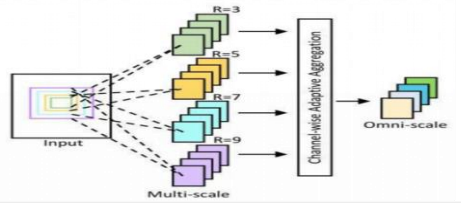
Tracking

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Cross-Camera Tracking (REID)

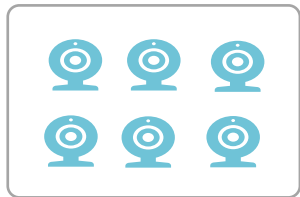
Key idea: Recognize same person across cameras by searching a continuously updated database of previously-seen entities

OSNet¹ multi-scale feature learning
robust to scale, camera-distance



Tracklet ID - Assign to entire track
(smooth per-frame noisy recognition)

t	2	3	4	5	6	7
id	014	014	014	012	014	012
m	0.8	0.9	0.9	0.7	0.5	0.8



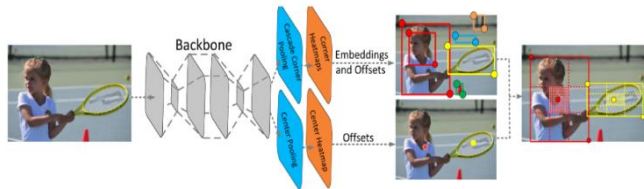
Camera Group

ID-Aware
Tracking

Omni-Scale
Embedding

Cross-Camera
ReID Search

Tracklet
ID Assign



ID-aware tracking
dramatically improves
tracking MOTA and high FPS

CenterNet (detection) +
Kalman (trajectory) +
Embedding Distance (identity)

Seen Entities

id	camera	exemplars	m
013	c41	[...]	0.8
013	c44	[...]	0.9

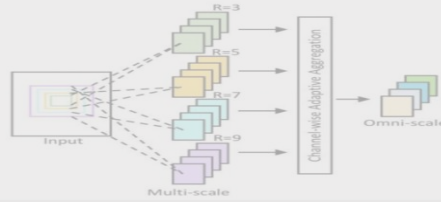
Detect **new entities**
Add **new exemplars** for
known entities

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Cross-Camera Tracking (REID)

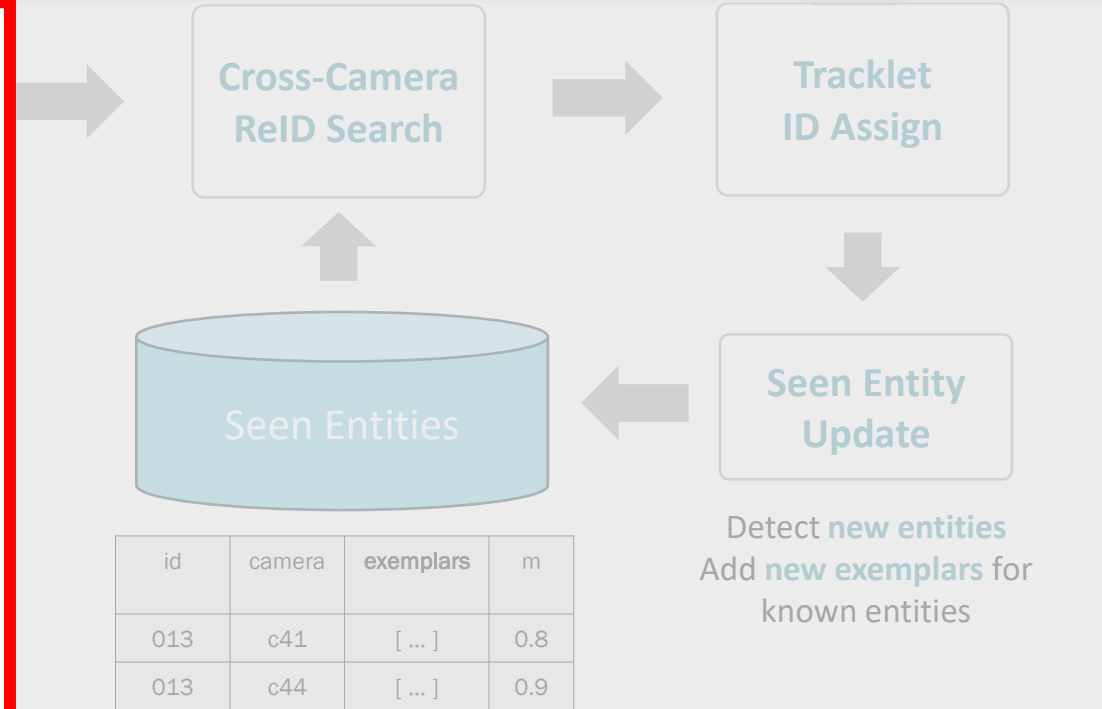
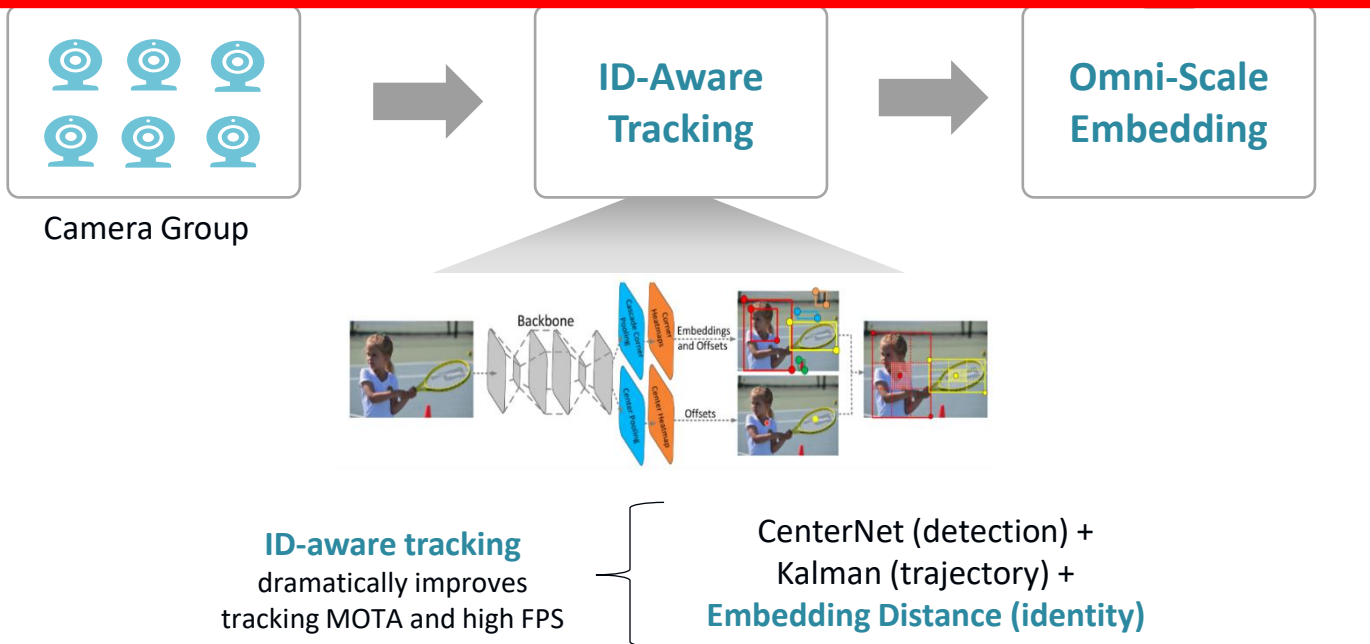
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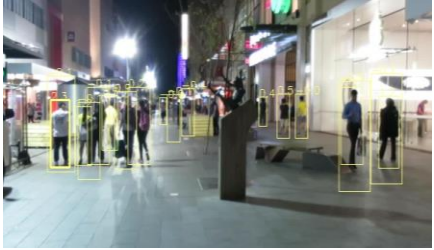


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Tracking Tasks and Public Benchmarks

Single-Camera Tracking – Detect and track a person within one camera



MOT16



MOT20

Multiple Object Tracking Benchmark

T	↑MOTA	IDF1	MOTP	MT	ML	FP	FN	Recall	Precision	FAF	ID Sw.	Frag	Hz		
SeedTrack 1.	68.7	±9.8	70.7	75	79.0	1,131	396	52,980	121,122	78.5	89.3	3.0	2,571	4,080	591.9
				(48.0)	(16.8)						(32.7)	(52.0)			
Fair 2.	68.5	±9.9	71.6	80.3	915	417	36,831	138,351	75.5	92.0	2.1	2,562	7,656	25.9	
				(38.9)	(17.7)						(33.9)	(101.4)			
<small>Y. Zhang, C. Wang, X. Wang, W. Zeng, W. Liu. A Simple Baseline for Multi-Object Tracking. In arXiv preprint arXiv:2004.01888, 2020.</small>															
CTTrack17 3.	67.8	±15.9	64.7	73.3	78.4	816	579	18,498	160,332	71.6	95.6	1.0	3,039	6,102	3.8
				(34.6)	(24.6)						(42.5)	(85.2)			
<small>X. Zhou, V. Koltun, P. Kr'ahenb'uhl. Tracking Objects as Points. In ECCV, 2020.</small>															

Cross-Camera Tracking (REID) – Identify the same user across different cameras



Duke MTMC Set



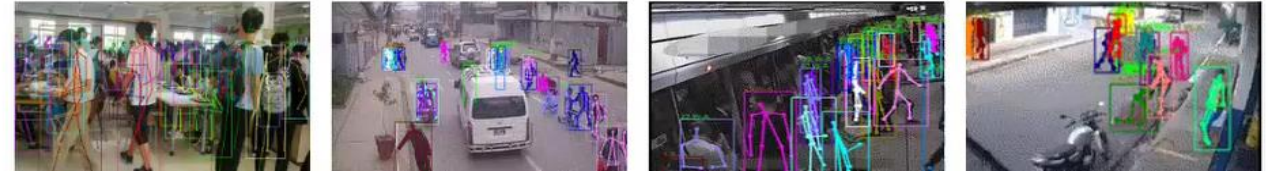
MSMT17 Set

Method	Benchmark Set Top-1 Accuracy			Model Memory Size
	Market1501	DukeMTMC	MSMT17	
Seedland	95.1	88.6	78.7	11M
PCB	93.8	83.3	68.2	120M
BFE-Net	94.0	88.9	-	130M
DG-Net	94.8	83.6	77.2	108M

Tracking is critical for down-stream understanding

Learnings from ACM2020 Grand Challenge

- #1 position in the ACM 2020 Multimedia Grand Challenge for Large-scale Human-centric Video Analysis international competition.
- High ranked in down-stream intelligence (pose tracking, action recognition) **primarily because of improved dense crowd tracking**

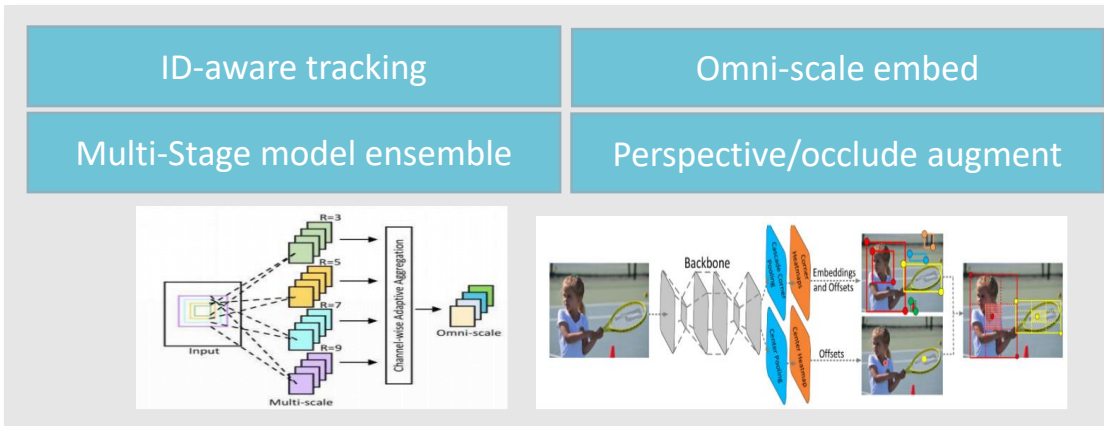


Team Name	# Institution	MOTA
[1st] Adaptive FairMOT	iSEE-SYSU & ACCUVISION	60.2282
[2nd] JiaRen.AI	Seedland	56.0474
[3rd] Crowd-Tracker	Xidian University	55.5163
Try private	Amazon	47.8120
NewTracker	Tencent	46.4815

Team Name	# Institution	w_AP@avg
[1st] Seedland.Tech	Seedland	57.5091
[2nd] ccc	YiTu & National University of Singapore	56.3375
[3rd] DH_IBA	Dahua Technology	55.1719
JDAI	JD Tech & UESTC	55.0139
WanDan	UESTC	54.5282

Team Name	# Institution	MOTA
[1st] Seedland.Tech	Seedland	63.9686
[2nd] Try	YiTu & National University of Singapore	61.7941
[3rd] SimpleTrack	Chinese University of Hong Kong	56.9834
DeepBlueAI	DeepBlueAI	55.1543
Commander_test4	XFORWARDAI	53.7671

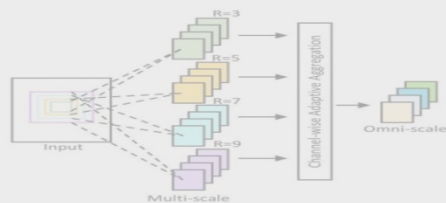
Team Name	# Institution	mAP@avg
[1st] MSF	YiTu & National University of Singapore	0.2605
[2nd] VM	Seedland	0.2548
[3rd] CF	City University of Hong Kong	0.1531
8A	Tencent	0.1509
only_person_rgb	Sun Yat-sen University	0.1086



Cross-Camera Tracking (ReID)

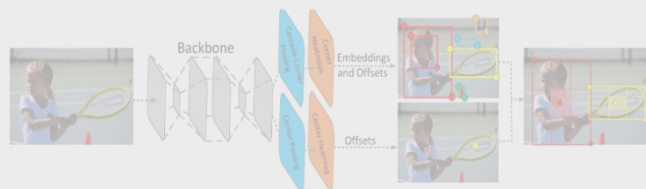
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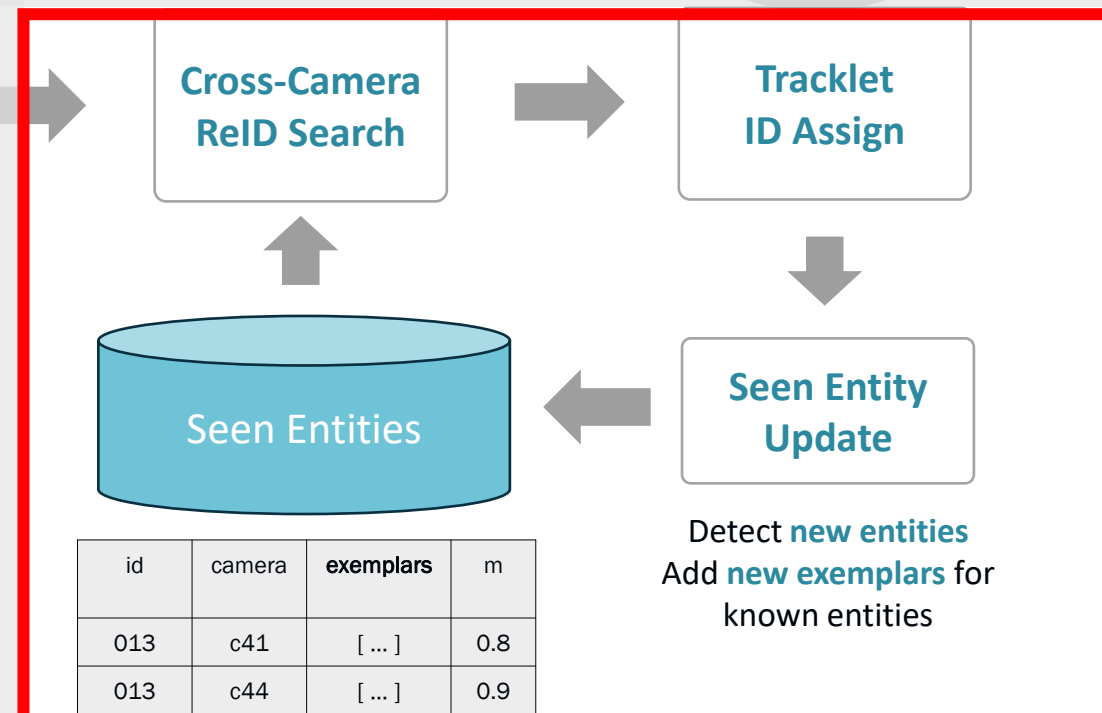
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ID-aware tracking
dramatically improves
tracking MOTA and high FPS

CenterNet (detection) +
Kalman (trajectory) +
Embedding Distance (identity)



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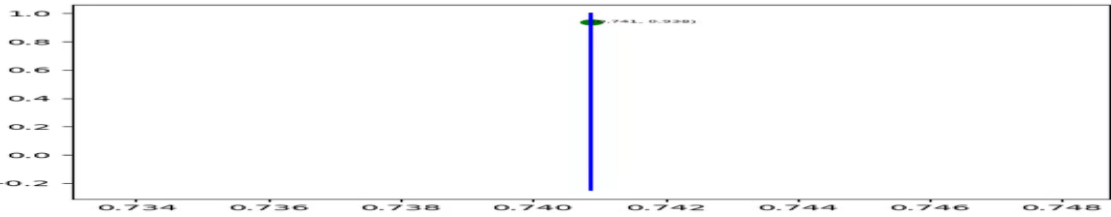
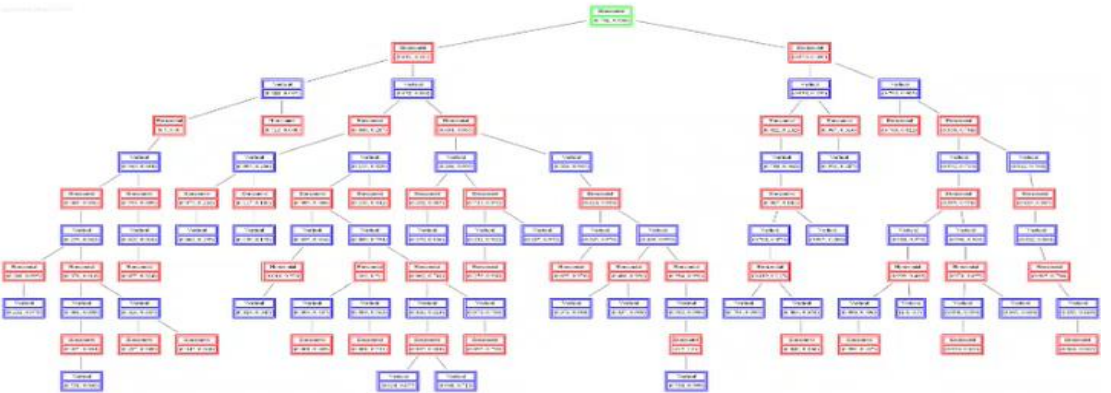
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ReID Search – Efficient Vector Search

KD-Tree Search – Efficient large database vector search

Partition tree at N level using $(N \bmod k)$ th vector dimension



from [DataScienceCentral](https://www.data-science-central.com/)

Tracklet ID - Assign to entire track (smooth per-frame noisy recognition)

t	2	3	4	5	6	7
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Cross-Camera ReID Search

Tracklet ID Assign



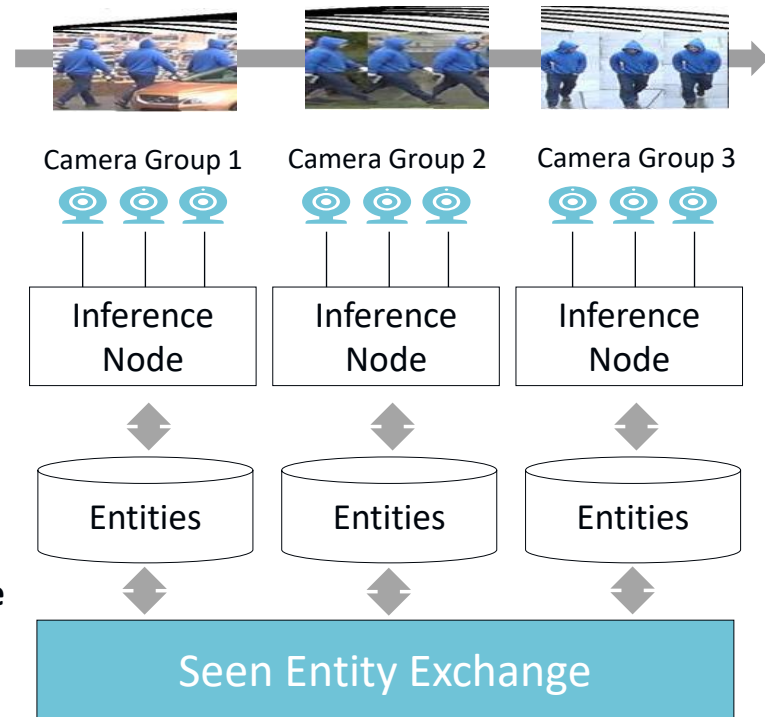
Seen Entity Update

id	camera	exemplars	m
013	c41	[...]	0.8
013	c44	[...]	0.9

Detect **new entities**
Add **new exemplars** for known entities

ReID Search Optimization

- Tracking requires **New IDs** to **propagate in real-time** across camera groups.
- **Multiple approaches** to balance **Recall** vs. **Compute Cost** vs. **Complexity**.
- Near-exchange provides best balance for community



Approach	Max IDs	REID Search	Seen Entity Exchange	Latency	Recall	Complexity
Node-Inner Don't exchange IDs	N	Linear Scan (MatMult)	None	Real-Time	Poor (mitigate by heavy nodes more cameras/node)	Trivial
Global Exchange All nodes get IDs	NxK	ANN Search (NxK large)	Periodic ANN Build	>30s	Good but with latency	Complex
Near Exchange Only physically near nodes get IDs	NxV	Linear Scan (MatMult)	Real-time	Real-Time	Good	Near-Trivial
Global + Near Best of both	NxK	ANN + Linear Scan (delta)	Periodic ANN Build + Real-time (delta)	Real-Time	Best	More Complex

N = Expected Max # People / Node K = Number Nodes V = Propagate to V-nearest nodes (complexity vs. recall trade-off)

Community Tracking Challenges

Similar
clothing



Workers
all look
the same



Old
analogue
cameras





Multi-Modal Identity Stamping

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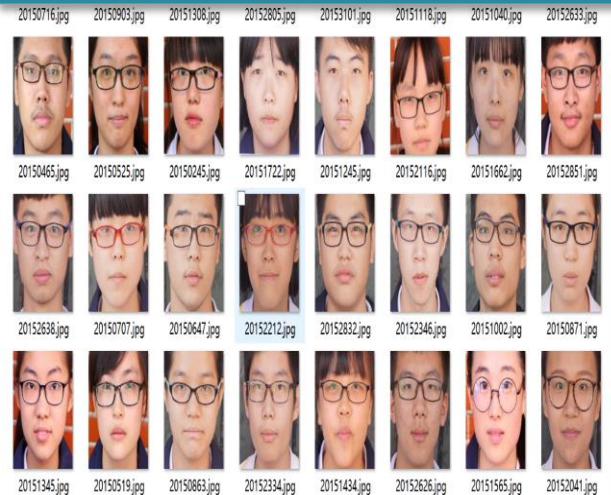
Easy



Frontal face, good lighting.

LFFW	Seedland	99.8
	Tencent	99.8
	Baidu	99.8
	Dahua	99.8

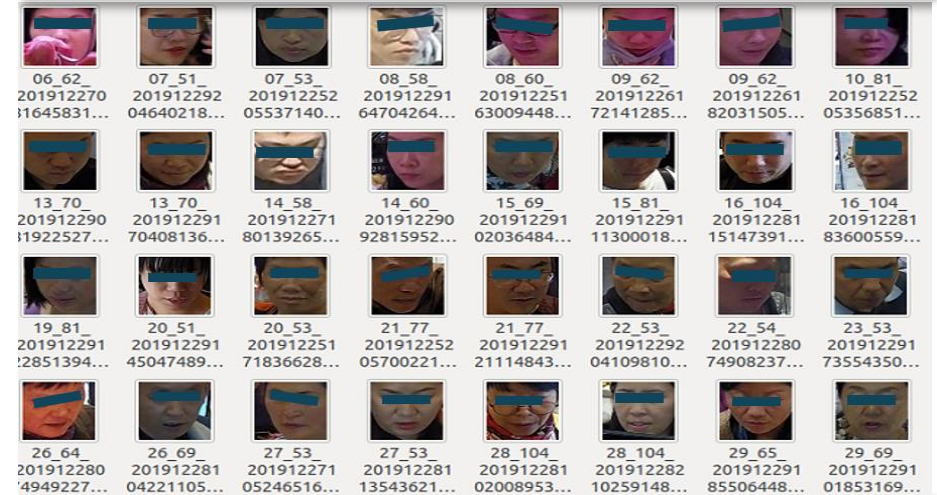
Challenging



Population bias e.g. Students

Students	Seedland	95.5@1e-6
	SenseTime '18	94.2@1e-6

Real Community Data

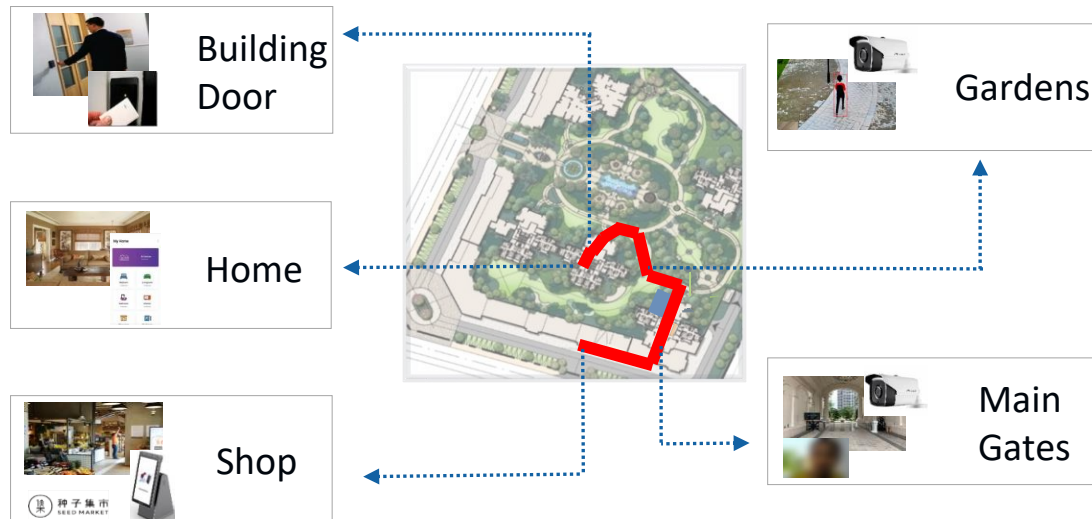


Lighting, angle, expression, occlusion, blur
very adverse! ~92% accuracy

Multi-Modal Identity Stamping

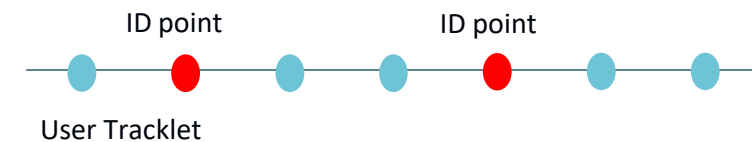
A typical user journey through the residential community has multiple touch points where user interacts with a device.

Multi-Modal identity stamping uses device interaction + visual features to convert the difficult identity task into an easier verification task.



Multi-Modal Identity Stamping

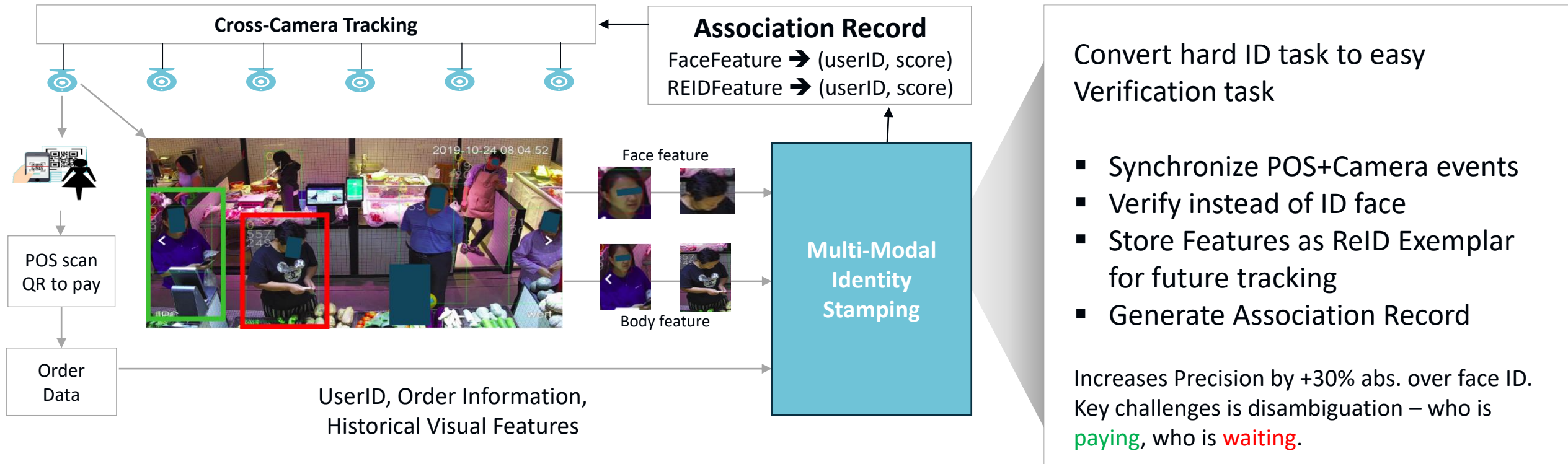
1. Use ReID to build cross-camera tracklet
2. Assign ID at key device interaction points
3. Stamp ID on tracklet after disambiguating disagreements



Examples of device Interactions in a user journey

- Face access systems
- Bluetooth pairing
- Swipe card access systems
- Shop POS payment

POS data for Multi-Modal Association



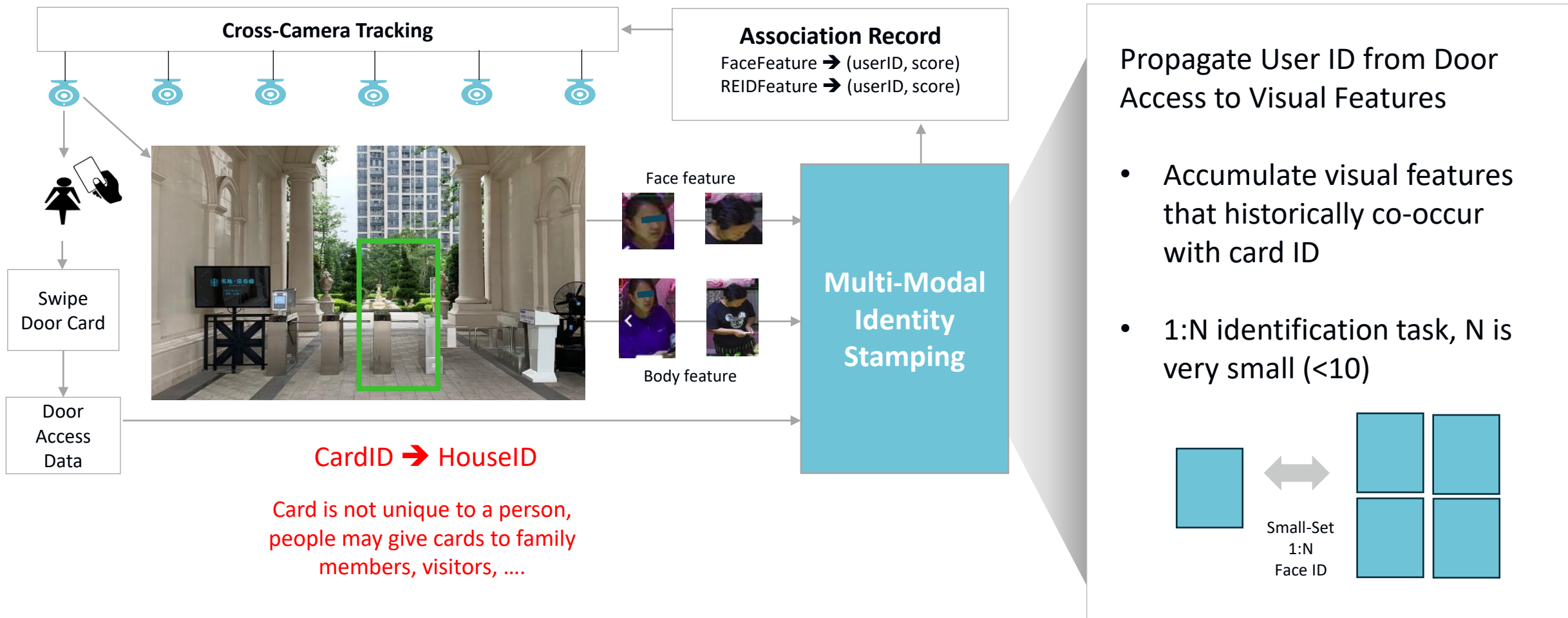
Camera Distance/Angle	Recall	Precision
Face ID only	55%	63%
Ceiling - 3m/20°	52%	90%
POS camera - 0.5m/10°	30%	93%



Retail Shop Community Tracking



POS data for Multi-Modal Association

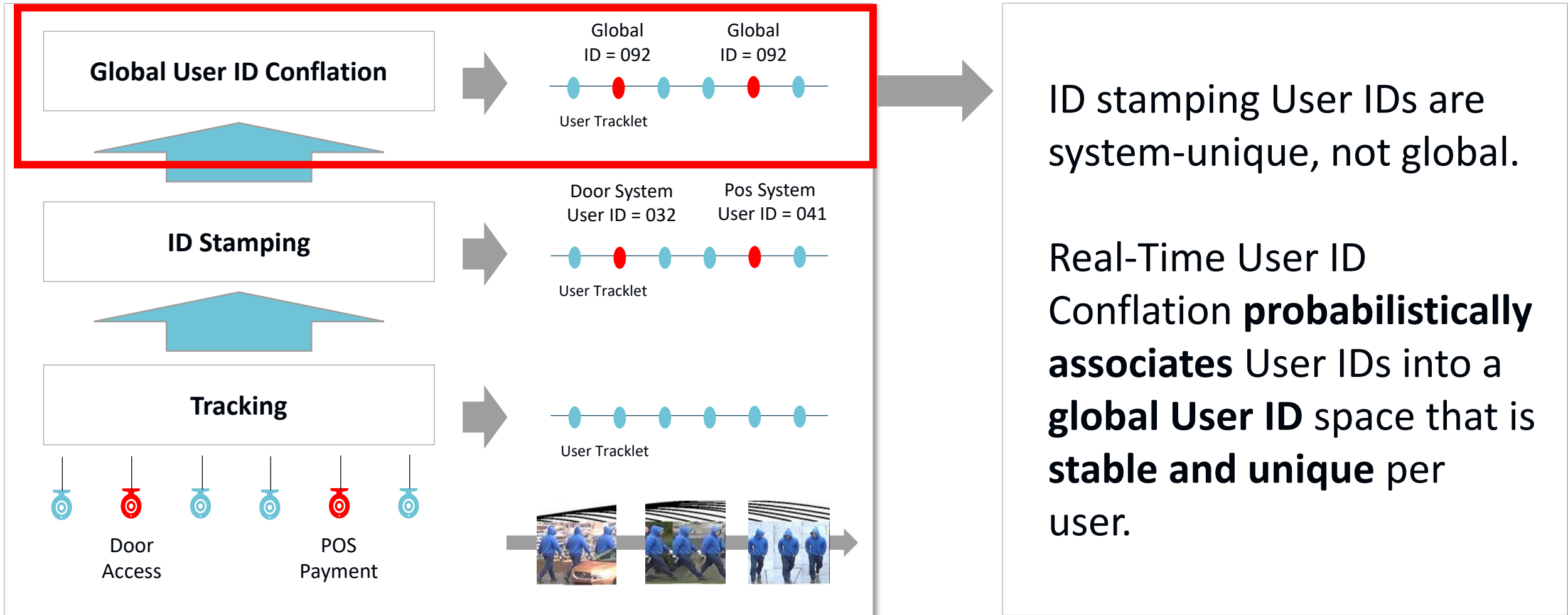




Global User ID Conflation

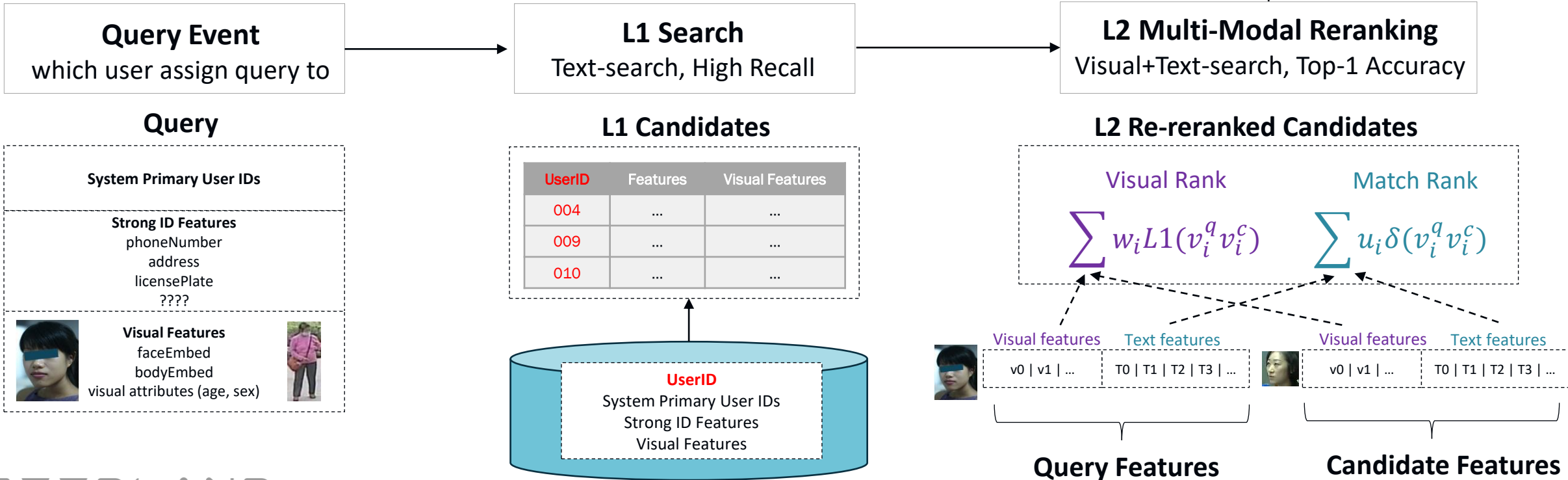
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Cross-Day Stable User ID Tracking

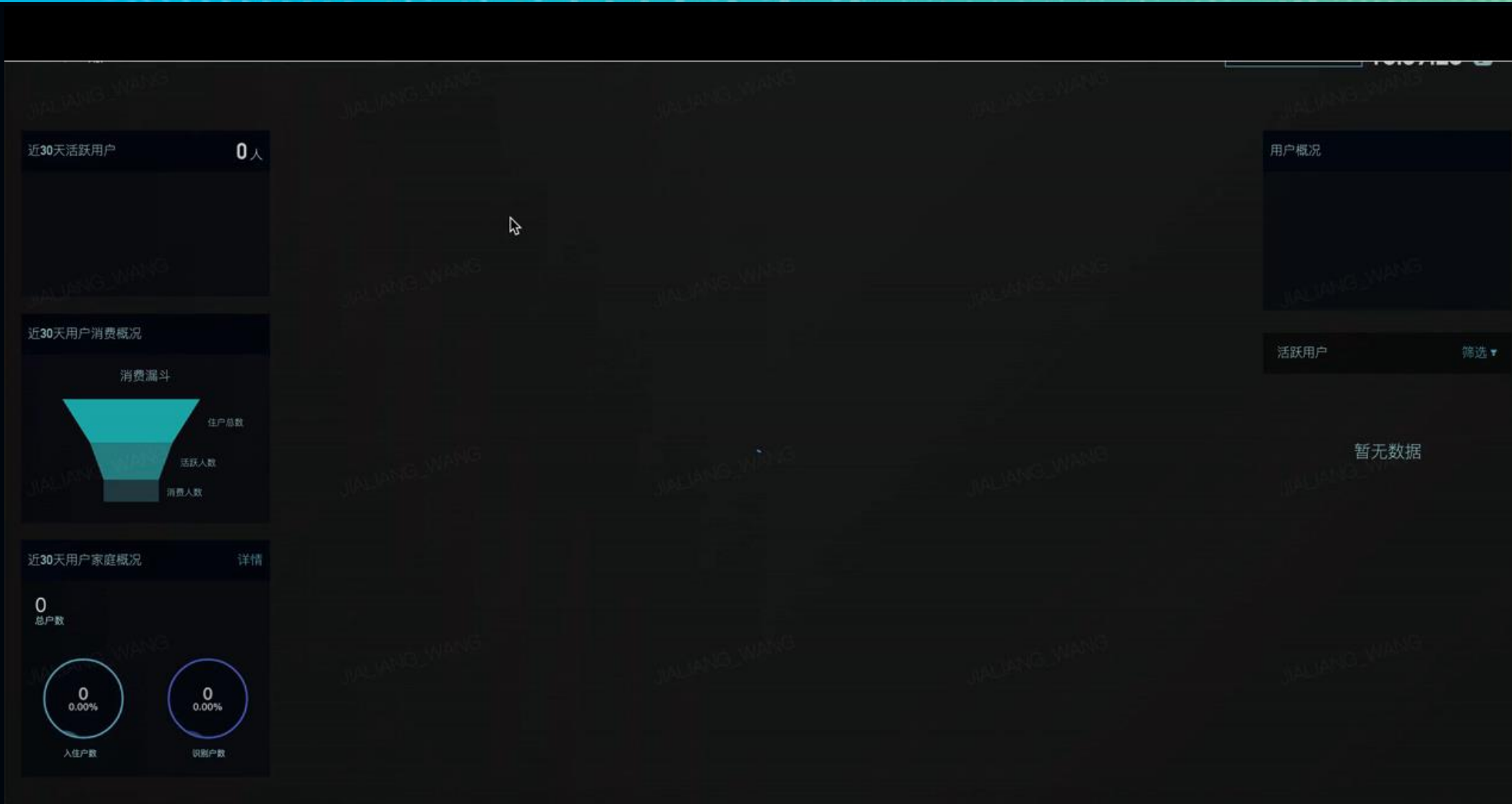


Real-Time User ID Conflation

Retrieval-based conflation does **real-time assignment** of a **Multi-Modal Query** event (text+visual) to a known set of stable **Global User IDs**. Multi-modal increases recall by **20%** at same precision.



Real-Time User Tracking and ID Conflation





Conclusions

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Residential communities have many opportunities to **transform** community, lifestyle and retail **with AIOT.**



Multi-modal tracking is a vehicle for intelligent user experiences with **significant benefits over vision-only.**



Privacy and **true value** for residents must be a **first-class citizen** when playing in the tracking space.