# From Sight to Insight: Visual Memory for Smarter Assistants

Xin Luna Dong, Meta Reality Labs GenAlRecP@KDD, 8/2025

This talk does not represent the company's point of view

# What Is An Ideal Personal Assistant?

#### Recommendation & Personalization

# **OBSERVE**

what you do in your life

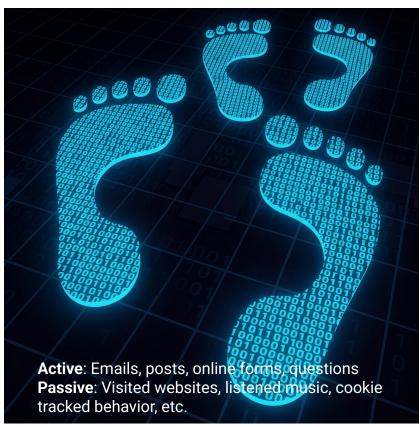
# **UNDERSTAND**

what you enjoy or find intriguing

PROVIDE

what you need tailored to your interest

### **Digital Footprints**



#### **Example: Social Network P13N**



## **OBSERVE**

what you read and post in social network

# **UNDERSTAND**

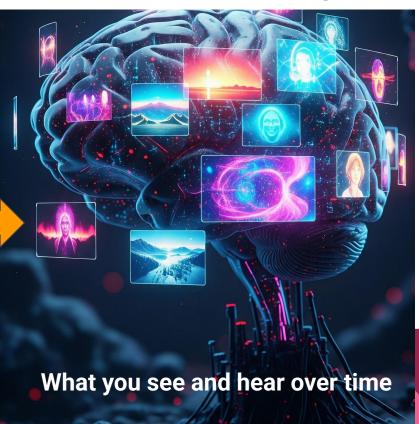
what you enjoy or find intriguing

PROVIDE what you may enjoy reading

#### **Digital Footprints**

# Active: Emails, posts, online forms, questions Passive: Visited websites, listened music, cookie tracked behavior, etc.

#### **Visual Memory**



#### What Can Visual Memory Provide In Addition?

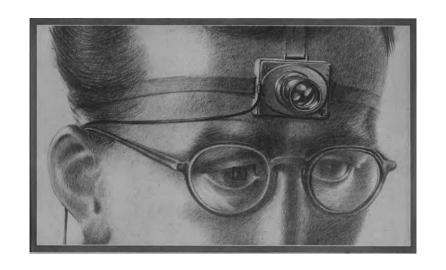








#### Visual Memory Is NOT IMPOSSIBLE





MEMEX (MEMory & EXpansion)
by Vannevar Bush (1945)

Wearables as great vehicles for life recording

#### Visual Memory QA & Personalization To The Property of the Prop



## **OBSERVE**

what you see and hear

# **UNDERSTAND**

what you enjoy or find intriguing



#### Challenges in Recording & Leveraging Vis. Mem.

#### **OBSERVE UNDERSTAND**

can be offline

#### **PROVIDE**

#### mostly on cloud

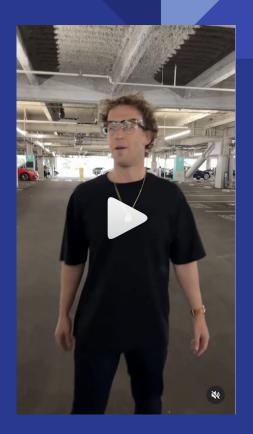
mostly on-device

- Hardware limitations
  - battery life
  - thermal constraints
  - storage capacity
  - transfer bandwidth

- Diverse and noisy history
  - o not relevant to a task
  - not accurately reflecting preferences
  - o not preferenceindicative or memory-worthy
- Special role of time and location

- Limited visual context windows for LLMs
- Runtime latency requirement
- Across-domain topics

# Application I. Memory QA



#### **Memory QA Examples**

| Memory Question                               | Memory Retrieval Results | Answer                         |
|---|--------------------------|--------------------------------|
| What's the restaurant<br>I saw last Thursday? | MIRATE                   | The restaurant is named Mirate |
| Which floor did I park?                       | 18                       | You parked on Floor 8.         |

#### **Memory QA—Remember This**



## **OBSERVE**

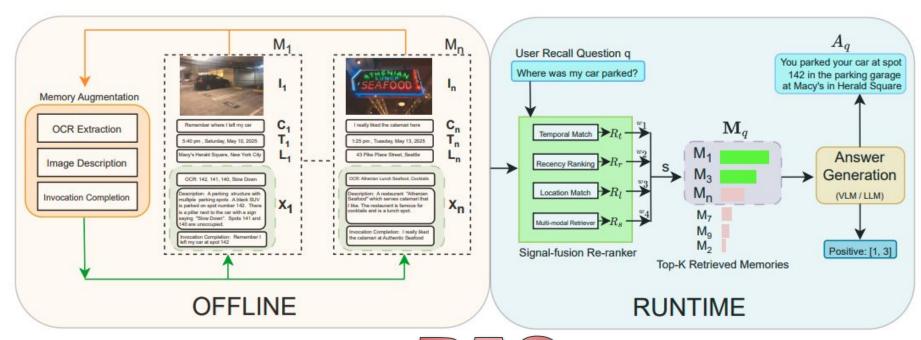
what you ask to remember

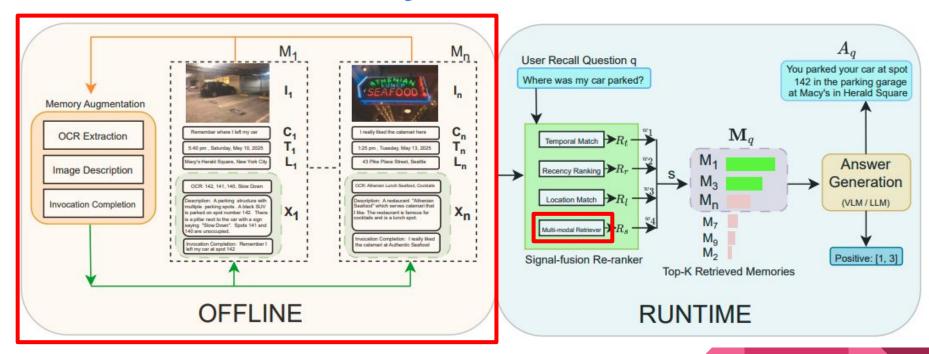
# **UNDERSTAND**

what is important to you

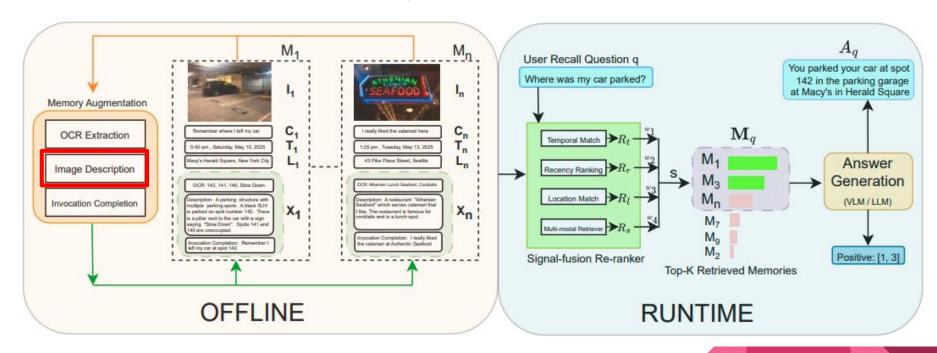
# PROVIDE

answers to your memory questions

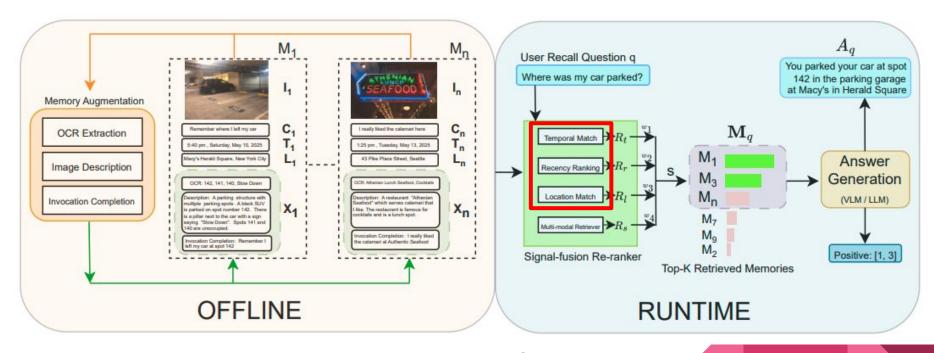




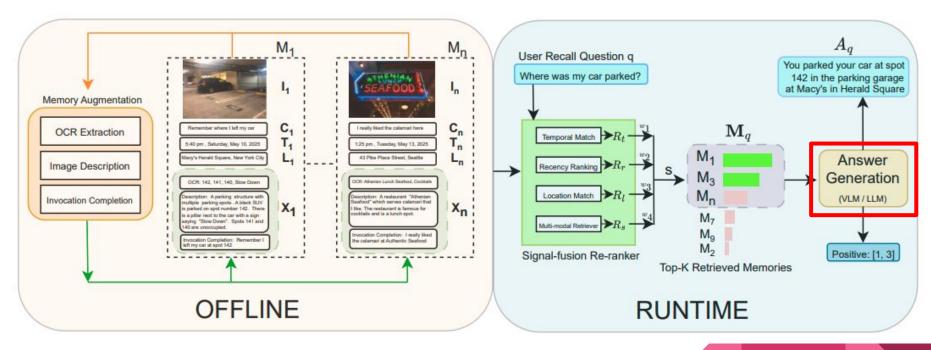
Better memory selection through offline augmentation & dual-modality retrieval



More targeted to memory questions through recall object prediction

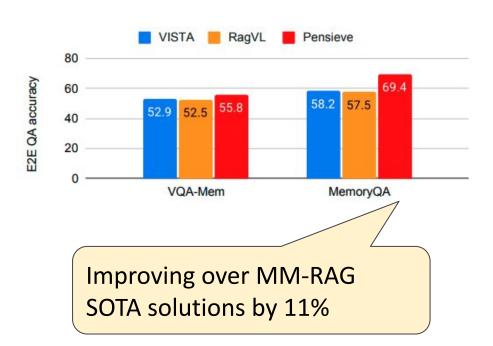


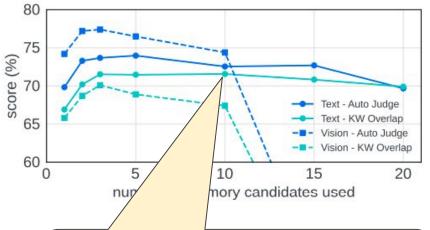
Better understanding of temporal & location requirements through temporal- & location-aware query rewriter and retrieval



Better answer generation and faster multi-memory aggregation through generation fine-tuning and text-based QA

#### The Pensieve Memory QA Quality





Text-based answer gen scales up to more candidate memories

#### **Memory QA—Pensieve Recap**



## **OBSERVE**

**User-triggered** 

# **UNDERSTAND**

Augmentation by identifying memorization points and capturing details

# PROVIDE

User-triggered RAG w. multi-memory aggregation

# Application II. P13N Through Visual History







#### **P13N Through Visual History Examples**

| Recommendation Question   | Retrieval Candidates   | Answer  |
|---|--|---|
| Which {museums, restaurants, parks, etc.} nearby should I visit?  User visual history | <i1> Vulcan Park and Museum, Statue symbolizing city's industries sits atop Red Mountain, surrounded by a park &amp; a museum. <i2> Sloss Furnaces Visitor Center, Blast furnace plant where iron was made from 1882–1971, now an arts &amp; education center with tours. <i3> Historic Cahaba Pumping Station, a working pump station located on the banks of the Cahaba River <i4> Alabama Jazz Hall of Fame, with a mission to foster, encourage, educate, and cultivate a general appreciation of the medium of jazz music</i4></i3></i2></i1> | You may enjoy the Cahaba Pumping Station, an active facility on the riverbank showcasing local waterworks.  Also consider Vulcan Park and Museum for industrial history, and Sloss Furnaces Visitor Center for a look at a preserved blast furnace. |
|   |  |   |

#### VisualLens—Personalized Recomm.



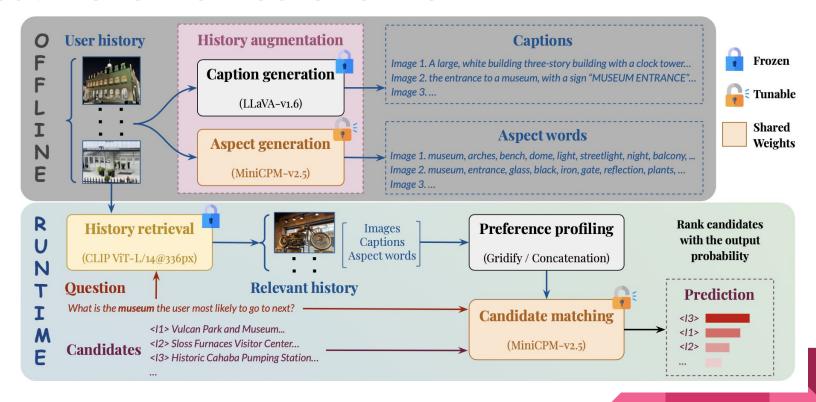
## **OBSERVE**

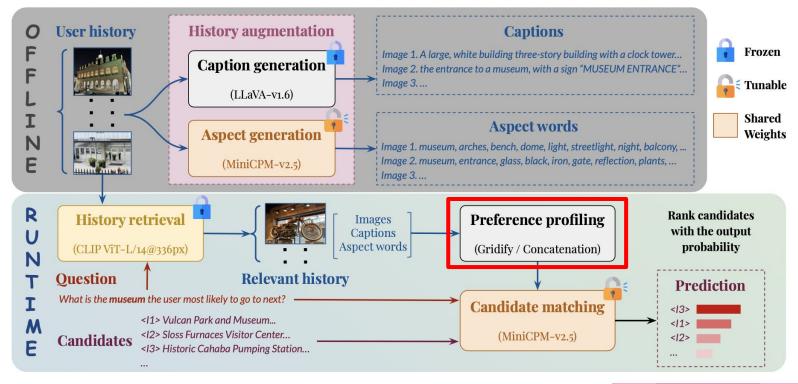
which photos you take

# **UNDERSTAND**

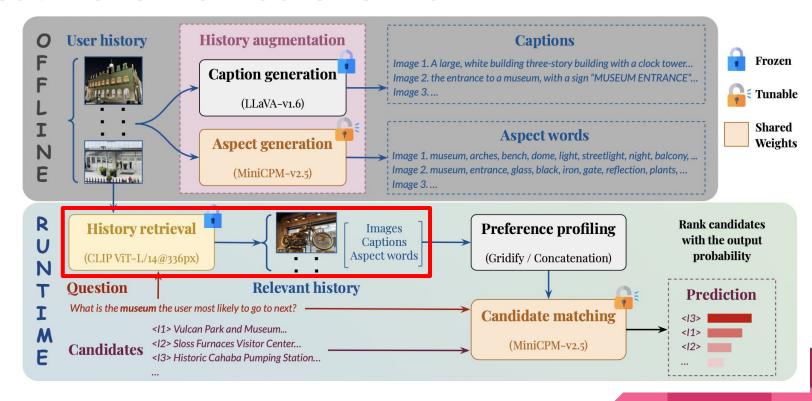
what you enjoy and find intriguing

# PROVIDE recommendations tailored to your interest

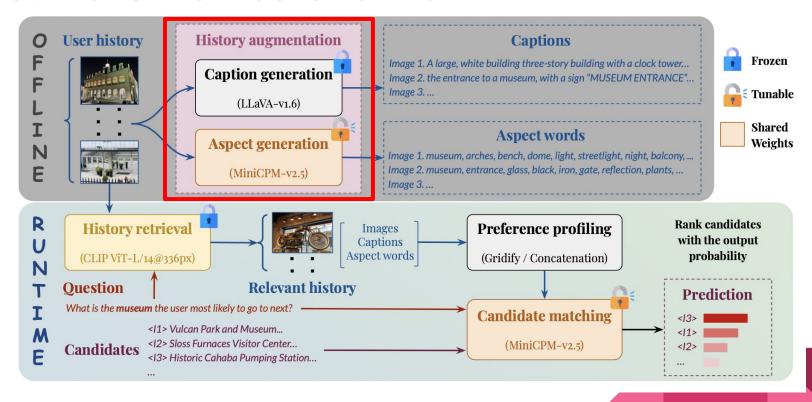




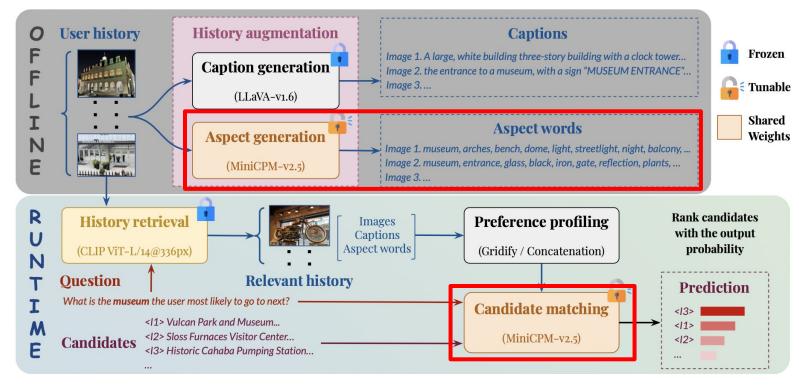
**Efficient generation through** grid-based multi-image inference



Relevant memories through history retrieval



Better signal capturing through offline augmentation



Better interest reflection through joint fine-tuning of aspect word extraction and recommendation

Filtered aspect



industrial, rust, pipes, towers, metal, structure, machinery, abandoned, overgrown, bluesky

Didesky

Extracted aspect

Aspect Q, Cand One image

VL validator aspects to predict the **museums** the user will go after?

Aspect Q1: What are useful

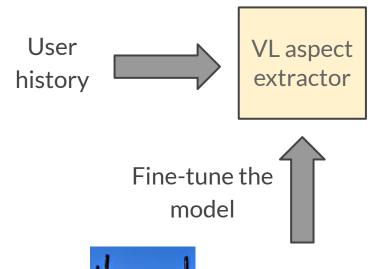
GT: Sloss Furnaces Visitor Center

Pred: industrial, rust, pipes, towers, metal, structure, machinery, abandoned

Aspect Q2: What are useful aspects to predict the **restaurants** the user will go after?

GT: Mugshots GMI and Bar -Birmingham, Al

Pred: industrial, metal, structure



industrial, rust, pipes, towers, metal, structure, machinery, abandoned



industrial, rust, pipes, towers, metal, structure, machinery, abandoned

VL aspect extractor

Extracted aspect

Fine-tune the model

Aspect Q, Cand One image



User

history

Filtered aspect



VL validator Aspect Q1: What are useful aspects to predict the **museums** the user will go after?

GT: Sloss Furnaces Visitor Center

Pred: industrial, rust, pipes, towers, metal, structure, machinery

Aspect Q2: What are useful aspects to predict the **restaurants** the user will go after?

GT: Mugshots Grill and Bar Birmingham, AL

Pred: industrial, metal, structure

industrial, rust, pipes, towers, metal, structure, machinery

Wang Bill Zhu, Deqing Fu, Kai Sun, et al. VisualLens: Personalization through Visual History. arXiv, 11/2024.



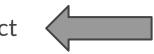
industrial, rust, pipes, towers, metal, structure, machinery

VL aspect extractor

Extracted aspect

Fine-tune the model

**Aspect Q**, Cand One image



VL validator

Aspect Q2: What are useful aspects to predict the restaurants the user will go after?

GT: Mugshots Grill and Bar -Birmingham, AL

Aspect Q1: What are useful

aspects to predict the

after?

Center

**museums** the user will go

GT: Sloss Furnaces Visitor

Pred: industrial, metal, structure, machinery

Pred: industrial, metal, structure



User

history

Filtered aspect

industrial, metal, structure, machinery



industrial, metal, structure, machinery

machiner

Extracted aspect

**Aspect Q**, Cand One image **GT** 

VL validator Aspect Q1: What are useful aspects to predict the **museums** the user will go after?

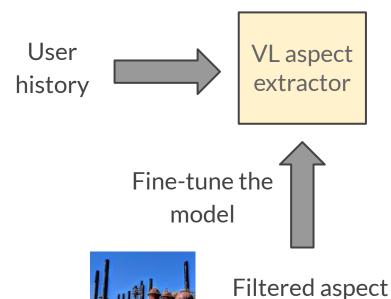
GT: Sloss Furnaces Visitor Center

Pred: industrial, metal, structure, machinery

Aspect Q2: What are useful aspects to predict the **restaurants** the user will go after?

GT: Mugshots Grill and Bar Birmingham. AL

Pred: industrial, metal



industrial, metal, structure, machinery

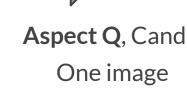
#### **Aspect Self-Converge (Neg)**



User VL aspect extractor history

man, sitting, table, wall, pictures, shadows, light, blue shirt, black pants, wooden chair

Extracted aspect



Fine-tune the model

**GT** 



Filtered aspect



validator

Aspect Q1: What are useful aspects to predict the **museums** the user will go after?

GT: Sloss Furnaces Visitor Center

Pred: man, sitting, table

Aspect Q2: What are useful aspects to predict the **restaurants** the user will go after?

GT. Mugchote Grill and Rar

Pred: table. chair

man, sitting, table

#### **Aspect Self-Converge (Neg)**



man, sitting, table

User history

VL aspect extractor

Extracted aspect

Fine-tune the model

**Aspect Q**, Cand One image

**GT** 



Filtered aspect



VL validator Aspect Q1: What are useful aspects to predict the **museums** the user will go after?

GT: Sloss Furnaces Visitor Center

Pred: man,

Aspect Q2: What are useful aspects to predict the **restaurants** the user will go after?

GT: Mugshots Grill and Bar - Birmingham, AL

Pred: bar, sitting

man, sitting

#### **Aspect Self-Converge (Neg)**



man, sitting

User history

VL aspect extractor

Extracted aspect

Fine-tune the model

**Aspect Q**, Cand One image

**GT** 



Filtered aspect



VL validator Aspect Q1: What are useful aspects to predict the **museums** the user will go after?

GT: Sloss Furnaces Visitor Center

Pred: industrial, iron, statue, landmark, historic

Aspect Q2: What are useful aspects to predict the **restaurants** the user will go after?

GT: Mugshots Grill and Bai - Birmingham, AL

Pred: bar, burgers, beer

man, sitting

#### VisualLens Effectiveness

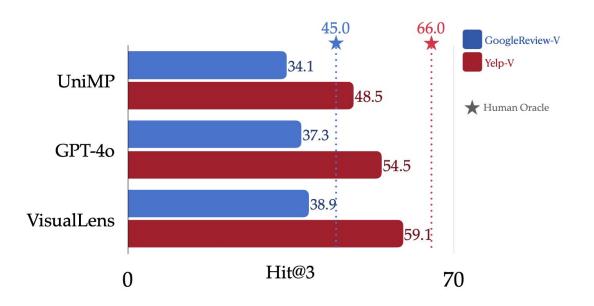
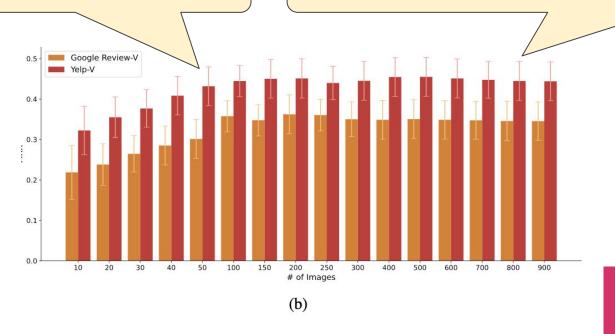


Figure 1. VISUALLENS leverages a user's task-agnostic visual history to provide personalized recommendations. Our method outperforms GPT-40 by 1.6% ~ 4.6% on Hit@3.

#### VisualLens Effectiveness vs. History Length

Quality improves with *richer* history

Robust against *longer* & *noisier* history



(b) MRR distribution over number of images. Both are on the User ID test set.

#### Personalization—VisualLens Recap



## **OBSERVE**

user-triggered, but not task-specific

# **UNDERSTAND**

augmentation by pinpointing user interest signals

# PROVIDE

Retrieve relevant memories as recommendation contexts

# Proactive Memory Capture & Compression

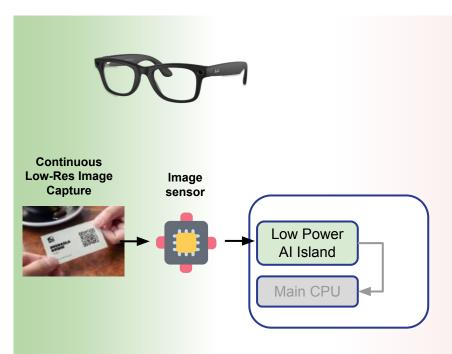
#### **Always-on Proactive Capturing**

# Continuous video recording

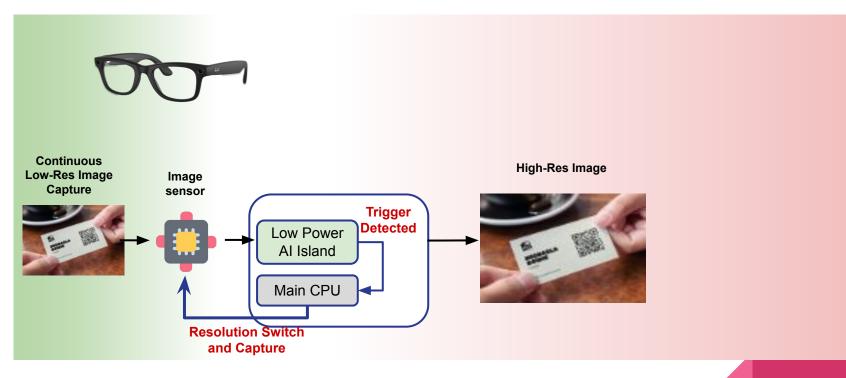
- **Challenge**: Hardware limitations
  - battery life
  - thermal constraints
  - storage capacity
  - transfer bandwidth
- Challenge: Repeated info and needle in a haystack
- Benefit: Lossless information

# Periodic low-resolution photo capture

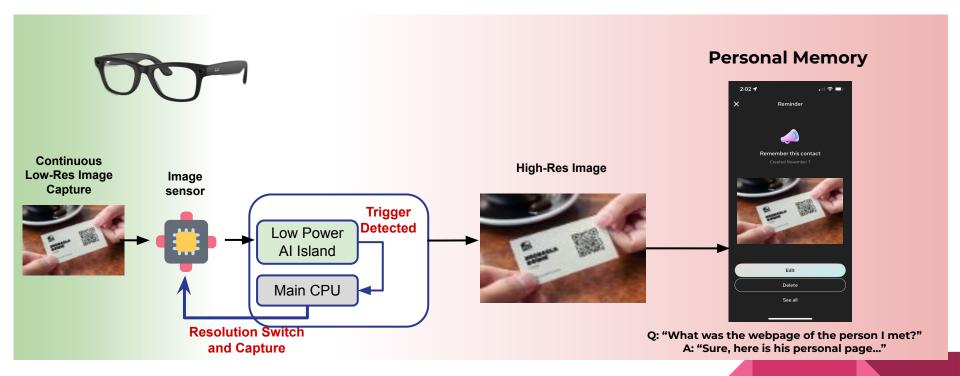
- Challenge: Lost information
  - o frames
  - details
- **Challenge:** Still repeated info
- Benefit: Saved hardware resources



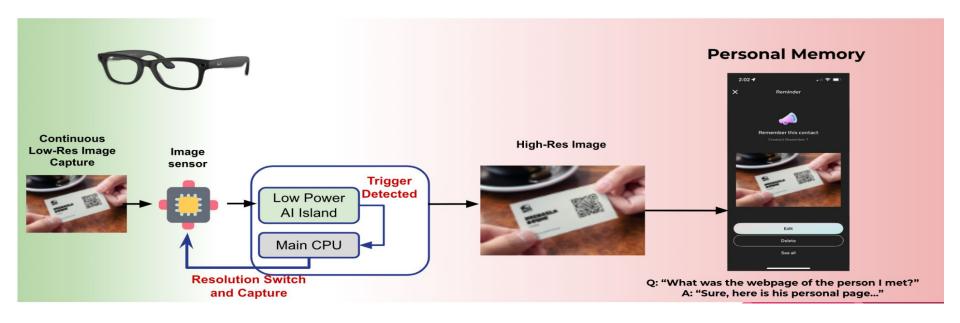
- Continuous low-resolution image capturing
- Event detection on low-power-Al-island



- Detected event wakes up main CPU
- CPU triggers high-resolution image capturing



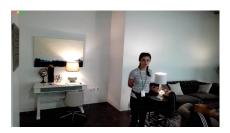
- High-resolution images transferred to personal memory
- Memory-QA on user's request



|                  | Model<br>Params | Model<br>Size | Precision<br>(HOI, AP50) | Precision*<br>(OCR, AP50) | Precision<br>(QR, AP50) |
|------------------|-----------------|---------------|--------------------------|---------------------------|-------------------------|
| Low power island | 1.78MB          | 1.88MB        | ~80%                     | ~74%                      | ~96%                    |
| Main CPU         | 14MB            | -             | ~88%                     | ~88%                      | -                       |

| Memory Usage    | 2.9MB |
|-----------------|-------|
| Running Latency | 74ms  |
| Running Power   | 180mW |

#### **Solution 2. Proactive Memory Compression**



Frame1



Frame4



Frame2



Frame5



Frame3



Frame6

#### **Solution 2. Proactive Memory Compression**



Frame1 - high



Frame4 - high



Frame2 - drop



Frame5 - low



Frame3 - low



Frame6 - high

#### **Proactive Captured Memory**



## **OBSERVE**

always-on event-triggered proactive capturing

# **UNDERSTAND**

memory compression to reduce memory usage

# PROVIDE

memory answers and recommendations

# Towards Ultimate Visual Memory Enhanced Smart Assistant

#### **Vision 1. Second Brain**

Build a second brain to offload thoughts and knowledge for memory search, personalization, and memoir







Memex (1945)





Automatic data recording
Digital footprint + Visual Memory

#### Vision 2. Smart Lens

Build smart lens to offer information relevant to what the user is seeing, saying, or asking, reactively and proactively



Free delivery on \$99+

#### **Future Research Directions**

#### **OBSERVE UNDERSTAND PROVIDE**

- Always-on event detection for proactive capturing
- Duplicate detection for capture and transmission suppression
- Memory compression to save space
- Memory organization and episode extraction for easier retrieval
- Seamless integration of multi-source memory
- Usage prediction for better augmentation
- Personality embedding for better recomm.

- Dense-memory search
- Seamless blending of personal and public knowledge
- Personalized LLM generation
- Proactive information offering

#### **Take-Aways**

- Wearables devices provide a great vehicle for life recording and for personalized assistance
- Visual memory allows interesting memory QA (Pensieve) and personalized recommendation (VisualLens)
- We envision building Second Brain and Smart Lens based on dense visual memory

KDDCup Workshop—Comprehensive RAG Multi-modal Multi-turn Challenge. 8:00-12:00.

Towards a Knowledgeable Assistant: A Federated RAG Approach. MLoG, 9:55-10:30.

Managing Data, or Letting Data Manage Themselves. SKnowLLM, 3:00-3:30. Room: 717

### Thank You!