

## Research Objective

An age-adapted video saliency algorithm that can predict the relative saliencies of billboards in a street video for adult and elderly

## Contributions

1. An eye-gaze data collected for adults and elderly while viewing a street video in free-viewing and task viewing.
2. Suitable metrics to analyze the collected data to reveal the age-related differences.
3. An age-adapted video saliency algorithm which predicts the relative saliencies of the billboards

## Materials

### Participants and Stimuli:

- 30 participants - 15 participants - mean age 24 and 73.1 each
- Stimuli - 2 min 30 sec street video full of restaurants.
- Tobii x2-60 eye tracker used for recording the eye-gaze data,
- fixations and saccades were detected by Tobii fixation filter.

### Procedure and Task:

- Participants were divided into two groups of 15 (mix of both age groups)
- Group 1 watched the video in free viewing, and group 2 watched it in task viewing mode
- Task – search a place for lunch



Figure 1: Illustration of the experiment setup

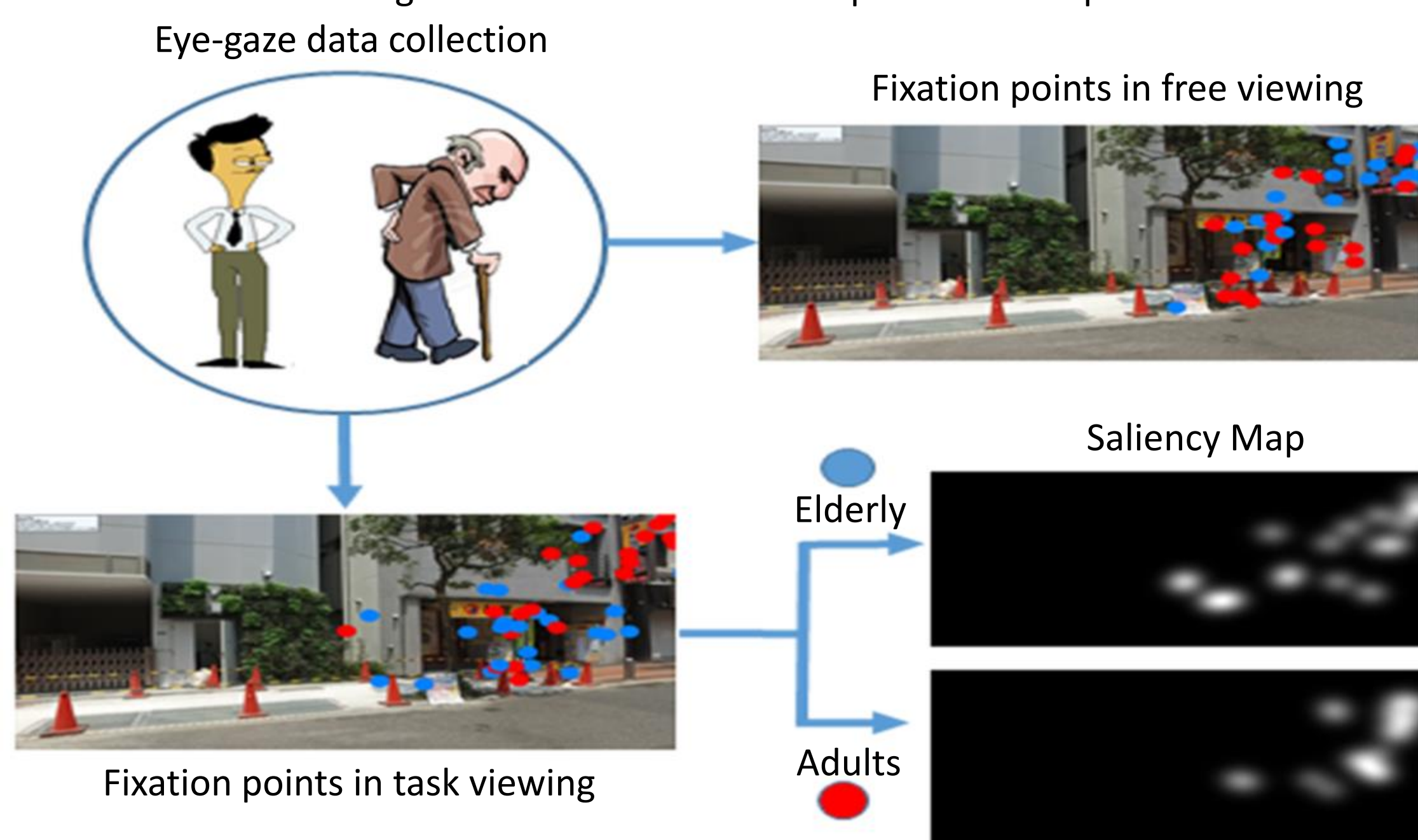


Figure 2: Process of generating fixation map and saliency map

## Methodology



Figure 3: Visualization of different tendencies of gaze landings around the billboards in free viewing for adults (first row) and elderly (second row) participants



Figure 4: Few instances of manually labelled billboards (only restaurants).

Metrics used to quantify the age-related differences

1. Total fixation count
2. Explorativeness
3. Center bias.

$$\text{Explorativeness} : H(I_{A/E}^{task}) = \sum_l h_{I_{A/E}^{task}}(l) * \log(L / h_{I_{A/E}^{task}}(l)) \quad (1)$$

$$H(I_{A/E}^{free}) = \sum_l h_{I_{A/E}^{free}}(l) * \log(L / h_{I_{A/E}^{free}}(l)) \quad (2)$$

- Result :
  - Free viewing - adults are significantly more explorative than elderly
  - Task viewing - both age groups showed similar tendency.

Center bias - elderly have higher center bias than adults in both modes

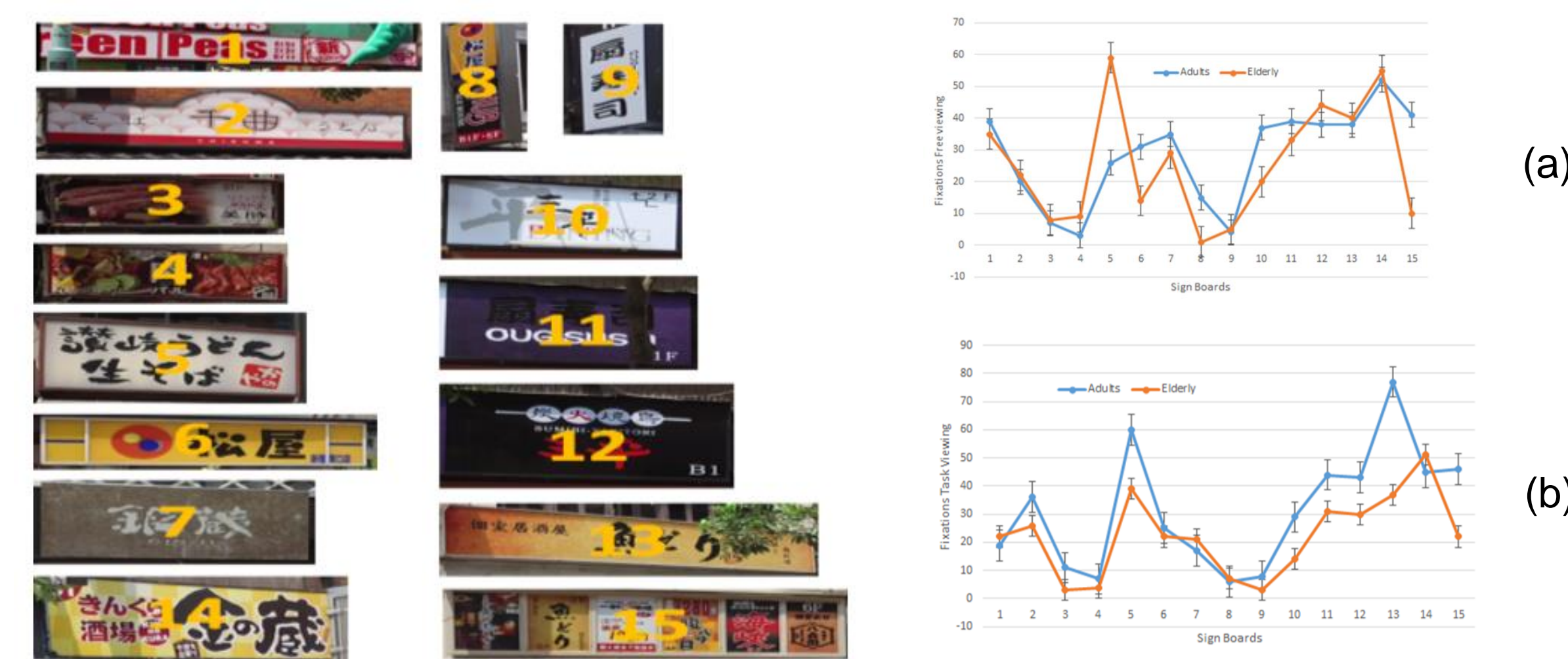


Figure 5: Annotated billboards in the street video

## Saliency Model

We have developed a learning based model, based on idea proposed in a previous study [6] with following changes to adapt with the age-related differences:

- Selected optimum set of scales for different features (bottom, mid, top-level features) extracted for adults and elderly.
- The models weights for feature combination are learned for the adults and elderly participant's gaze separately.
- Center-maps are tuned according to age-related differences in center-bias tendency.

Experiment setup:

- The top 10% strong positive from top 5% of the ground truth saliency map and similarly 10% strong negative pixels from bottom 20% of the map were selected
- For a given set of features and levels (positive and negative samples) SVM is used to learn the optimal weights i.e., model parameters for adults and elderly.

$$SM_A(I_i) = w_A F^T(I_i) + b_A \quad (3)$$

$$SM_E(I_i) = w_E F^T(I_i) + b_A \quad (4)$$

Table 1: The prediction accuracy (AUC score) billboards for adults and elderly participants

		Free Viewing				Task Viewing					
Board Saliency		Highly Salient	Least Salient		Board Saliency		Highly Salient	Least Salient			
Board number	(adults)	5	14	4	9	Board number	(adults)	5	13	8	9
Board number	(elderly)	5	14	8	9	Board number	(elderly)	5	14	4	9
Itti [1]	adults	0.87	0.65	0.78	0.60	Itti [1]	adults	0.76	0.69	0.72	0.72
	elderly	0.75	0.67	0.73	0.63		elderly	0.75	0.77	0.60	0.68
GBVS [2]	adults	0.83	0.71	0.83	0.57	GBVS [2]	adults	0.78	0.76	0.63	0.58
	elderly	0.79	0.70	0.58	0.57		elderly	0.71	0.72	0.81	0.65
s_map [3]	adults	0.74	0.73	0.82	0.58	s_map [3]	adults	0.70	0.73	0.60	0.57
	elderly	0.72	0.71	0.62	0.61		elderly	0.83	0.71	0.70	0.71
m_map [4]	adults	0.67	0.64	0.65	0.68	m_map [4]	adults	0.78	0.73	0.76	0.59
	elderly	0.65	0.60	0.69	0.56		elderly	0.69	0.70	0.70	0.59
e_map [5]	adults	0.71	0.65	0.71	0.55	e_map [5]	adults	0.65	0.66	0.56	0.53
	elderly	0.66	0.66	0.54	0.58		elderly	0.58	0.78	0.58	0.61
Ours	adults	0.83	0.79	0.67	0.65	Ours	adults	0.80	0.76	0.69	0.63
	elderly	0.78	0.72	0.63	0.68		elderly	0.75	0.77	0.72	0.69

## Conclusions

- Adults and elderly while viewing street videos have difference gaze landings during free viewing and task viewing as has been verified by our metrics.
- The proposed model outperforms others in predicting saliency for billboards in paved areas of street videos for different age groups.

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