

Where are they looking?



But why?







# Goal

- Perform gaze-following in natural settings.
- Only a single view available.
- Without restrictive assumptions.

# Related work

- Saliency:
  - Mainly fixations of an observer free-viewing pictures.
- Gaze:
  - Only estimate gaze direction.
  - Rely on face detection (so only work for people facing the camera).



# GazeFollow: the dataset

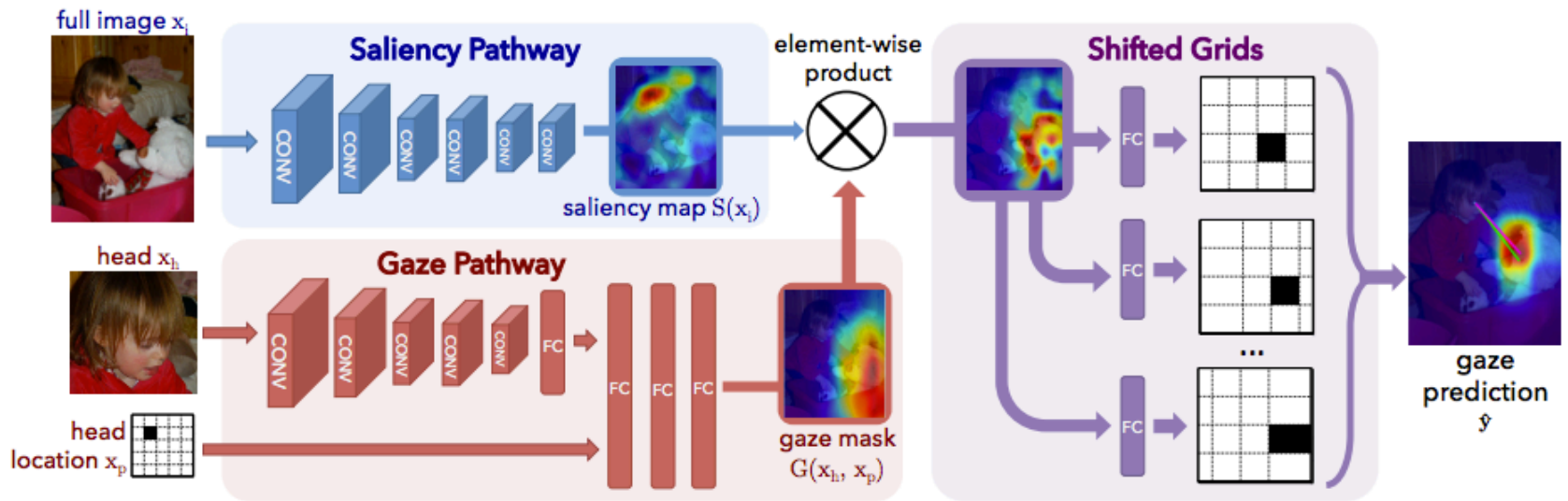


**ENOUGH SETUP**



**SHOW ME HOW THEY DID IT!!!**





$$\hat{y} = F(G(x_h, x_p) \otimes S(x_i))$$

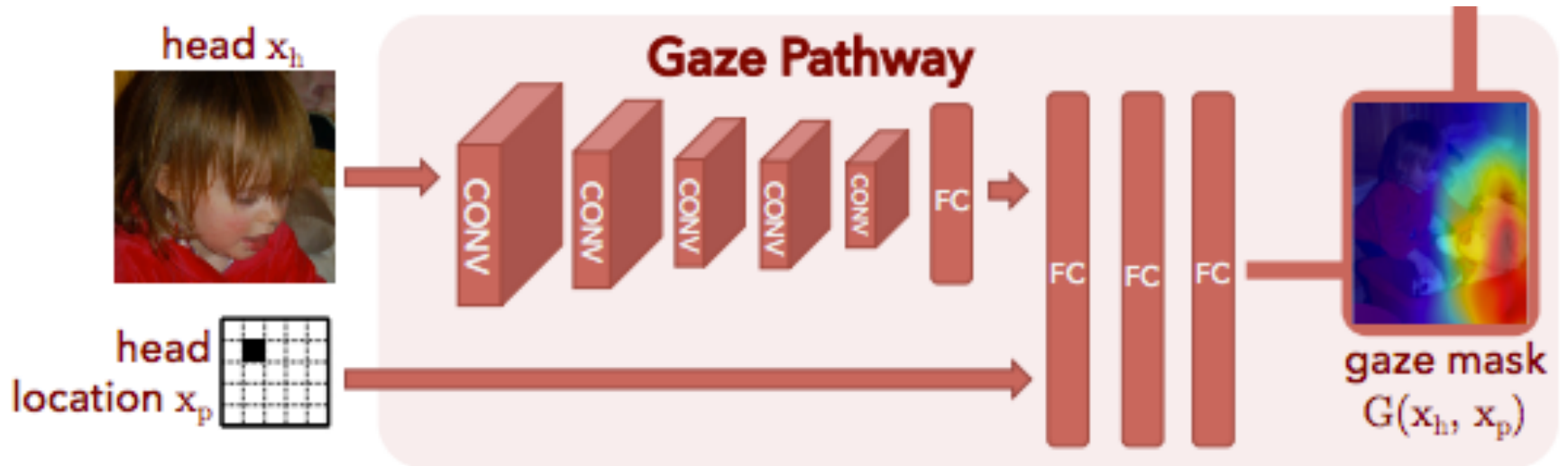
# Saliency map

full image  $x_i$



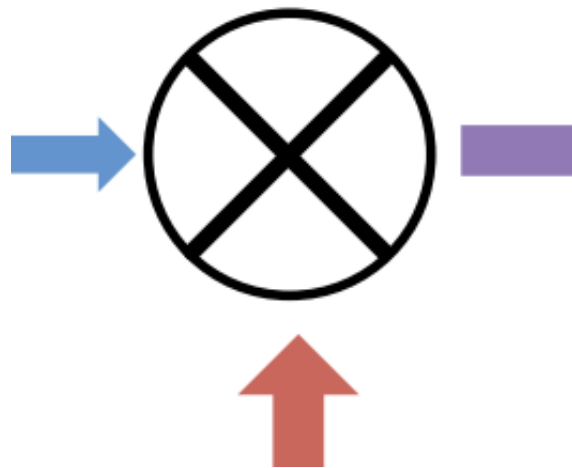
$$S(x_i)$$

# Gaze mask



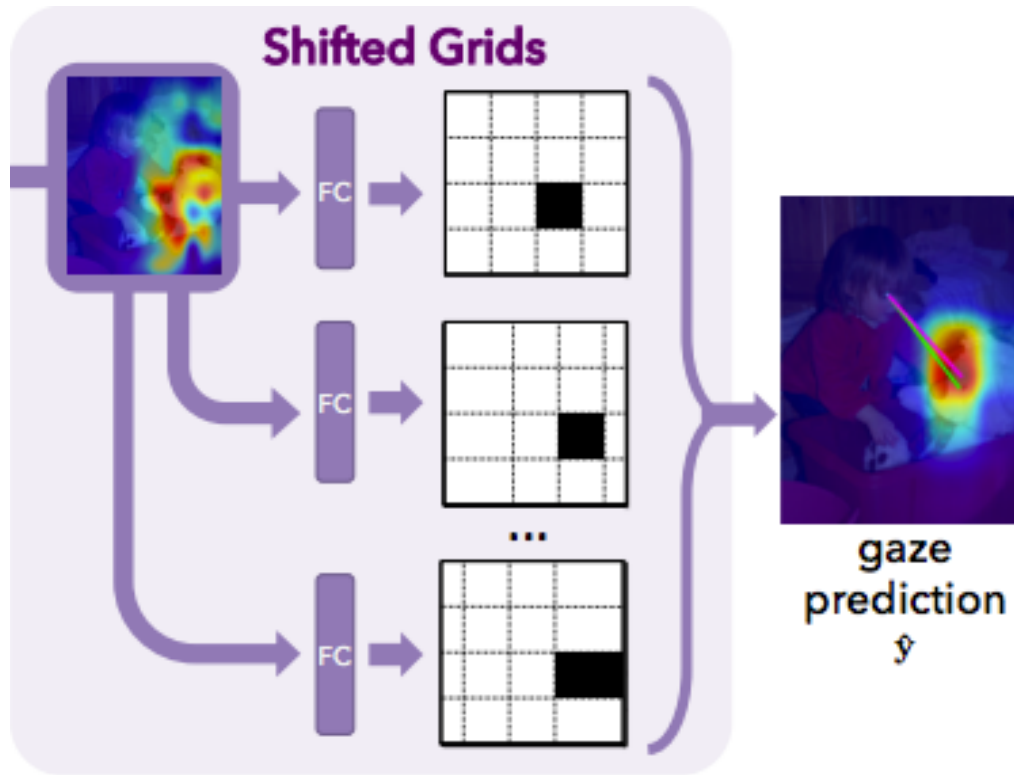
$$G(x_h, x_p)$$

element-wise  
product



$$G(x_h, x_p) \otimes S(x_i)$$

# Shifted grids



$$F(G(x_h, x_p) \otimes S(x_i))$$



# Quantitative results (pears to apples)

Model	AUC	Dist.	Min Dist.	Ang.
Our	<b>0.878</b>	<b>0.190</b>	<b>0.113</b>	<b>24°</b>
SVM+shift grid	0.788	0.268	0.186	40°
SVM+one grid	0.758	0.276	0.193	43°
Judd [11]	0.711	0.337	0.250	54°
Fixed bias	0.674	0.306	0.219	48°
Center	0.633	0.313	0.230	49°
Random	0.504	0.484	0.391	69°
One human	0.924	0.096	0.040	11°

# Ablation study (removing stuff)

Model	AUC	Dist.	Min Dist.	Ang.
No image	0.821	0.221	0.142	27°
No position	0.837	0.238	0.158	32°
No head	0.822	0.264	0.179	41°
No eltwise	0.876	0.193	0.117	25°
5 × 5 grid	0.839	0.245	0.164	36°
10 × 10 grid	0.873	0.218	0.138	30°
L2 loss	0.768	0.245	0.169	34°
Our full	0.878	0.190	0.113	24°

# Qualitative Results (pew pew pew)





# Visualization of internal representations

input image

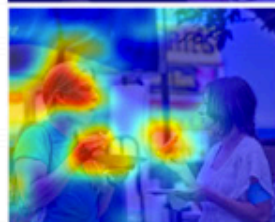
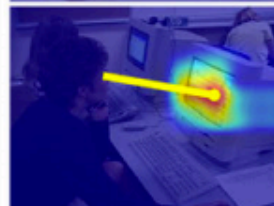
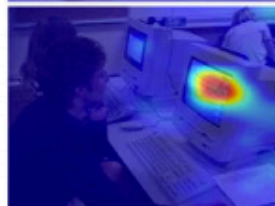
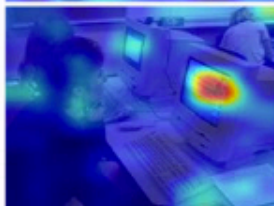
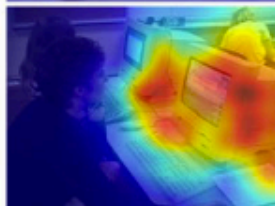
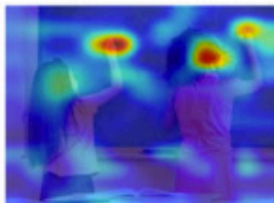
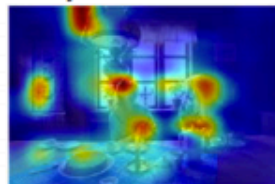
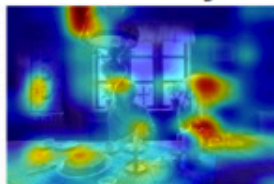
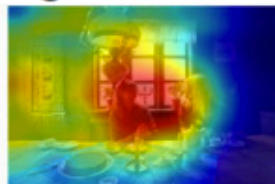
gaze mask

saliency

product

output

prediction



# Critique

## Negatives:

- Not using “just one image as input” (but they do in the demo).
- In the quantitative analysis, they compare with methods that solve different problems.



# Critique

## Positives:

- Awesome new dataset.
- Nice first attempt (and baseline for future work).
- Lots of room to improve.
- Great as a future project.

Thanks!