Visual Attention Analyzer

Al-powered Chrome extension for advanced webpage attention analysis

Presented By Chandra Prakash S

Problem Statement

- Web designers and marketers struggle to predict where users will focus
- Traditional eye-tracking studies are expensive and time-consuming
- Need for accessible tools to understand visual attention patterns
- Lack of solutions combining visual and structural webpage analysis



Solution Overview



01.

Chrome extension that analyzes visual attention patterns on webpages

02.

Combines computer vision with DOM analysis for comprehensive results

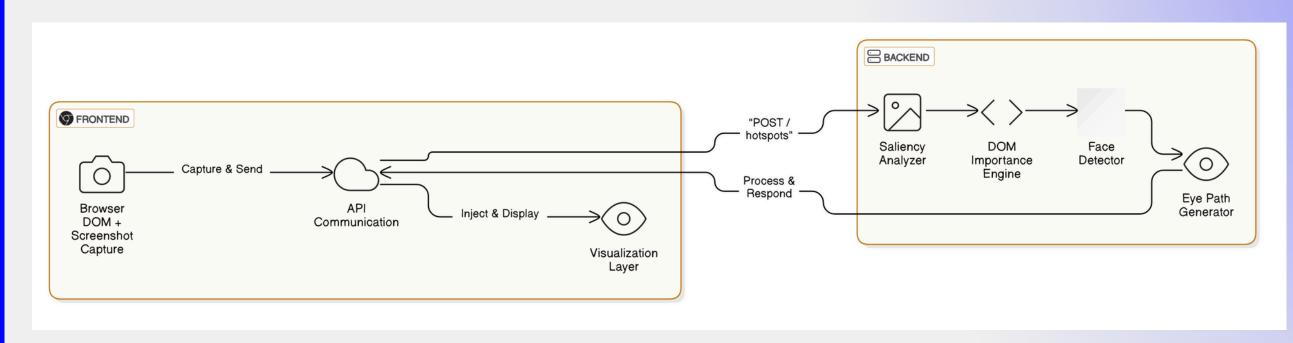
Provides interactive visualizations overlaid on the actual webpage

03.

Makes advanced attention analytics accessible to everyone

Technical Architecture

- Frontend: Chrome Extension APIs (content scripts, background workers)
- Screenshot Capture: Chrome tabs API for full page capture
- DOM Analysis: JavaScript parsing of page structure and attributes
- Backend Processing: FastAPI Python server with computer vision algorithms
- Data Flow: Screenshot + DOM data → API → Visualization overlays





Spectral Residual Saliency Detection:

Identifies visually distinctive regions

Center Bias Application:

Accounts for natural human tendency to focus on center

MediaPipe Face Detection:

Prioritizes human faces in attention analysis

Hybrid Scoring System:

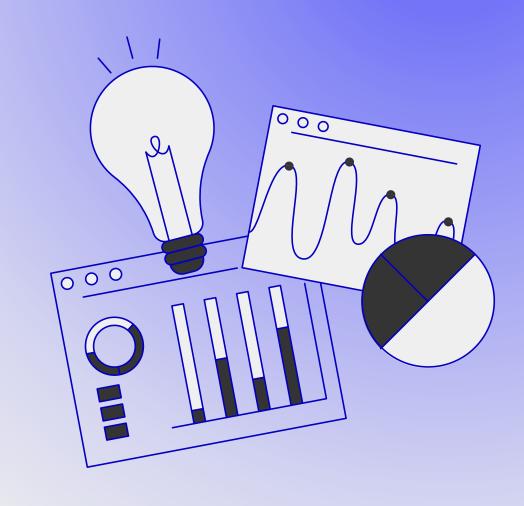
Combines visual saliency with DOM importance

Non-Maximum Suppression:

Merges overlapping regions for cleaner results

Visualization Features

- Hotspot Regions: Color-coded boxes with rank indicators
- Eye Movement Path: Numbered points with directional lines
- Face Detection: Purple outlines identify faces as attention anchors
- Attention Heatmap: Color-gradient overlay of attention distribution
- Interactive Controls: Toggle visibility of different visualization layers





Code Deep Dive: Backend

- FastAPI Framework: Efficient, async Python web server
- Key Components:
 - Saliency detection with OpenCV
 - Center bias application for natural viewing patterns
 - DOM importance calculation (tag weights, position, content)
 - Face detection integration with **MediaPipe**
 - Hotspot scoring algorithm that balances visual and semantic signals



Code Deep Dive: Frontend

• Extension Components:

- Content script for DOM analysis and visualization rendering
- Background script for screenshot capture
- o Interactive control panel for visualization toggles

Enhanced DOM Data Extraction:

- Element positions, attributes, and styles
- Interactive element detection
- Visibility checking and DOM depth calculation



UX designers optimizing webpage layouts



04

Accessibility specialists identifying focus issues

02

Marketing teams improving ad and CTA placement

Real-World Applications

Exploring creativity

Content creators enhancing visual hierarchy





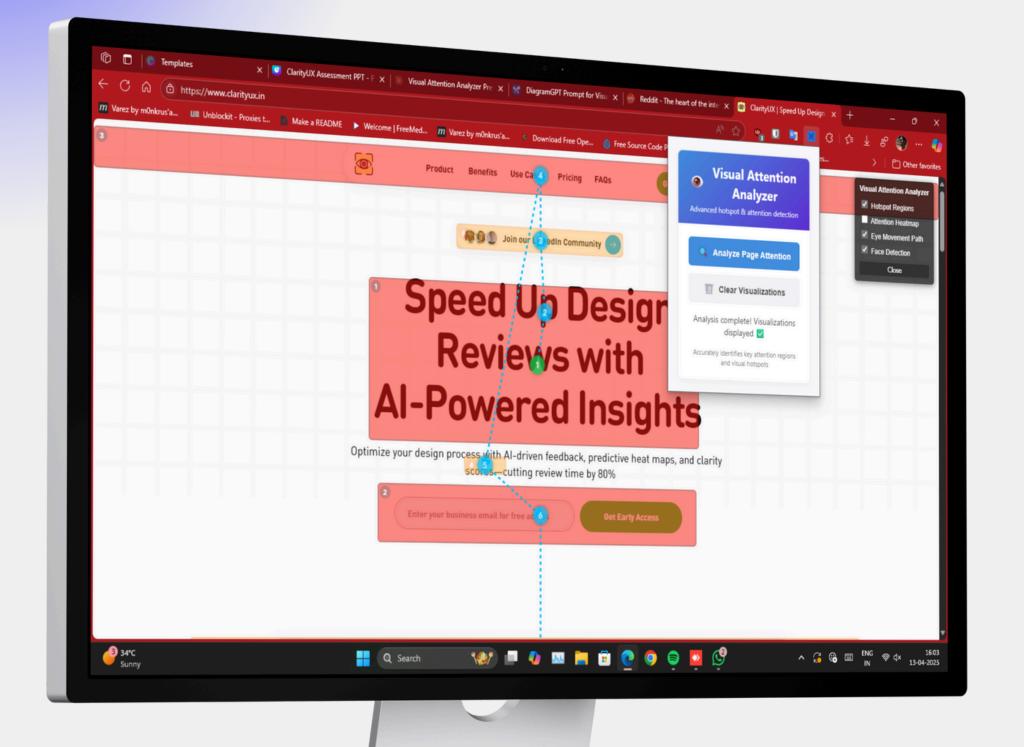
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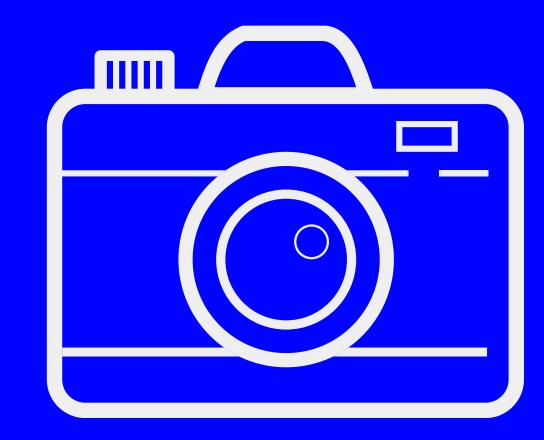
E-commerce sites optimizing product displays

06

News sites arranging content for maximum engagement

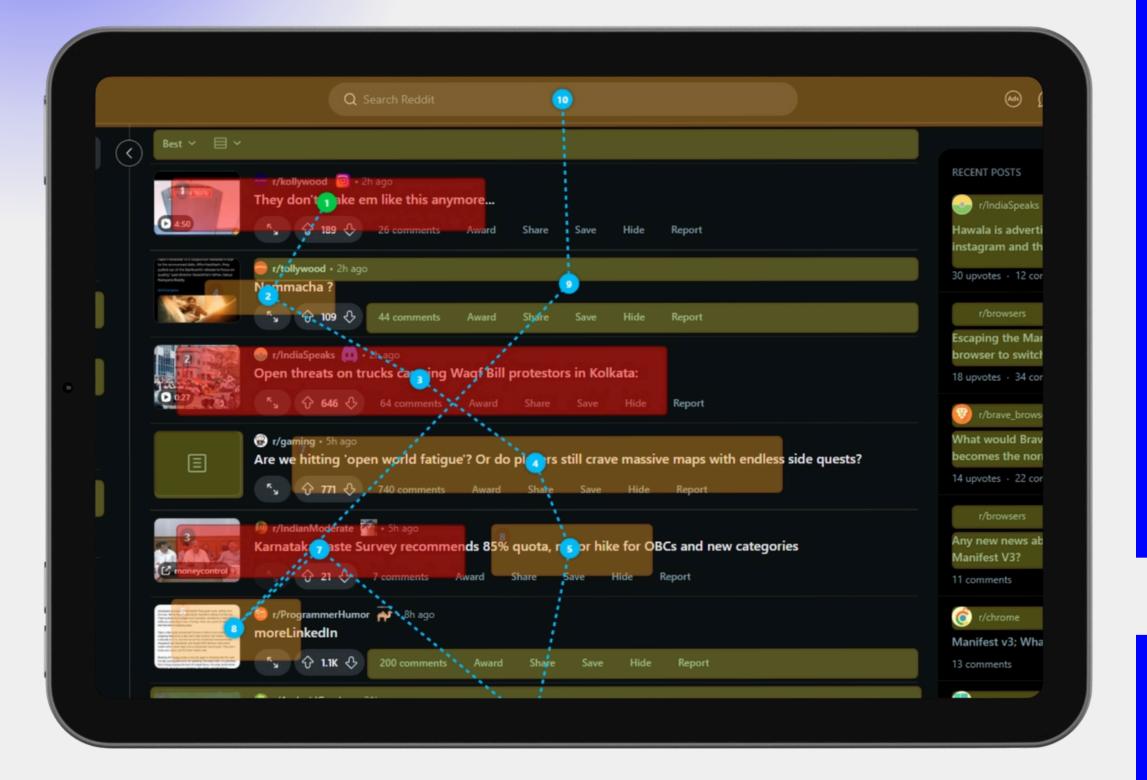
Screenshots

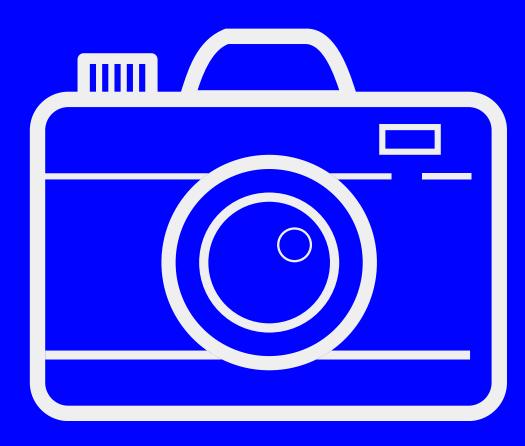


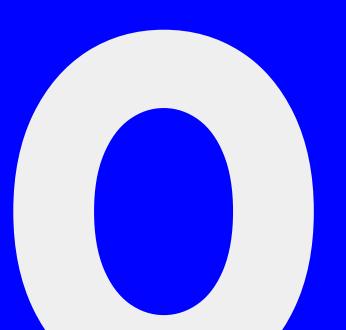




Screenshots







Code Snippet

```
saliency = cv2.saliency.StaticSaliencySpectralResidual_create()

# 2. COMPUTE SALIENCY MAP
(success, saliency_map) = saliency.computeSaliency(img)
if not success:
    raise HTTPException(status_code=500, detail="Saliency computation failed")

# 3. APPLY CENTER BIAS TO SALIENCY MAP (people tend to look at center)
saliency_map = apply_center_bias(saliency_map, img.shape)

# 4. APPLY ADAPTIVE THRESHOLDING
thresh_value = saliency_map.mean() + 1.5 * np.std(saliency_map)
    _, thresh_map = cv2.threshold(saliency_map, thresh_value, 1, cv2.THRESH_BINARY)
thresh_map_uint8 = (thresh_map * 255).astype(np.uint8)

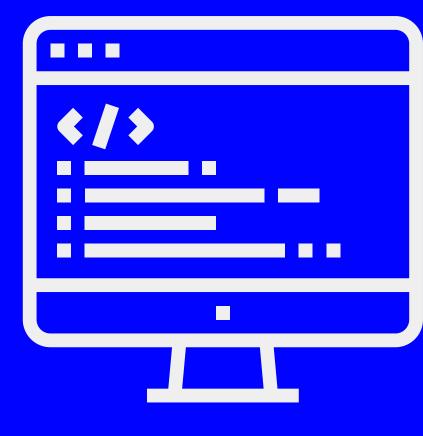
# 5. FIND AND FILTER CONTOURS (REMOVE TINY NOISE)
contours, _ = cv2.findContours(thresh_map_uint8, cv2.RETR_EXTERNAL, cv2.CHAIN_APPROX_SIMPLE)
min_area = img.shape[0] * img.shape[1] * 0.001 # Minimum 0.1% of image area
filtered_contours = [cnt for cnt in contours if cv2.contourArea(cnt) > min_area]

# 6. EXTRACT BOUNDING BOXES AND INITIAL SCORES FOR HOTSPOTS
hotspot_regions_bbox = []
hotspot_scores = {}
dom_importance = {} # Store DOM element importance
```



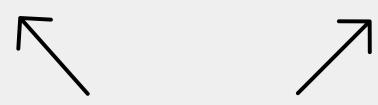
Code Snippet

```
function extractEnhancedDOMData() {
   const importantElements = [];
   const elementsToConsider = document.querySelectorAll('input, button, a, h1, h2, h3, h4, h5, h6, p,
nav, ul, ol, li, form, textarea, select, img, video, div, span, label, header, footer');
   elementsToConsider.forEach(element => {
       const rect = element.getBoundingClientRect();
       if (rect.width > 0 && rect.height > 0 && isElementVisible(element)) {
           const attributes = {};
           for (const attr of element.attributes) {
               attributes[attr.name] = attr.value;
           const computedStyle = window.getComputedStyle(element);
           importantElements.push({
               tag_name: element.tagName.toLowerCase(),
               bounding_box: [rect.left, rect.top, rect.right, rect.bottom],
               text_content: element.textContent.trim().substring(0, 200),
               attributes: attributes,
               style: {
                   backgroundColor: computedStyle.backgroundColor,
                   color: computedStyle.color,
                   fontSize: computedStyle.fontSize,
                   fontWeight: computedStyle.fontWeight.
```





Machine learning model to incorporate user behavior data



Mobile website/app support

Comparative analysis between multiple webpage versions

Future Development

Accessibility focus mode for inclusive design

Integration with design tools (Figma, Adobe XD)

Exploring creativity





API for integration with analytics platforms

Conclusion

- Visual Attention Analyzer bridges the gap between design and user perception
- Combines cutting-edge computer vision with website structure analysis
- Provides actionable insights for design optimization
- Open-source approach encourages community improvement

Chandra Prakash

Thank You

Chandra Prakash