

Saliency Estimation for Advanced Imaging Scenes Using Pixel Statistics

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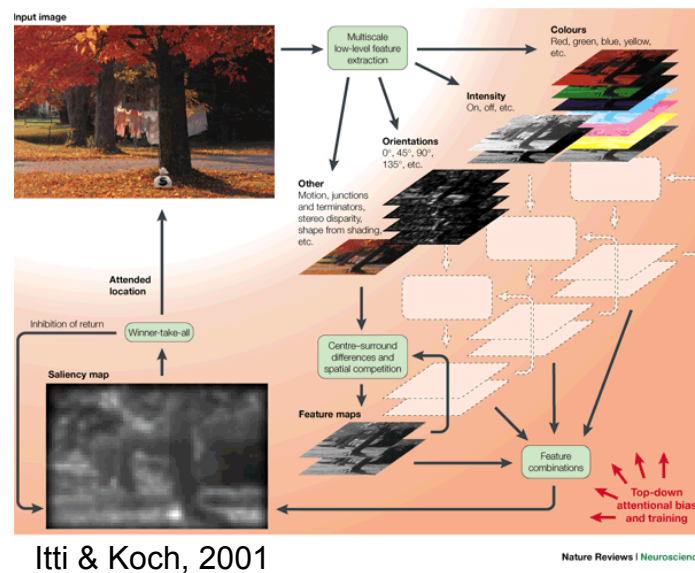
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Background

- Large body of work on estimating visual saliency of natural scene imagery
- “Standard” models readily available for downloading

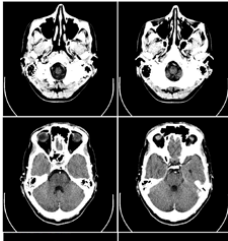


- Some efforts to continue making improvements
 - “Large-scale Scene Understanding Challenge” -
<http://lsun.cs.princeton.edu/>

Advanced Imaging Sensors

- But many of today's advanced sensors produce image products with novel visual characteristics

X-ray



- Shadowing
- Orientation

IR & Thermal



- Saturation
- Resolution
- False color

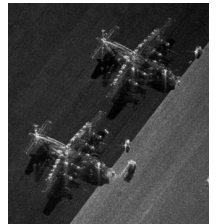
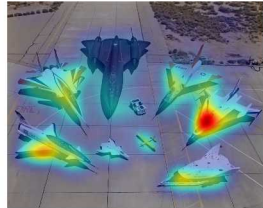
Radar



- Layover
- Shadowing
- Noise

Challenge/Problem

- Data from advanced sensor systems are ultimately interpreted by human analysts - traditional saliency models will have some applicability
 - Information still encoded and displayed using standard visualization parameters such as contrast and color
- Developing technologies will continue to provide challenging imagery
 - *“While dual-energy imaging is now a reality in medical practice, **multienergy** is still in its early stage, but a promising research activity.”¹*



¹Pacella, D., Reports in Medical Imaging, Vol. 8, 2015

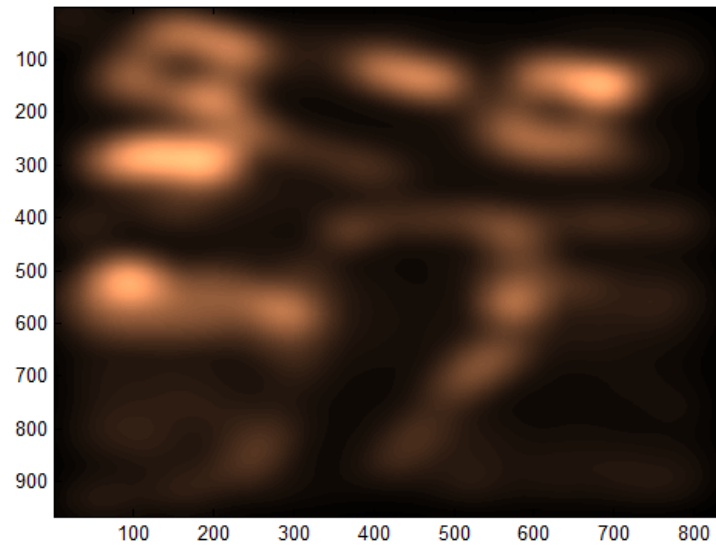
Study Overview

- How well does existing model (Itti & Koch) predict saliency in synthetic aperture radar (SAR) imagery?
- How can standard saliency estimation be improved to better predict gaze patterns of sensor-knowledgeable viewers?

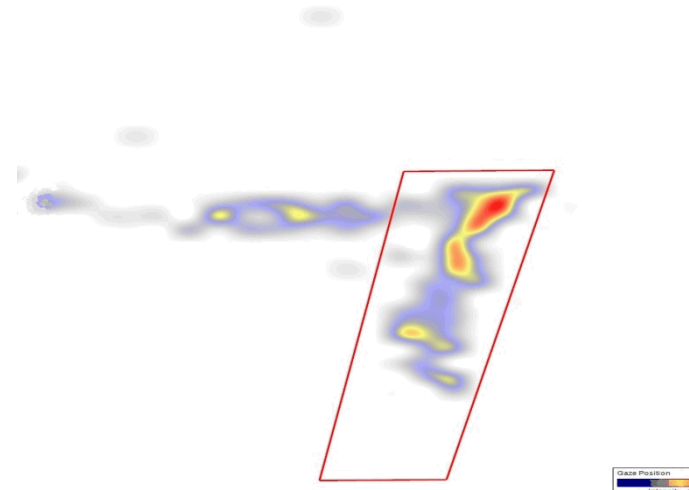
- Study task - change detection in SAR imagery
- Participants
 - 3 with no SAR experience (“novices”)
 - 6 radar engineers familiar with SAR (“engineers”)
 - 3 professional SAR imagery analysts (“experienced IAs”)

SAR Example – Saliency vs. Actual Gaze

Saliency Map



Gaze Map



Saliency Comparison Metrics*

1. Linear Correlation Coefficient (CC)

- Measure of the strength of a linear relationship between fixation map (G) and saliency map (S)

- $CC(G, S) = \frac{cov(G, S)}{\sigma_G \sigma_S}$ When CC is close to ± 1 , there is almost a perfectly linear relationship

2. Normalized Scanpath Saliency (NSS)

- Average of saliency values at human gaze positions (saliency normalized to have zero mean and unit standard deviation)
 - NSS = 1 indicates that the subjects' gaze positions fall in a region whose predicted saliency is one standard deviation above average
 - When $NSS \geq 1$, the saliency map exhibits significantly higher saliency values at human gaze locations compared to other locations
 - $NSS \leq 0$ indicates the saliency model performs no better than picking a random position

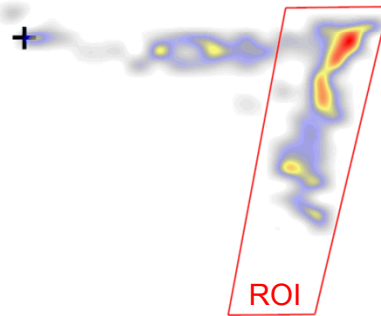
3. Area Under Curve (AUC)

- Human gaze positions are considered positive set, other points are negative set
- Saliency map is treated as binary classifier to separate positive and negative sets

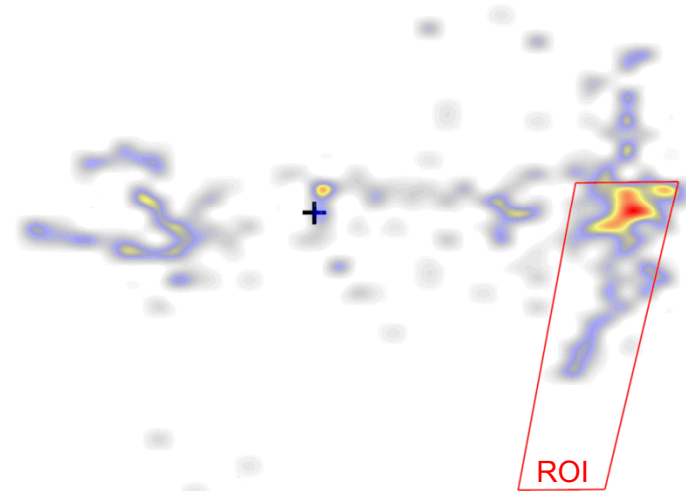
*Borji, A., et al. (2013). "Quantitative Analysis of Human-Model Agreement in Visual Saliency Modeling: A Comparative Study." IEEE Transactions on Image Processing **22(1)**: 55-69.

Example Gaze Maps By Expertise

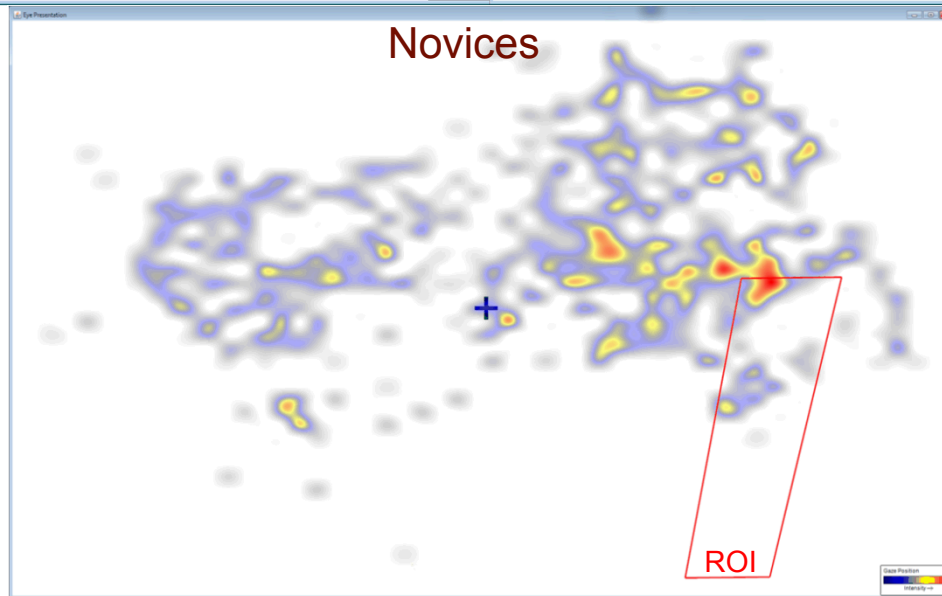
Image Analysts



SAR Engineers - Same Domain

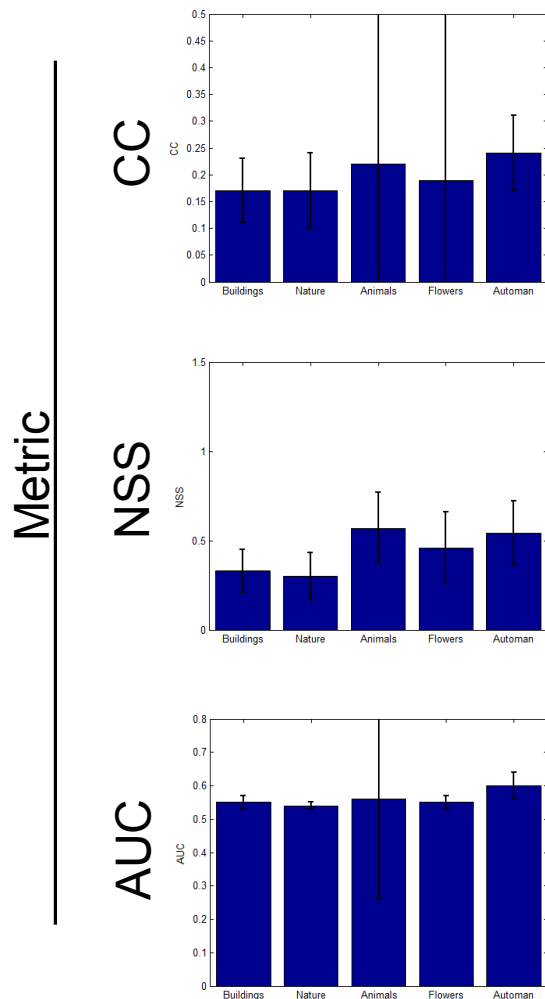


Novices



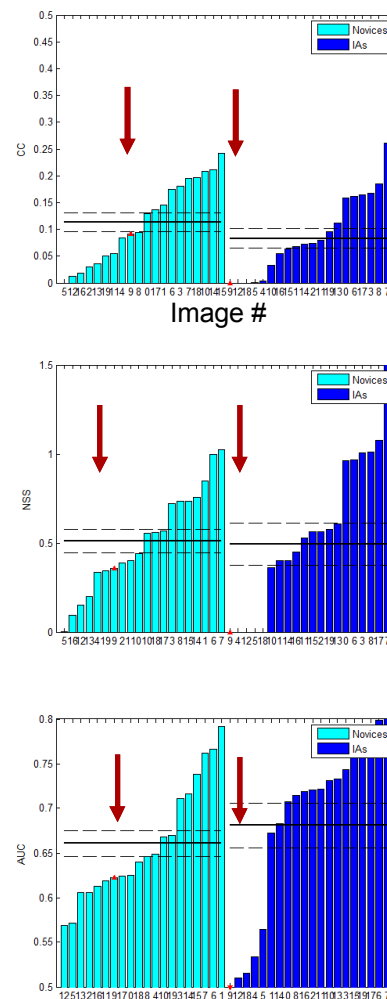
Standard Saliency Estimation Applied to SAR Imagery

Image Type*
Kootstra & Shomacker Image Set



Buildings - Nature - Animals - Flowers - Automan

SAR Images

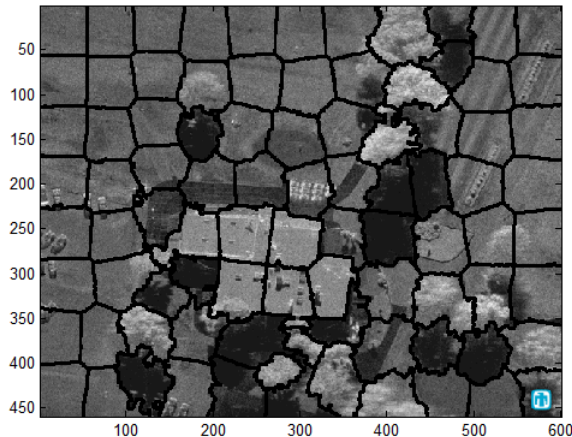


Itti COI2 Used for all
SAR Images

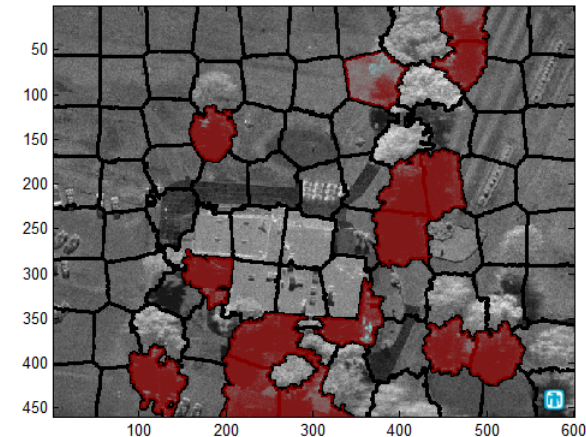
Reducing Saliency Estimates in Shadow Regions

- Pixel-statistical methods used to segment¹ the scene and characterize the segment properties²
- These properties can serve as filters to modulate traditional saliency estimates
 - SAR Phenomenology - shadow regions have low coherence

Segment



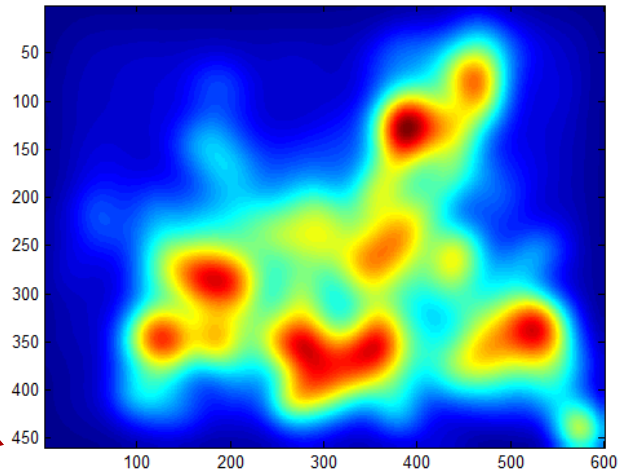
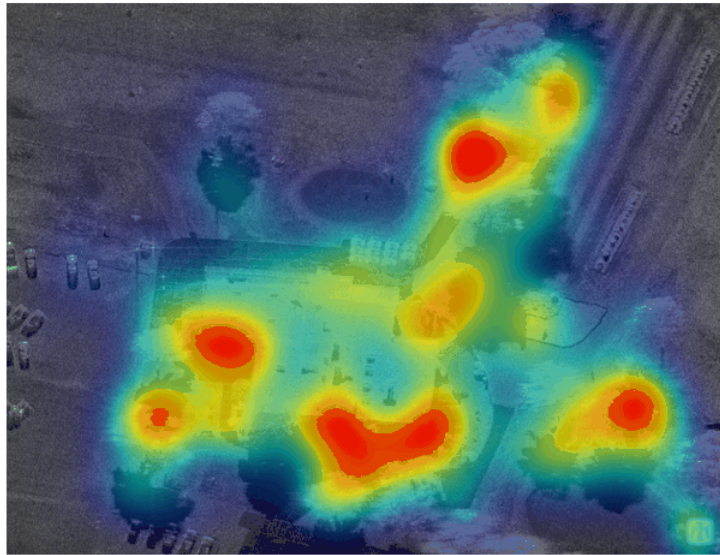
Classify



¹ M. M. Moya, et al, "Superpixel segmentation using multiple SAR image products" RADAR SENSOR TECHNOLOGY XVIII, Proceedings of SPIE VOL 9077, Conference on Radar Sensor Technology XVIII, MAY 05-07, 2014, Baltimore, MD

² M.M. Moya, et al., "Superpixel Classification for Signature Search in Synthetic Aperture Radar Imagery," Conference on Data Analysis (CoDA), March, 2014, Santa Fe, NM.

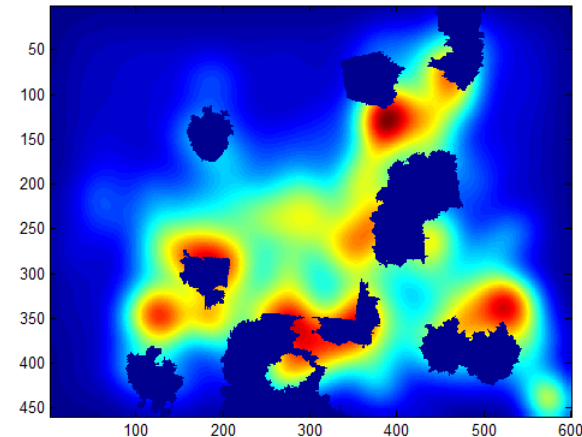
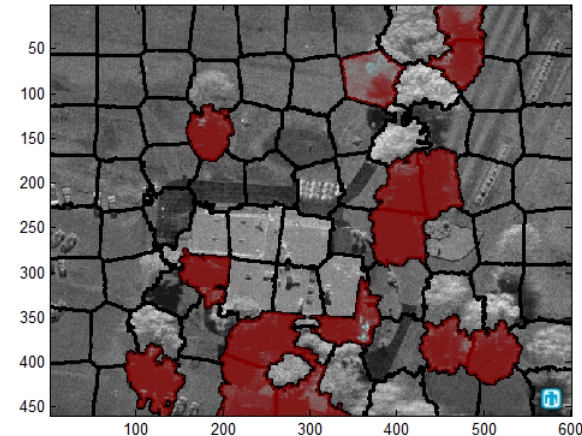
Method (1): Natural Scene Saliency Map



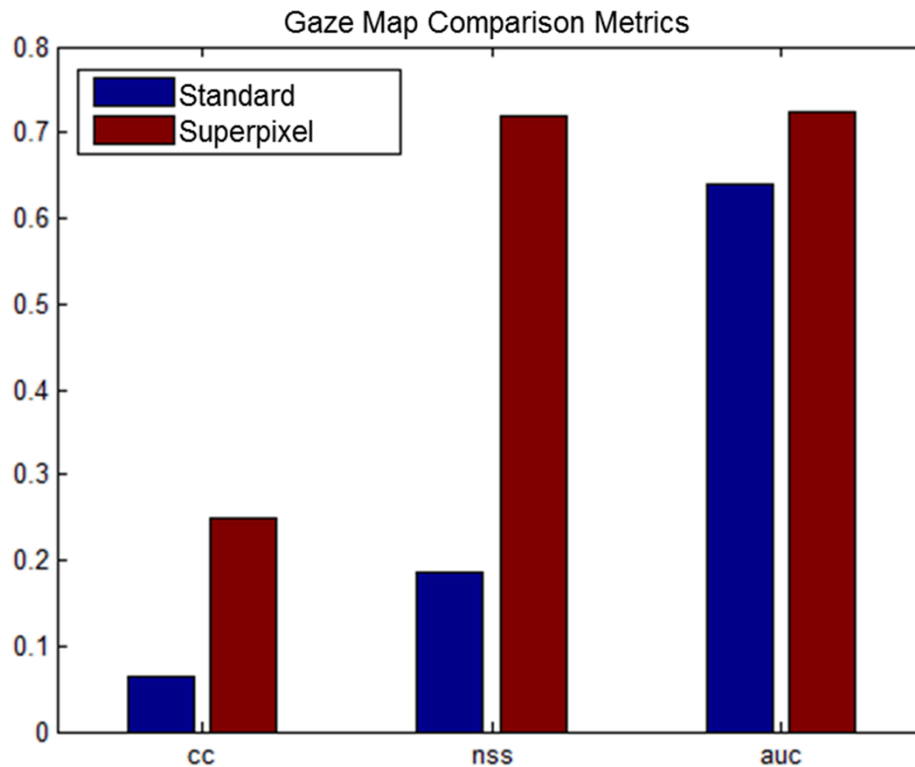
¹ Itti citation

Method (2): Select and Filter Based on Superpixel Characteristics

- Select superpixels with certain characteristics (i.e. shadows)
 - Classify using pixel statistics within each superpixel
- Apply mask to original saliency map
 - Can add Gaussian, or other smoothing to reduce discontinuities



Study Results



Saliency map modulated by superpixel characteristics is more similar to analyst fixation maps

- Linear correlation (cc) improvement factor is 3.8X
- Normalized scan path saliency (nss) improvement factor is 3.9X
- Area under receiver-operator curve (auc) improvement factor is 1.1X

Conclusion

- Saliency estimates from “standard” model have high variance in agreement with gaze maps
- Modulating standard model using superpixel segmentation and classification based on sensor phenomenology can improve salience – gaze agreement
- Using eye tracking technology to explore relationships between traditional saliency models and pixel-statistical properties we can understand eye movements of domain experts interacting with imagery from today’s most advanced sensors