Perceptually Based Tone Mapping for Low-Light Conditions

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Purkinje Effect



Bright scene



Low-light => bluish

Key idea

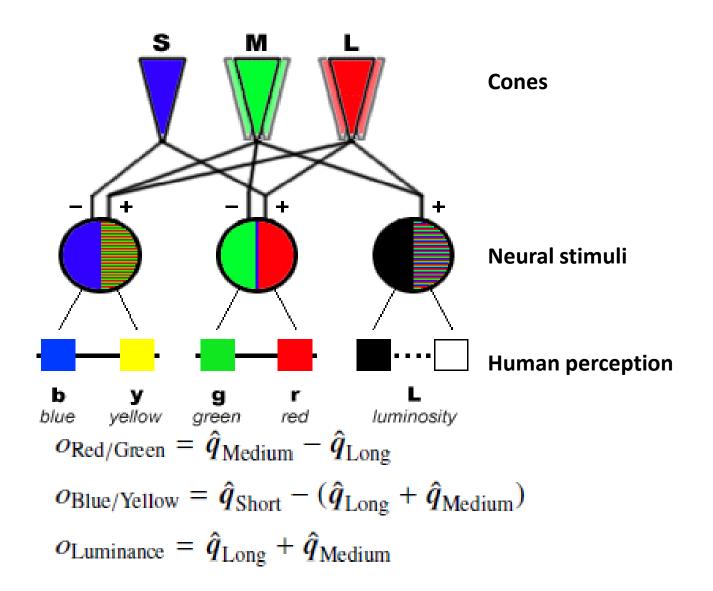


Match human vision in low-light conditions

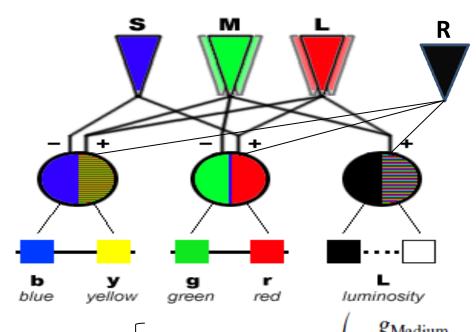
Human vision

- Bright scene: dominated by cones (photonic vision)
- Dark scene: dominated by rod (scotopic vision)
- In between? Low-light vision (mesopic vision)
 - rod and cones all contributes (4D-> 3D)
 - This is why we perceived color change in low-light
 - Cameras do not have this mechanism!

Opponent color model in photopic vison



Low light vision



Cones and Rod

Neural stimuli

Human perception

$$\Delta o_{\text{Red/Green}} = x \, \kappa_1 \left(\rho_1 \frac{g_{\text{Medium}}}{m_{\text{max}}} - \rho_2 \frac{g_{\text{Long}}}{l_{\text{max}}} \right) q_{\text{Rod}}$$

Neural stimuli diffs

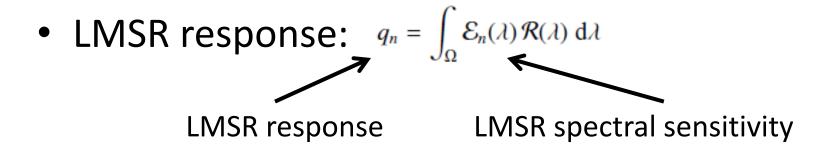
$$\Delta o_{\text{Blue/Yellow}} = y \left(\rho_3 \frac{g_{\text{Short}}}{s_{\text{max}}} - \rho_4 \left(\alpha \frac{g_{\text{Long}}}{l_{\text{max}}} + (1 - \alpha) \frac{g_{\text{Medium}}}{m_{\text{max}}} \right) \right) q_{\text{Rod}}$$

$$\Delta o_{\text{Luminance}} = z \left(\alpha \frac{g_{\text{Long}}}{l_{\text{max}}} + (1 - \alpha) \frac{g_{\text{Medium}}}{m_{\text{max}}} \right) q_{\text{Rod}}$$

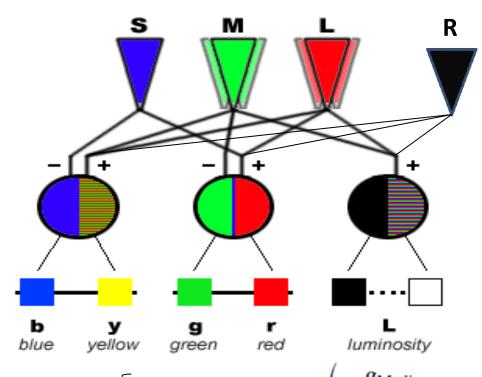
Method Overview

- The only work apply human vision model to deal with Purkinje effect in images
- Flow
 - Acquiring Spectral Images to estimate LMSR response
 - Compute nerual stimuli difference due to rod
 - Convert neural stimuli back to RGB

Acquiring Spectral Images



- Each pixel is represented by a spectral function R(λ)
- Take snapshot of the same scene with 10 band pass filters
- Cubic B-spline curve fitting to get R(λ)



Cones and Rod

Neural stimuli

Human perception

$$\Delta o_{\rm Red/Green} = x \, \kappa_1 \left(\rho_1 \frac{g_{\rm Medium}}{m_{\rm max}} - \rho_2 \frac{g_{\rm Long}}{l_{\rm max}} \right) q_{\rm Rod}$$
 Neural stimuli diffs
$$\Delta o_{\rm Blue/Yellow} = y \left(\rho_3 \frac{g_{\rm Short}}{s_{\rm max}} - \rho_4 \left(\alpha \frac{g_{\rm Long}}{l_{\rm max}} + (1-\alpha) \frac{g_{\rm Medium}}{m_{\rm max}} \right) \right) q_{\rm Rod}$$

$$\Delta o_{\rm Luminance} = z \left(\alpha \frac{g_{\rm Long}}{l_{\rm max}} + (1-\alpha) \frac{g_{\rm Medium}}{m_{\rm max}} \right) q_{\rm Rod}$$

(12)

Convert back to RGB image

- Compute RGB values best represent the neural stimulus
- Apply standard HDR range compression first
- Looks weird (a low light image with full range)
- Suppress pixel value by mesopic factor w
 - w=0 -> fully photonic, w>1 -> fully scotopic

$$\dot{P} = P * \max(1 - w(1 - \gamma), \gamma), \qquad \gamma \in [0.25, 0.5]$$

Result (Range Compression)

Exposure _____





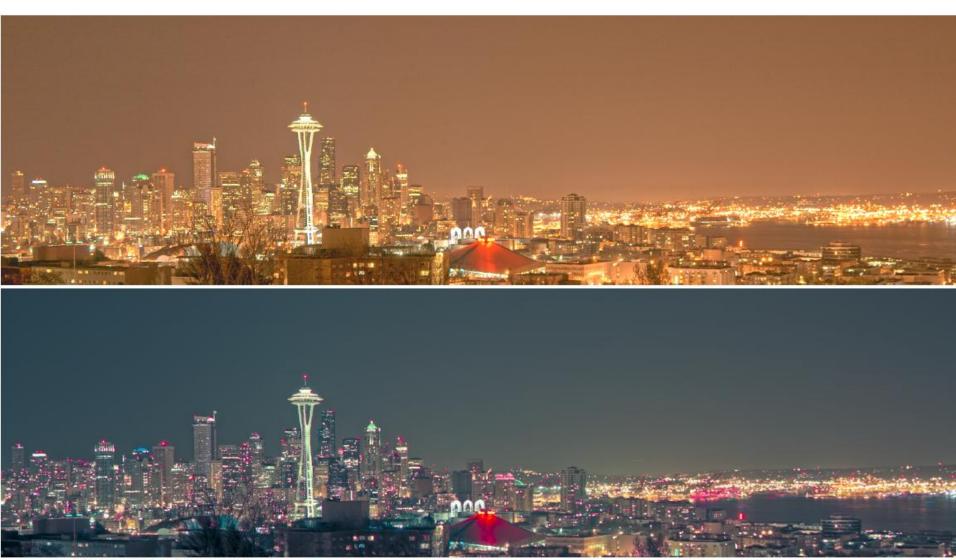








Results



Images copyright Adam Kirk and James O'Brien.

Results





Images copyright Adam Kirk and James O'Brien.

Results





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Approximation of Non-Spectral Images

- Spectral images are hard to acquired
- Build a RGB->LMSR mapping matrix by data trained from spectral images
- Apply that mapping to RGB images to create approximated LMSR value

Results (Non-Spectral Image)



Original image "Foggy Night" copyright Jack Tumblin, Northwestern University.

Results (Non-Spectral Image)





Thanks



Differences/Alternatives

-What happens when using RGB input images instead of multiple spectual images?



Spectral image tone-mapped



Non-spectral HDR image tone-mapped

– How do the two methods compare?

— Does the approximate work well?
Why or why not?

– Do the results fit your perception?

- Other techniques?

- Limitations?
 - Focus only on interplay between rods and cones in early stage of vision
 - Other adaptive mechanisms not incorporated
 - Additional effects not modeled

- Future works
 - Apply method to all sort of light intensities
 - Predict actual viewer experiences
 - Apply to videos