

(yet more) Inverting Bioluminescent Radiance

(and)

A Very Brief Look at the Spectral
Dependence of Forward Modeled

$$T^*f/Q=R_{rs}/[b_b/(b_b+a)]$$

Two unrelated projects by

Wayne Slade

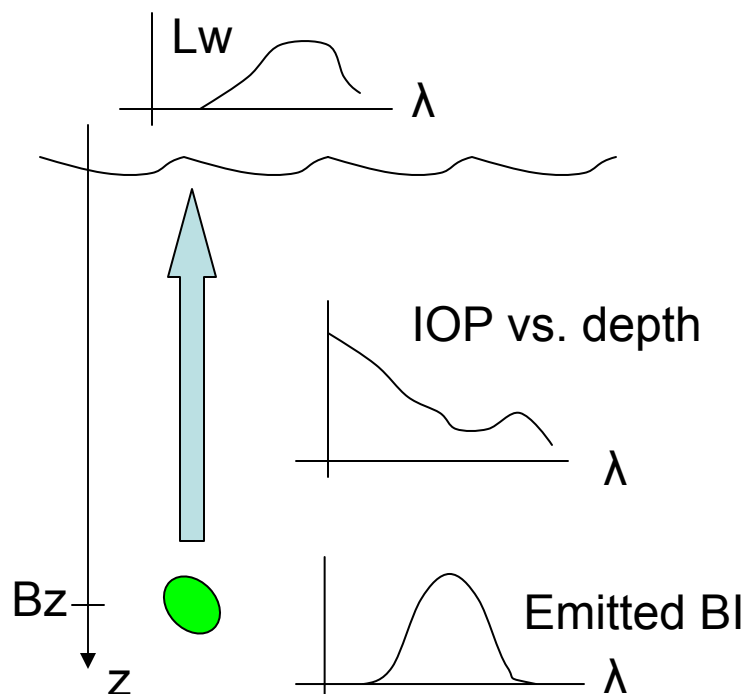
DMC Ocean Optics 2004

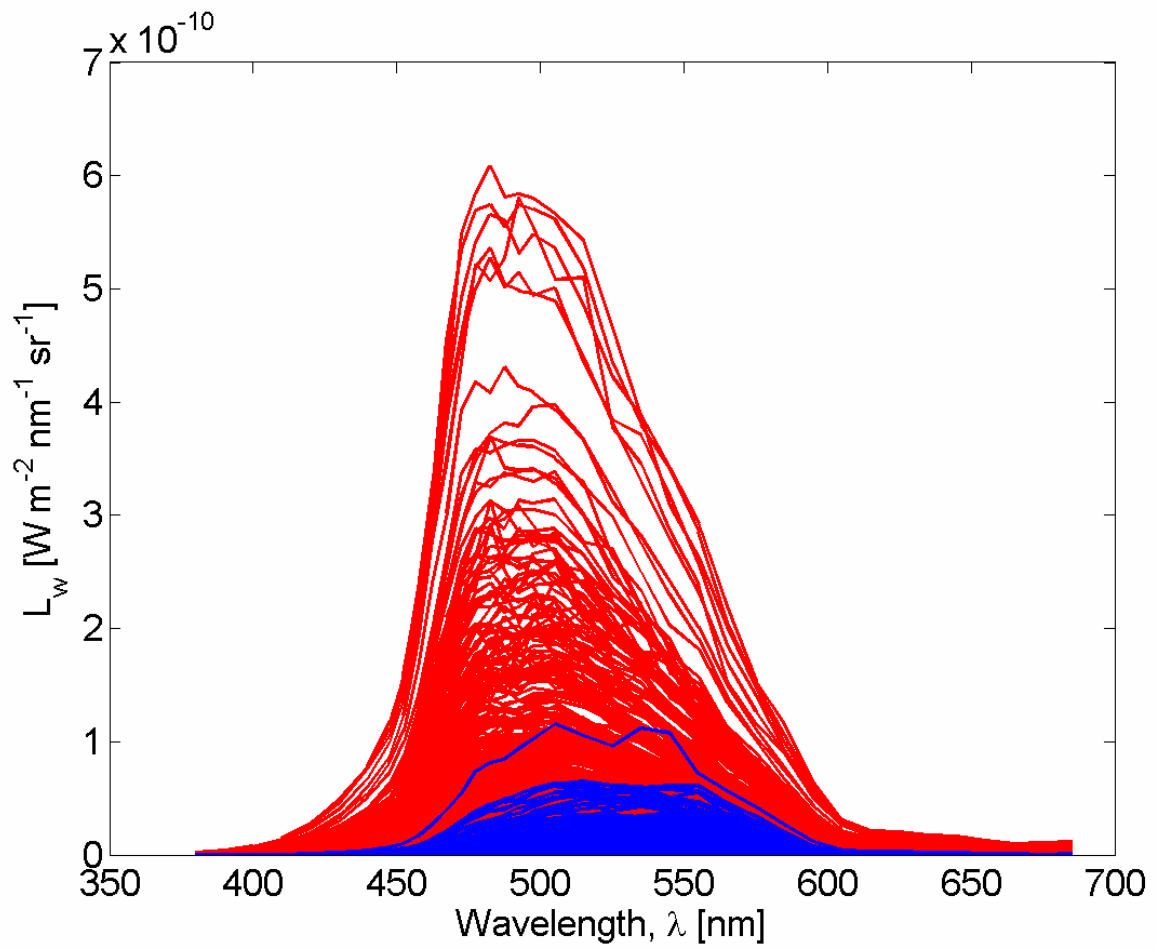
Part 1 – Bioluminescent Radiance Inversion

- Movement of objects in water stimulate bioluminescence within the water column
- Depth of stimulation dictates how far the light travels before reaching the surface
- Radiance that reaches the surface depends strongly on its absorption and scatter by water column IOP, as well as spectral shape and intensity of bioluminescence (i.e. how many phyto)
- Inverse Problem – can we use the bioluminescent radiance seen at the surface to determine the depth where the bioluminescence was stimulated?

Bioluminescent Radiance Inversion

- Dataset consists of ~7400 measurements of IOP vs. depth with bioluminescence measured (stimulated) at different depths from bathyphotometer and ac-9 [from the LEO-15 crew, see Matt Oliver]
- HYDROLIGHT was used to derive L_w from bioluminescent layer, propagated through water column IOPS and sea surface (no sun)





Bradt (2004) derived a special (analytic) form of Conservation of Misery (a la Mobley)

$$\frac{N}{K} = F_B$$

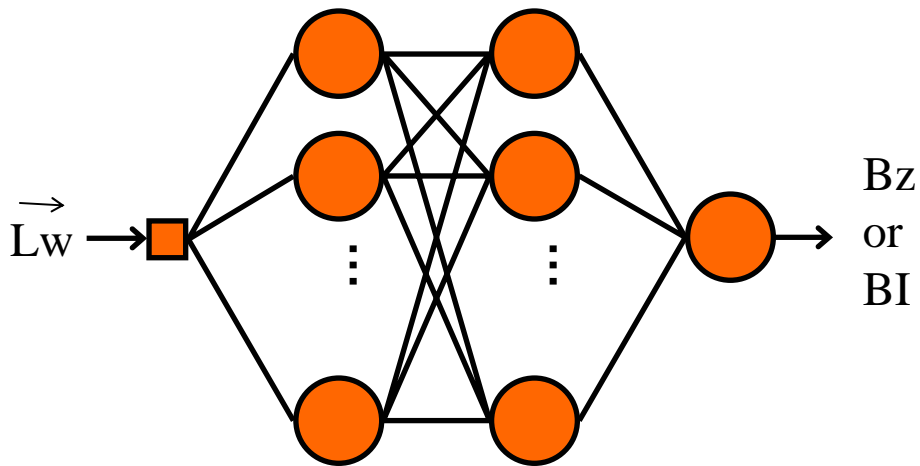
Furthermore, Bradt empirically related brute force to the elegance parameter

$$\frac{N}{K} = F_B = \eta \frac{1}{E}$$

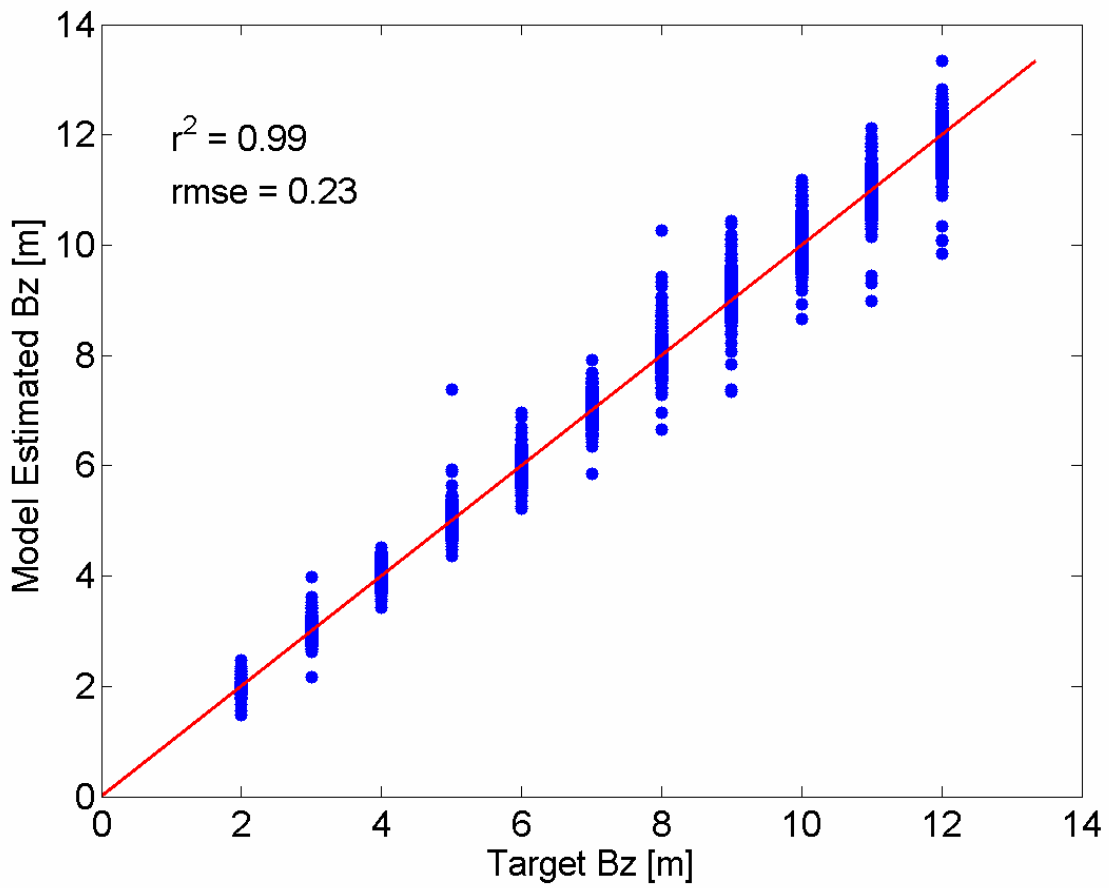
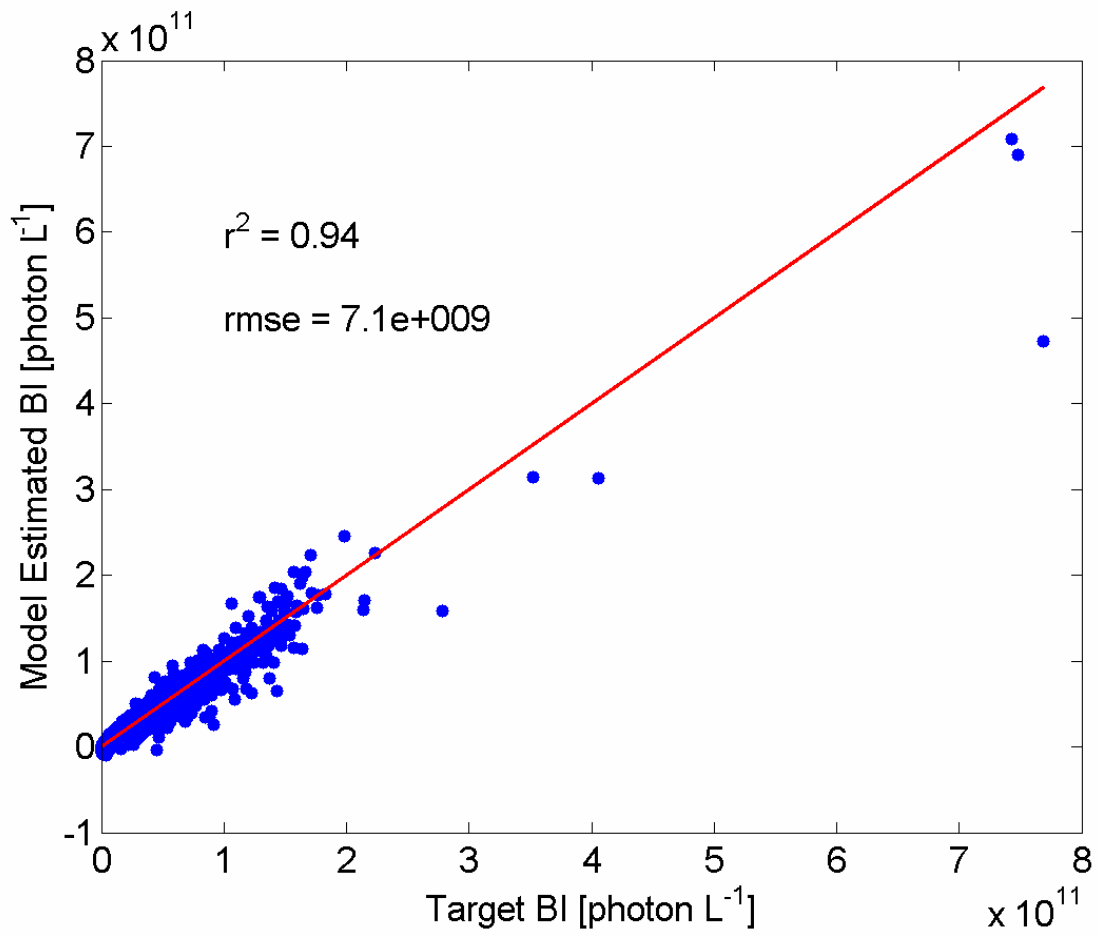
Application of Bradt's relationship to bioluminescent radiance

$$\begin{array}{ll} \lim_{k \rightarrow 0} \frac{N}{K} & F_B \rightarrow \infty \\ & E \rightarrow 0 \end{array}$$

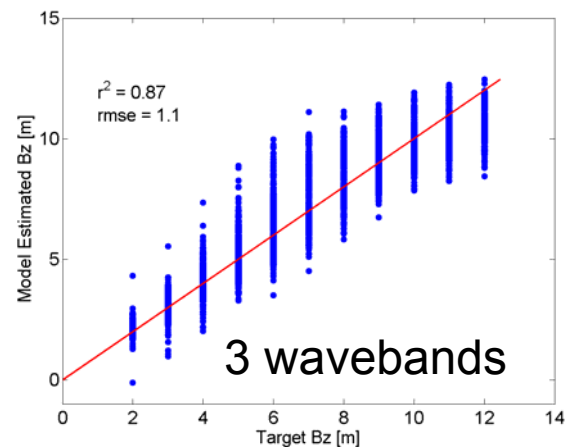
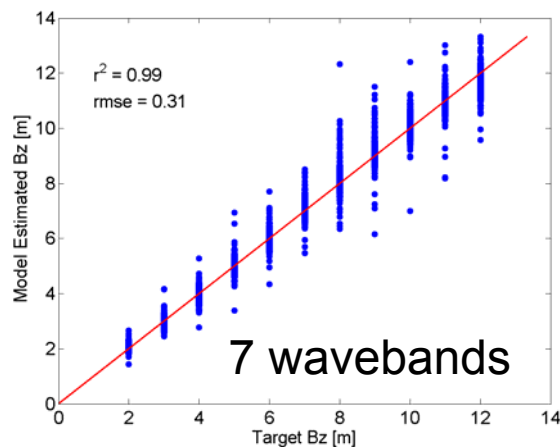
NN Inversion Model



- Development of simple empirical model is tough since IOP, BI, and sea state are variable
- Two separate NN models were trained to estimate Bz and BI from spectral Lw
 - Network input Lw (~37 wavebands from 380-700)
 - Network output Bz or BI (depth or intensity)
- Generalization ability of network is very important!
 - Avoid too few or too many nodes (so that noise is not learned)
 - Stop network training early by monitoring performance on validation dataset
- NN models are tested on an independent dataset (60% used for training, 20% for validation, 20% testing)



More NN Inversion



- NN model was able to extract information from hyperspectral data
- Trying wavelength subsets (every 5 wavebands and [400 480 550]) yielded degraded but reasonable results
- Empirical and analytic approaches seem particularly tough for this problem
- Analytic approach is preferable, but if the best you can do is empirical (i.e. band ratio), then consider a NN black box!

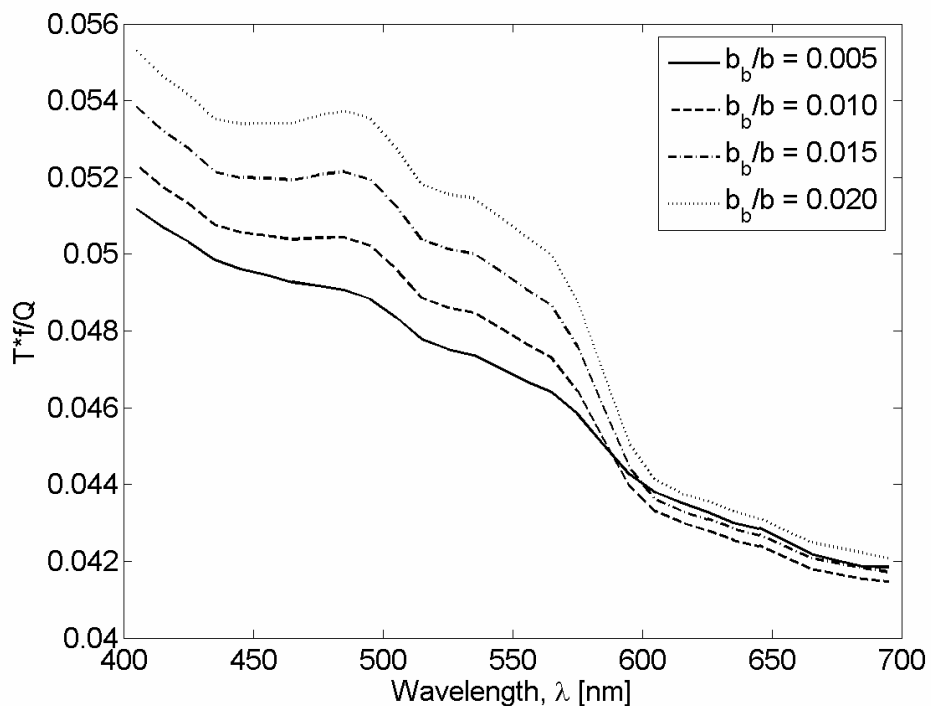
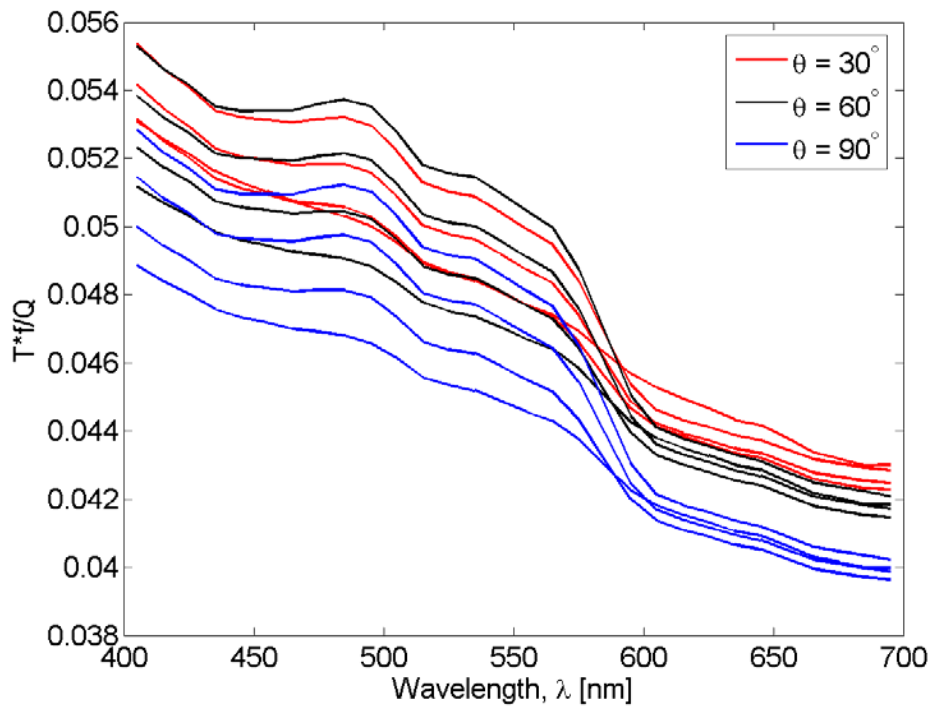
Part 2 – Forward Modeled Spectral Dependence of T^*f/Q

- Irradiance Reflectance is defined as $R=Eu/Ed$, and has been approximated (such as by Gordon et al.) as $R=f*[bb/(bb+a)]$, with $f\sim 0.33$ at $z=0$ -
- f/Q depends on sun and satellite geometry as well as IOP
- $Rrs=Lu/Ed$, so define $Q=Lu/Eu$ and include $T\sim 0.54$ to propagate across sea surface

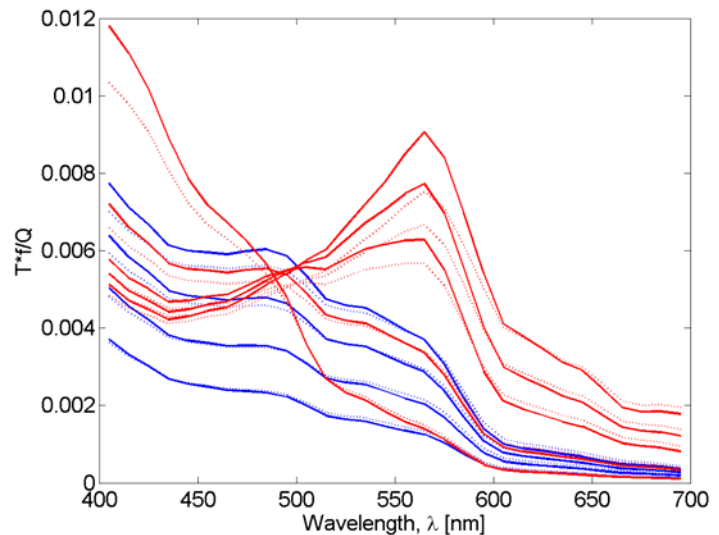
$$Rrs = T \frac{f}{Q} \left[\frac{bb}{bb + a} \right]$$

- T^*f/Q can be an important parameter in Rrs inversion (typically assumed constant 0.051)
- Spectral dependence could influence spectral matching based algorithms

Forward Modeled Spectral Dependence of T^*f/Q (HYDROLIGHT)



Finale



- Forward modeling shows T^*f/Q has spectral variation, but assumption of spectrally flat $T^*f/Q=0.051$ leads to 5-10% error in R_{rs}
- More accurate T^*f/Q will probably become more important for inversion algorithms on calibrated hyperspectral data where we aim to retrieve more subtle features
- 20 forward HYDROLIGHT runs only taps the surface of this topic; next steps are to understand works of Morel and Gentili and others, serious HYDROLIGHT runs, and a good look at variation in the real world (i.e. Satlantic HyperPro can easily measure Q , combined with more sensors for bb,b)

**Diffuse reflectance of oceanic waters.
III. Implication of bidirectionality for the
remote-sensing problem**