

## Optimizing Image Analysis Processing in Thin Transparent Aquifers

## Application to Pixel Wise Regression of Salt-Water Intrusion

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## Outline

- Experiment & Motivations
- Pixel-wise regression procedure:
  - Power-law fitting (grayscale, standard)
  - Objectives
- Novel alternatives:
  - Laurent series fitting (grayscale)
  - Bead-correcting Beer-Lambert fitting (monochromatic)
  - Reduction-deviation metric fitting (color image)
- Comparison & Summary









Lab device & camera setup



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## **Background:** Salt Water Intrusion (SWI)



### Salt water intrusion in a coastal aquifer





- SWI increases via anthropogenic environmental changes:
  - Pumping
  - Evapotranspiration (e.g. higher temperatures and crops)
  - Urban runoff
  - Sea level rise
- Physical complexity of systems
  - Pressure-density-diffusion transport in porous media
  - Ordered heterogeneity: Stratification and fractures
  - Random heterogeneity
  - Tidal/seasonal variability
  - Pumping
- Important and complex system that needs rigorous experimental study

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## **Experimental System**

- Transparent aquifer (1 cm thick) is filled with glass beads of approximately 1 mm diameter
- Head-driven fresh water flows from left
- Salt and red dye (Allura Red) pre-mixed with water in separate tank
- Salt water recirculates by density-driven flow from right reservoir
- White LED backlights illuminate the transparent aquifer
- Cameras (grayscale and color) record the state of the aquifer
- Transparent aquifer setup gives precise information on length of SW intrusion and mixing zone of interface







Smaller SWI sandbox apparatus with original (left) and new (right) cameras

## **Pixel-wise Image Post-Processing Experimental Procedure**





### Prior to intrusion test: 8 calibration images taken of aquifer filled with fractional amounts of saltwater

- Aquifer flushed to pure water, then SW fills reservoir to intrude into aquifer
  - Images taken every 5 minutes to steady state interface
- Experiments prior to this work were carried out with grayscale camera only.
- To take calibrations images, 1 intrusion test requires an extended workday.

### **Post-processing Procedure**

- Crop image set automatically
- At each pixel, calibrate the fitting function via the dilution images.
- Apply fitting sets to intrusion images









Map

## **Power Law Fitting**



• Power Law Fitting (PLF) on grayscale images:  $C = m I^{x_2} m$ 

$$C = x_1 I^{x_2} - x_3$$

- Applied at each pixel to determine coefficient values, x<sub>i</sub>, for light curve
- Takes > 1 day to calibrate full RGB image!
- This specific power-law equation cannot be linearized
  - Non-linear optimization is **computationally expensive** relative to linear optimization
- For average data:  $x_2 = -2.86$
- Less accurate simplification of fitting:

$$C = x_1 + x_2 I^{-3}$$







**Power Laws fitting example calibration data** GR power law (black) & simplified law (dashed)

## **Objectives**



New larger-scale experimental apparatus (10x larger aquifer volume)

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### **Goals for Optimization**

- Speed up, simplify, and improve accuracy of experimental process at all stages: Setup & calibration, intrusion runs, and data processing
- Simplicity: Focus on *pixel-wise* methods using simple, fast fitting functions
- Speed up: Improve post-processing times by optimizing numerical methods
- Efficiency: Reduce number of calibrations to speed up time of experiment, and decrease water and salt resource use

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## **Laurent Series Fitting**

- Expanding linear regression equation into Laurent Series (LSF)  $C = x_1 + x_2I^{-1} + x_3I^{-2} + x_4I^{-3}$
- Significant benefits to LSF
  - Simpler computational method via linear regression solution
  - Radical speedup: 3 minutes (>10<sup>3</sup> x faster than power law!)
  - Improved accuracy near extremes
- Some concerns with LSF method
  - Not a physically motivated method
  - Subject to errors near lowconcentration limit (non-monotonic)
  - Does not simplify experimental procedure (similar # of calibrations)







### Laurent series fitting example calibration data Laurent series (red) visually gives best fit among methods

# Color Absorption Behavior Prediction with Beer–Lambert Law





### Backlights, Red dye, and BW-camera spectra

Sensitivity of camera is poorly correlated with dye absorption







Spectral prediction for BW-camera after white LED light passes through red dye



### **C-I Curve for spectral prediction** Mismatch indicates additional effects in spectrum

## **Bead-Correcting Beer–Lambert Fitting**



Light transmission through cross-section

- Lowest-order light decay is 1/C going through axis, other terms are higher order with e<sup>-C</sup> like in the Beer–Lambert Law
- Solid-solid contact points hypothesized to dominate light transmission in highconcentrations
- Suggests adding inverse intensity to BLL

$$C = x_1 \log\left(\frac{1}{I}\right) + x_2 + x_3 \frac{1}{I}$$

- Asymptotic expansion gives a Laurent series.
- Ideally applied to filtered monochrome pictures







— True Int (√(1-r<sup>2</sup>)) — Axial (r<sup>2</sup>) — BLL — BCBLL × Corr Pts C-I Curve: Comparison of numerical integrals of theoretical monochromatic data with BLL & BCBLL

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## Real Color Absorption-Emission Behavior

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Combination of reduction and deviation

(RDM) is nearly linear

100%

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Deviation between color

## **Reduction–Deviation Metric Fitting**





Concentration regression on metric (C-M curve)







- Low-order functions of the Reduction-Deviation metric reproduce accurate concentrations
- Near-linear influence of dye on metric-based light curve; requires less calibration
- Calculation of RDM field over image and pixelwise fitting coefficients are fast
- Fitting functions are stable and monotonic
- Requires RGB camera
- Significant improvements over grayscale methods

## **Performance of Methods**









# Fitting coefficient computation time vs. downscaling

- Power law fitting is three orders of magnitude slower than the other fitting methods
- Linear RDM is extremely fast





others

### **Calibration error vs. downscaling**

- Grayscale methods have very similar error values
- The quadratic RDM is much more accurate than the other fittings
- Linear RDM is nearly as good as

### RMSE vs. number of calibration images (0.01 ds)

- For most, fewer fittings have only small impact on accuracy
- At least one more image than number of fitting function coefficients is best

## **Summary of Methods**

BY NC ND



Method	Post- processing time	Minimum Calibration Images	Accuracy	Fitting Function Stability	Sensor Requirements
Power Law (Standard) $C = x_1 I^{x_2} - x_3$	>1 day (nonlinear!)	5	Acceptable (5% error)	Stable, Monotonic	N/A (grayscale image)
Laurent Series $C = x_1 + x_2 I^{-1} + x_3 I^{-2} + x_4 I^{-3}$	3 minutes	6	Acceptable (5% error)	Non-monotonic	N/A (grayscale image)
Bead-Correcting BLL $C = x_1 \log\left(\frac{1}{I}\right) + x_2 + x_3 \frac{1}{I}$	2 minutes	5	Acceptable (5% error)	Non-monotonic	(Monochrome light filter?)
1 <sup>st</sup> order Reduction– Deviation Metric $C = x_1 M$	0.01 second	2	Acceptable (6% error)	Linear, monotonic	RGB camera
2 <sup>nd</sup> order Reduction– Deviation Metric $C = x_1M + x_2M^2$	1 minute	4	Best (1.5% error)	Monotonic	RGB camera
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## Conclusions



- Linear regression methods are much faster than nonlinear.
  - The Laurent Series Fitting has similar accuracy to the Power-Law Fitting, but is orders of magnitude more rapid to postprocess.
- Beads may significantly effect the experimental light curve.
  - The simpler, physics-based linear regression adapting the Beer–Lambert Law to this experiment performs as well as the other fitting functions.
- Allura Red dye causes spectral effects in RGB camera via spectral-variable absorbance and red colored fluorescence.
  - RGB specific methods perform better than grayscale intensity methods (e.g. PLF & LSF).
  - The second order Reduction-Deviation Metric method reduces the error by two-thirds.
- All novel methods improve post-processing times and can reduce the calibration cost of the experiment.
  - In particular, the first-order reduction-deviation metric fitting reduces the concentration estimation to two calibrations, while maintaining comparable accuracy.





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