Machine Learning for Sustainable Development and Biological Conservation

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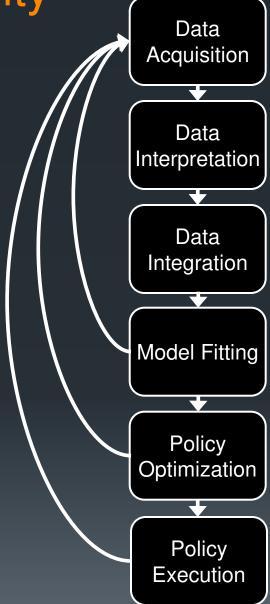
President, Association for the Advancement of Artificial Intelligence





Computational Sustainability

 The study of computational methods that can contribute to the sustainable management of the earth's ecosystems



Data Acquisition

- Africa is very poorly sensed
 - Only a few dozen weather stations reliably report data to WMO (blue points in map)
- Project TAHMO (tahmo.org)
 - TU-DELFT & Oregon State University
 - Deploy 20,000 stations across Africa
 - Provide data to farmers and to enable crop insurance industry
 - Increase agricultural productivity
- Computational Problem
 - Where to place the weather stations?
 - Krause, Singh & Guestrin, 2008



Data Acquisition







Data Interpretation

Insect identification for population counting

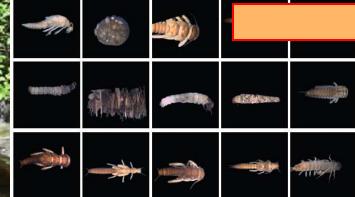
Raw data: image

Interpreted data: Count by

Method: Computer Vision

Lytle, et al., 2010





Data Acquisition

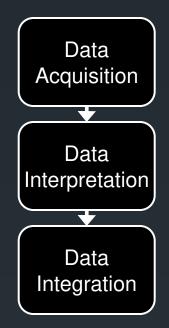
Data nterpretation

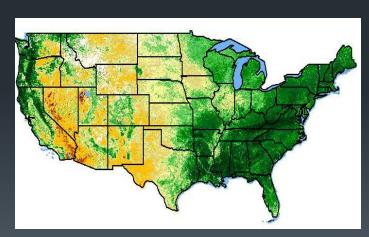
Species	Count
Limne	3
Taenm	15
Asiop	4
Epeor	25
Camel	19
Cla	12
Cerat	21

Al For Social Good

Data Integration

- Virtually all ecosystem prediction problems require integrating heterogeneous data sources
 - Landsat (30m; monthly)
 - land cover type
 - MODIS (500m; daily/weekly)
 - land cover type
 - Census (every 10 years)
 - human population density
 - Interpolated weather data (15 mins)
 - rain, snow, solar radiation, wind speed & direction, humidity



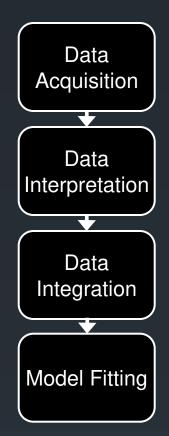


Landsat NDVI:

http://ivm.cr.usgs.gov/viewer/

Model Fitting with Machine Learning

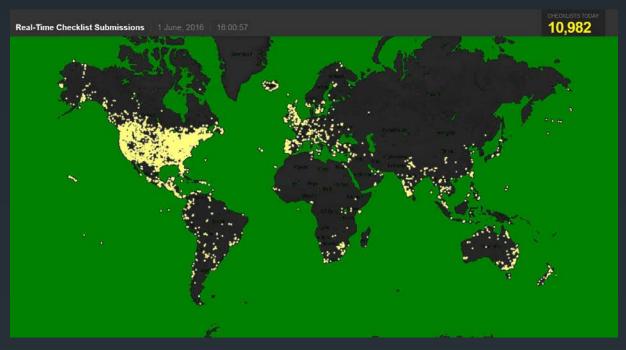
- Species Distribution Models
 - create a map of the distribution of a species
- Migration and Dispersal Models
 - model the trajectory and timing of movement



eBird Project



- Volunteer Bird Watchers
- Time, place, duration
- Species seen
- **8**,000-12,000 checklists uploaded per day



Computational Method: Collective Graphical Model (Sheldon) et al., 2011)







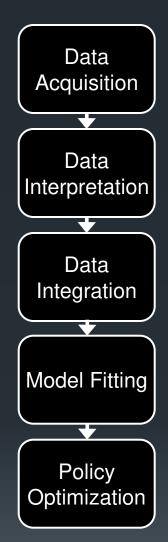
Sheldon, Sun, Liu, Dietterich unpublished

Fitted Migration Model Ruby-Throated Humming Bird



Policy Optimization

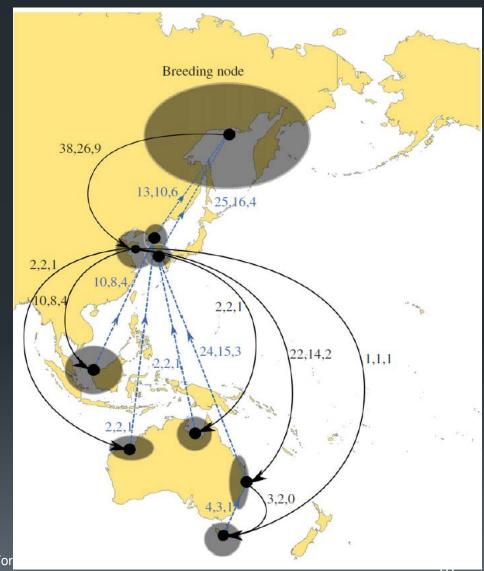
- Compute optimal policies for managing ecosystems
- Incorporate uncertainty about the future
- Computational Tools
 - MDPs (Markov Decision Problems)
 - POMDPs (Partially-Observable MDPs)
 - Point-based solvers (Pineau, 2003; Poupart, et al. 2005; Kurniawati et al, 2008)



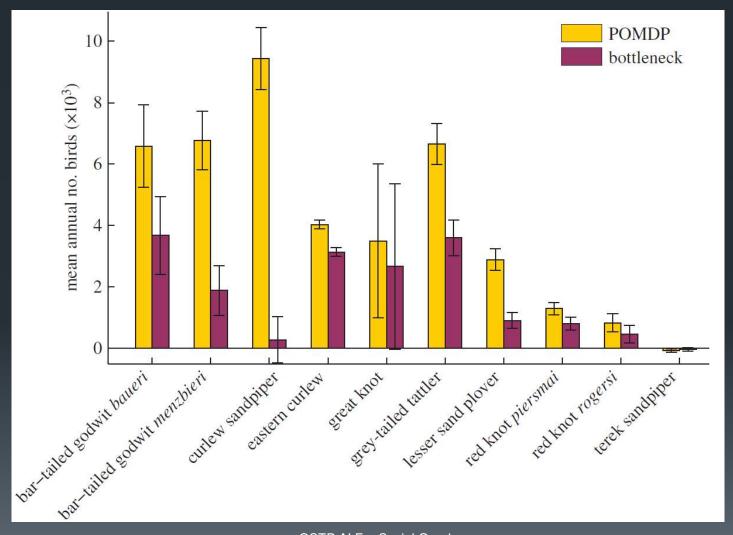
Protecting Coastal Habitat to Protect Migrating Birds from Sea

Level Rise

- East Asia-Australia migratory pathways
- Sea Level Rise destroys habitat unless areas further inland have been protected
- Timing and location of protection depends on the timing of future sea level rises
- POMDP formulation
- Nicol, et al. 2015

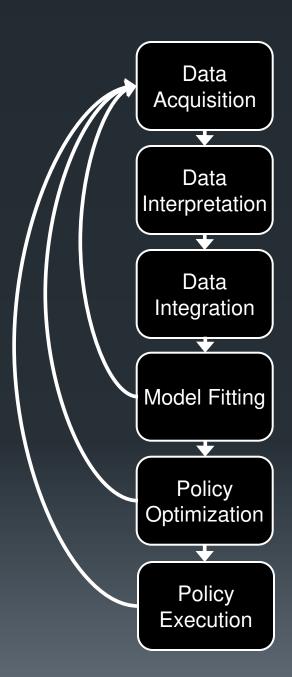


Results: Much More Successful than Existing Bottleneck Heuristic



Policy Execution

- Repeat
 - Observe Current State
 - Update Models and Re-Optimize
 - Choose and Execute Optimal Action



Summary

Locating weather stations in Africa

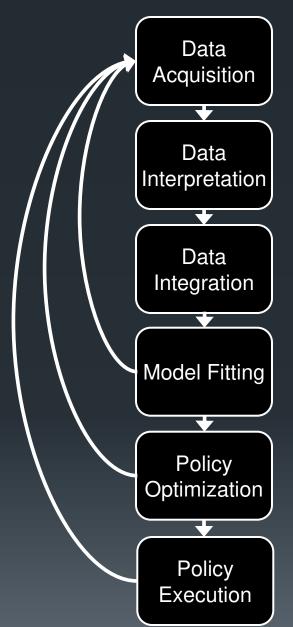
Images → Insect Species

Multiscale Data

Bird Migration Models fit to eBird Data

Where and when to purchase coastal habitat?

Action!



References

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- Lytle, D. A., Martínez-Muñoz, G., Zhang, W., Larios, N., Shapiro, L., Paasch, R., Moldenke, A., Mortensen, E. A., Todorovic, S., Dietterich, T. G. (2010). Automated processing and identification of benthic invertebrate samples. *Journal of the North American Benthological Society*, 29(3), 867–874.
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