case study of the Gulf of St. Lawrence Atlantic halibut



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10th International Flatfish Symposium Saint-Malo, November 12th, 2017



Centre for Fisheries Ecosystems Research

Marine species biotelemetry



Aquatic animal telemetry: A panoramic window into the underwater world

Hussey et al. 2015 Science, 348

Acoustic tags

- Provide direct position when individuals are in proximity o receivers
- Usually smaller spatial scale (10-100kms)
- Do not log / archive environmental data

Archival tags for fish: pop-up satellite tags (PSAT) and data-storage tags (DST)

- Provide only tagging and recapture / pop-up locations



Geolocation problem for flatfish when using satellite / archival tags For pelagic fish equipped with PSATs, positions are inferred

- For pelagic fish equipped with PSATs, positions are inferred from light intensity
 Geolocation problem for
- Geolocation problem for flatfish:
- Often distributed too deep to obtain reliable light intensity
- →Positions need to be inferred from recorded depth and temperature data (sometimes salinity)



Region specific solutions to fratish

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North Sea - Tidal location method - plaice Hunter et al. 2003. Mar. Biol. **Cloançoarcogenvironmental data (depth,** temperature, salinity) recorded by tags with regional oceanographic characteristics to infer individual position



Gulf of St. Lawrence oceanographic

Strong gradients in bathymetry and bottom temperatures

<1m

Assumptions: halibut is distributed at least once a day at the bottom.

Daily maximum depth recorded by tag corresponds to bottom and the associated temperature corresponds to bottom



Hidden Markov model (Pedersen et al. 2008. **CJFAS 65:2367-2377)** • Separation of the movement process from the observation process



X_t: unknown fish position at time t (hidden state)

- Y_t: observation at time t (depth and temperature data)
- 1: movement function: diffusion equation $\frac{\partial \phi(x,t)}{\partial t} = D\left(\frac{\partial^2 \phi(x,t)}{\partial x^2} + \frac{\partial^2 \phi(x,t)}{\partial y^2}\right)$

2: observation function: $L(z,tp|\mathbf{x}) = \int_{z-\Delta z}^{z+\Delta z} N(z; \mu_z(\mathbf{x}), \sigma_z(\mathbf{x})) \cdot \int_{tp-\Delta tp}^{tp+\Delta tp} N(tp; \mu_{tp}(\mathbf{x}), \sigma_{tp}(\mathbf{x}))$







Pop-up satellite archival tag (PSAT) data limitations



'Archived data': high resolution (2 min.) Data density: 100% (PSATmemory)

00%

N = 263,000 data per year

'Received data': lower resolution (30 or 60 min.) Data density: ~40% of 'Transmitted' (transmission gaps)

N = 4,760 data per year

Lag physical recovery using a "Goniometer" • *angle-meter* detects Argos signals



• Fisher et al. 2017. Animal biotelemetry, 5:21 -









Geolocation model validation

- **3** Methods
- <u>Simulation reconstruction of random simulated track</u>
- Observation stationary tags (known position and stationary behavior of model)
- Observation double tags (e.g. acoustic and archival) oUse for instance in Gulf of Maine - Liu et al. 2017 CJFAS 74: 1862-1877

Geolocation model validation - simulations Simulated 150 days track





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Geolocation model validation – observations

- Moored tags (2 at different locations and depths blue dots)
- mrPAT (10 double tagged large halibut throughout the Gulf red dots)





Conclusion

- Geolocation of flatfish is region specific
- Need for in depth geolocation model validation
- When possible, the physical recovery of PSAT greatly improves geolocation estimates

Advertisement

- Fully funded 2-year postdoctoral position available to work on halibut movement modeling
- www.arnaultlebris.com/PostDoc_MovementMo deling.pdf

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Observation function

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$$L(z,tp|\mathbf{x}) = \int_{z-\Delta z}^{z+\Delta z} N(z; \mu_z(\mathbf{x}), \sigma_z(\mathbf{x})) \cdot \int_{tp-\Delta tp}^{tp+\Delta tp} N(tp; \mu_{tp}(\mathbf{x}), \sigma_{tp}(\mathbf{x}))$$



Model sensitivity – structural errors? Observational likelihood

- $L(z,tp|\mathbf{x}) = \int_{z-\Delta z}^{z+\Delta z} N(z; \mu_z(\mathbf{x}), \sigma_z(\mathbf{x})) \cdot \int_{tp-\Delta tp}^{tp+\Delta tp} N(tp; \mu_{tp}(\mathbf{x}), \sigma_{tp}(\mathbf{x}))$
- Other data input possible? Light? Tide?
- Use daily variability in depth and temperature?
- Statistical assumptions: normal distributions? Other types of distribution? Model sensitivity observation errors?

Oceanographic data

• Interpolated observations? Or prediction from circulation model?

