

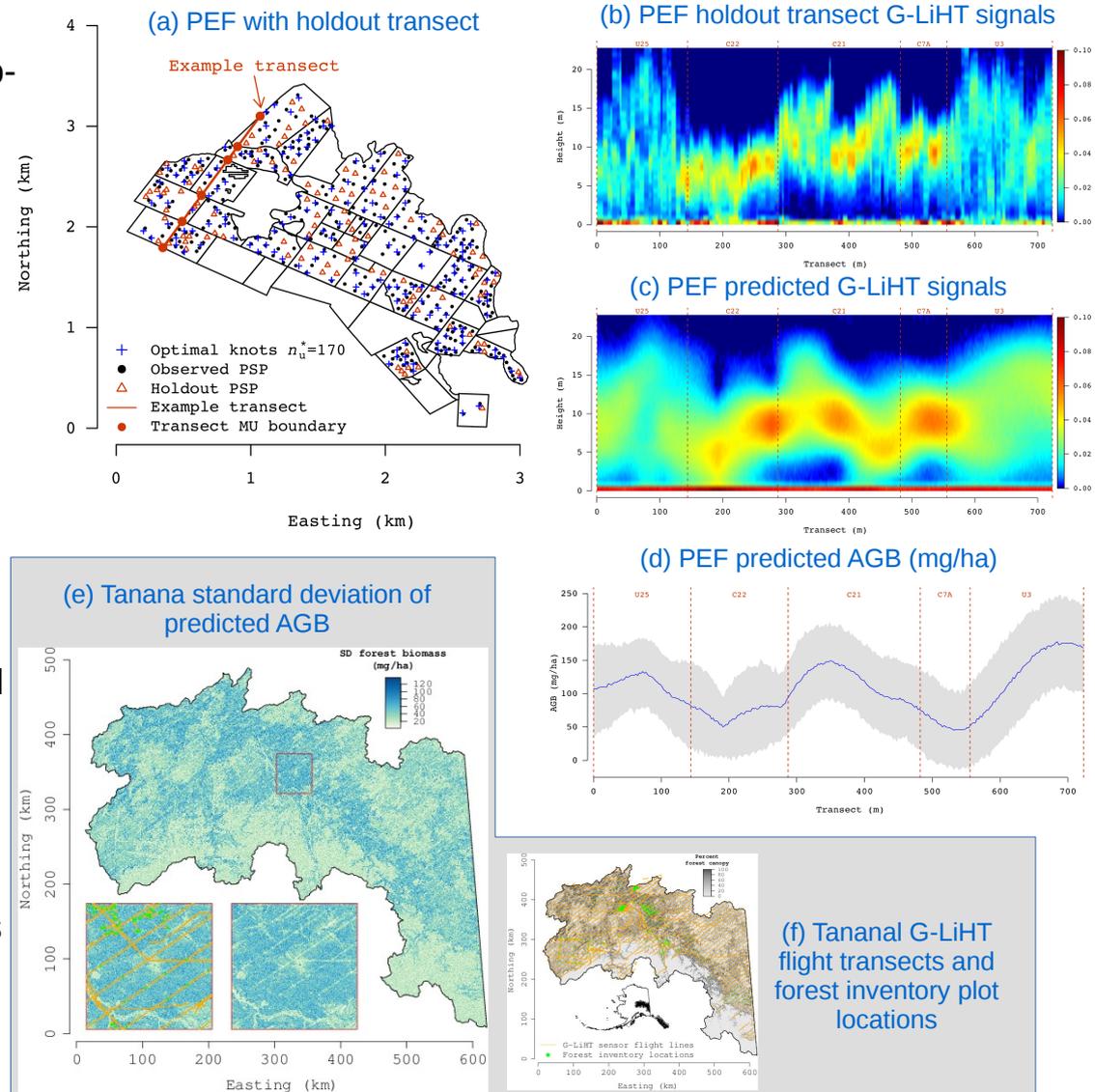
# Process-based hierarchical models for coupling high-dimensional LiDAR and forest variables over large geographic domains

**Challenge:** Few methods are available to deliver wall-to-wall forest variable prediction, e.g., above ground biomass (AGB), with associated uncertainty, using spatially sparse inventory plot data and partial coverage remotely sensed data (e.g., LiDAR transects/strips).

**New methods:** We detail a process-based Bayesian hierarchical model to jointly model AGB and LiDAR signals. The proposed framework address:

- (1) spatial misalignment between the AGB observations and LiDAR signals,
- (2) optimal and automated selection of LiDAR signal covariates,
- (3) prediction of both AGB and LiDAR signal at any location in the domain,
- (4) prediction uncertainty quantification for both AGB and LiDAR signals,
- (5) high-dimensional datasets.

**Significance:** Given the anticipated availability of sampling LiDAR and hyperspectral data products (e.g., not wall-to-wall coverage), the proposed model provides a statistically valid approach to use information where available to deliver improved wall-to-wall prediction products with uncertainty.



**Illustrations:** The proposed model was assessed using LiDAR data acquired from NASA G-LiHT and field inventory data from the Penobscot Experimental Forest (PEF) in Bradley, Maine and Tanana, Alaska.

CMS publication from: Hurtt-03, Dubayah-04, Morton-02, Cook-B-01

Finley, A.O., S. Banerjee, Y., Zhou, B.D. Cook. (2016) Process-based hierarchical models for coupling high-dimensional LiDAR and forest variables over large geographic domains. <http://arxiv.org/abs/1603.07409>