The New York Forest Carbon Assessment

Outputs, Applications & Insights for Adirondack Research & Stewardship

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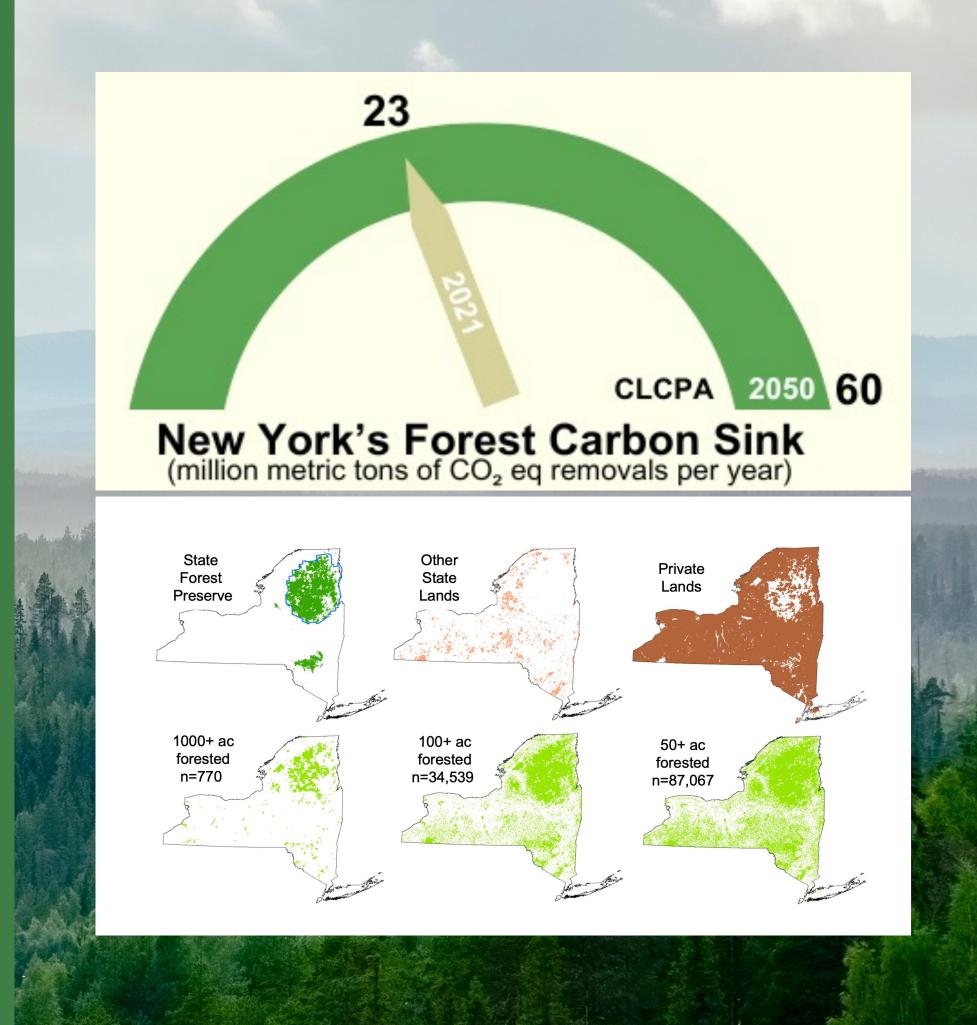






Overview

	 Objective
Background	 Approach
	• AGB & C Mapping
	 Stock-Change
	 Monitoring
Functions	 Change Detection
& Outputs	 Uncertainty
	 Parcel-level analysis
	Parcel-level analysisSolar facility siting
Applications	 Solar facility siting
Applications	Solar facility sitingStatewide MRVGHG Inventory
Applications	 Solar facility siting Statewide MRV GHG Inventory Disturbance trends
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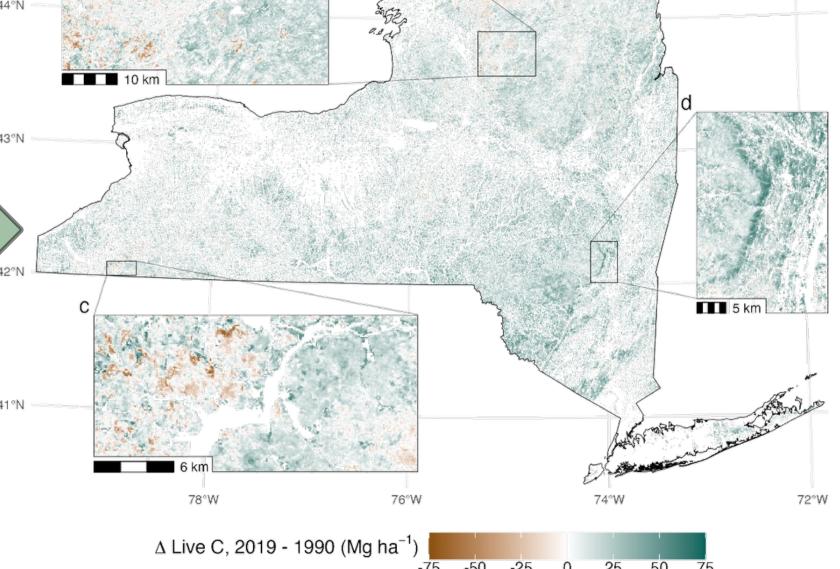


New York Table 349: Net CO₂ Flux from Forest Pools in Forest Land Remaining Forest Land (MMT CO₂ Eq.), New York Self-(Organic) Table 350: Net C Flux from Forest Pools in Forest Land Remaining Forest Land (MMT C), New York Aboveground Biomass Belowground Biomass Dwad Wood Self (Mineral) Said (Organic) Drained Organic Soil* + Absolute value does not exceed 0.05 MMT C No observations of drained organic seil available These estimates include C stock changes from drained organic soils from both Firest Lond Remaining Forest Lond and Land Converted to Firest Lond

Table 351: C Stocks in Forest Land Remaining Forest Land (MMT C), New York

Carbon Poels	3990	3981	1992	1965	1994	1995	1996	1997	1998	1989	2000	2001	2002	2906	2004	2005	2006	300T	3088	3089	2030	2011	2812	2013	2014	2015	3016	2017	2008	2019	2929
Forest Rossystem	1,698	1,706	1,713	1,721	1,728	1,736	1,743	1,751	1,758	1,766	1,773	1,780	1,787	1,795	1,802	1,809	1,816	1,822	1,830	1,837	1,844	1,851	1,858	1,865	1,872	1,879	1,886	1,892	1,899	1,906	1,912
Aboveground Bismuss	424	429	485	440	445	451	457	462	46T	473	476	-865	-869	494	459	904	548	51.4	519	584	529	534	530	544	548	554	559	564	569	574	578
Belowground Biomass	62	65	84.	85	86	87	58	88	90	93	92	95	54	95	96	97	26	99	180	101	182	165	104	305	306	107	308	189	130	LIL	111.
Doad Wood	39	40	41.	42	42	43	-84	45	-65	46	41	45	49	49	59	51	52	50	58	54	95	96	56	57	58	59	60	90	61	62	63
Litter	100	180	100	100	200	104	200	200	101	181	181	102	102	102	302	300	102	103	188	188	100	103	100	200	204	104	104	184	184	164	106
Soil (Mismal)	1,043	1,048	1,048	1,043	1,843	1,843	1,043	1,043	1,048	1,048	1,048	1,043	1,040	1,844	1,644	1,004	1,044	1,044	1,044	1,044	1,944	1,844	1,844	1,061	1,064	1,044	1,045	1,045	1,045	1,945	1,845
Stall (Organic)	1.1	11	11	11	11.	11.	11	11	1.1	11	11	11.	11	11.	11.	11	11	1.1	11	11	11.	11.	11.	11.	11	1.1	1.1	11	11	11	11.
Notes: The franct accounts	n Catal	a de not i	and taking the	and one make	france he	al face of	coloneto	e escriborne	wall age to	ment see	el. Thank	a manage most	eren dan	to indepe	adapt son	adina Da	market kom e	ed important	teritories	nday Pilk	data see	servered t	- nemerous	et attaches s	a of Town	enct of the	e Toneste	er man l	Day is the	and arms	ad abanco

in stack. Thus, an estimate of flux for 2019 requires estimates of C stacks for 2019 and 2020



Objective

To build a map-based forest carbon accounting system for NYS that replicates USFS/EPA methods for GHG inventory and supports scalerelevant public and private-sector decision-making for climate solutions

Notes. The funct occurrence C stock changes do not include those on non-funct hand [o.g., agredoment areas]. Providence indicate not C untake [i.e., a net removal of C from the atmosphere). Total net flux is an estimate of the actual net flux between the total forest of pool and the atmosphere. Totals may not sum due to independent rounding.

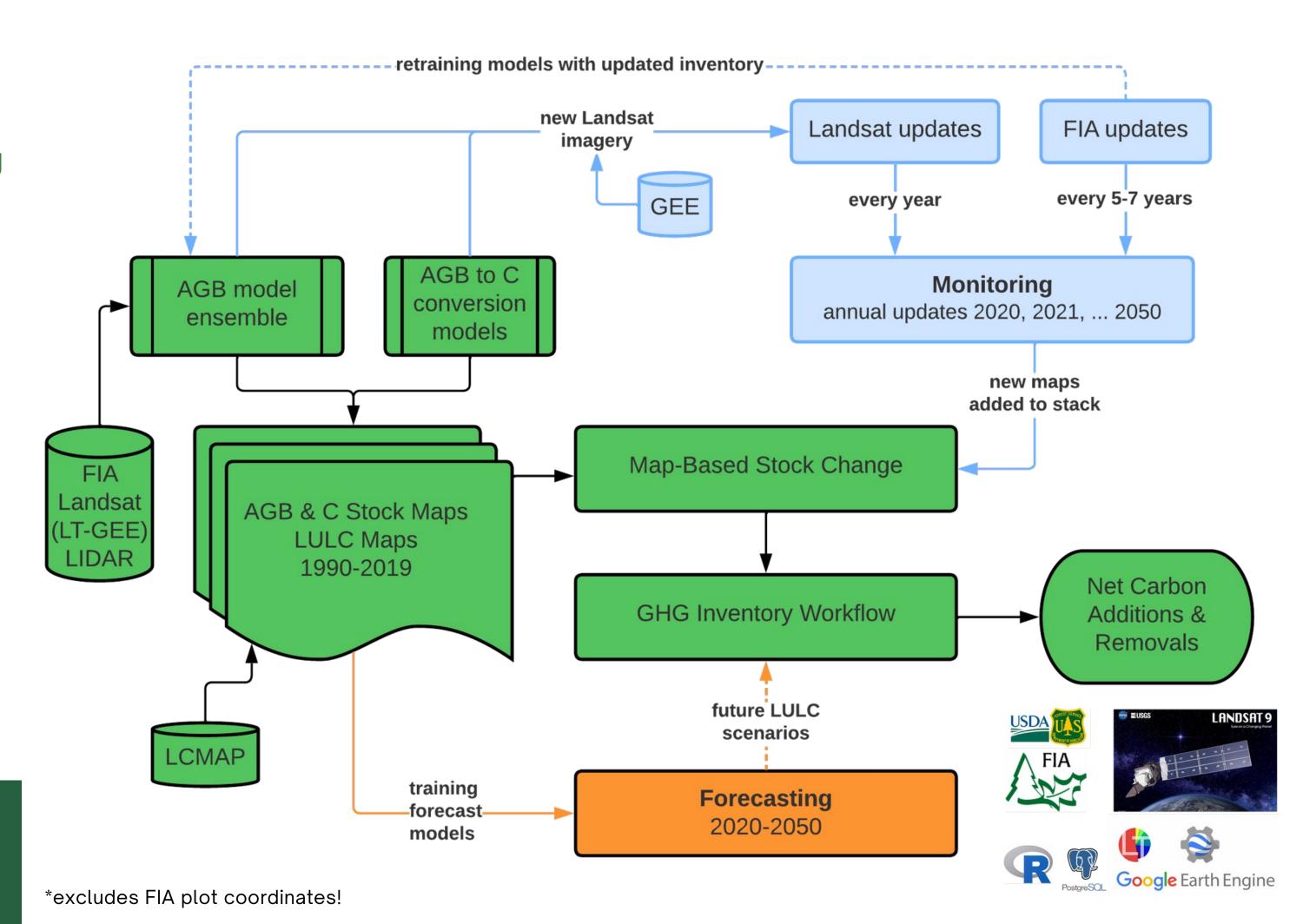
Integrated basis for

- AGB & C Mapping
- Stock Change
- GHG Inventory
- Monitoring
- Forecasting

Built on freely available data* and open-source computational tools



Approach



Ensemble machine-learning models, trained on FIA plots and EO data (LIDAR/ALS, Landsat C2, LT-GEE)

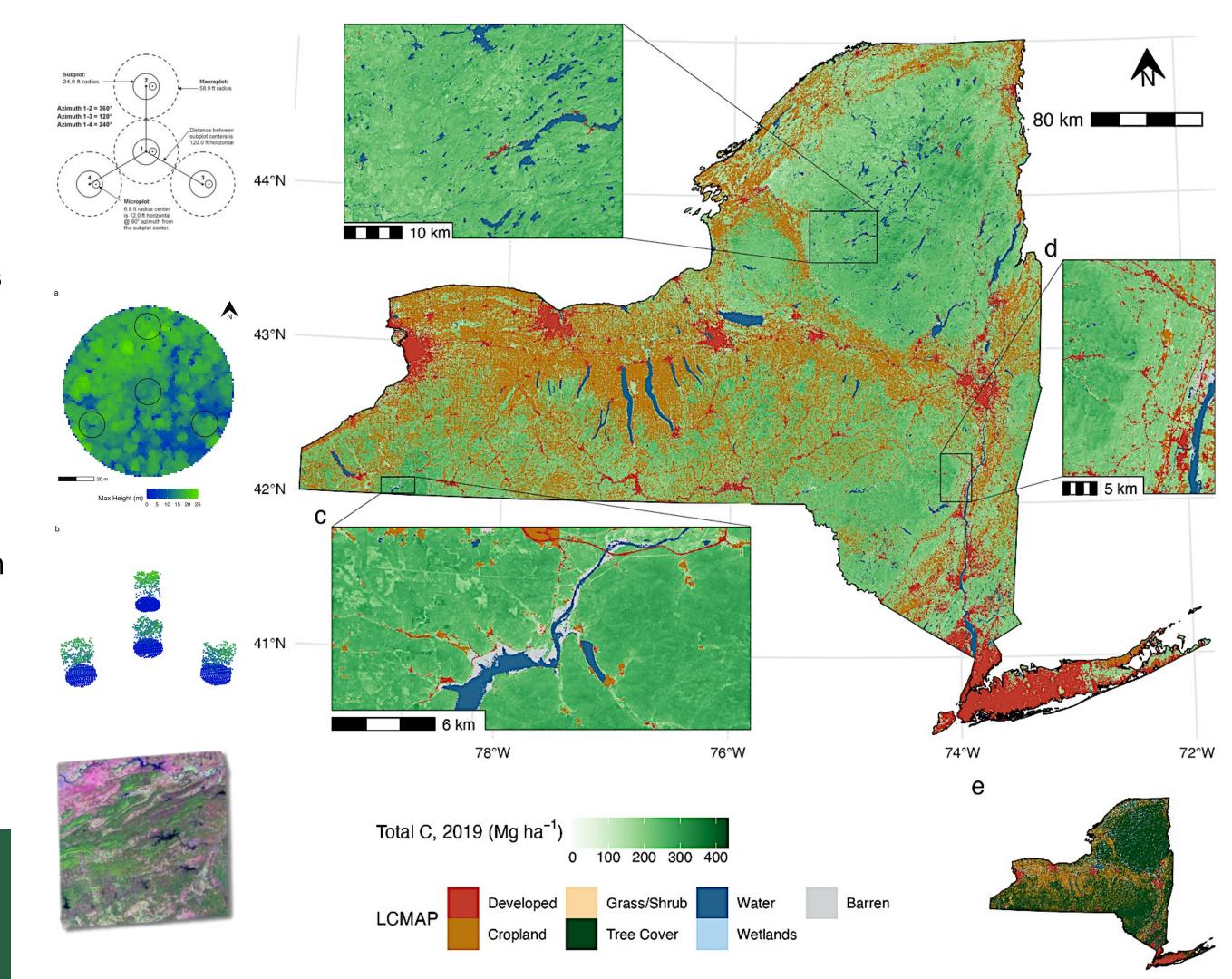
Created 30m annual AGB maps 1990 to 2019 statewide

AGB maps converted to 'live C' maps using FIA allometrics

Soil, litter, deadwood C pools mapped with models trained on FIA Phase 3 and EO data

Annual 30m C pools mapped only for forested pixels

AGB & C Mapping

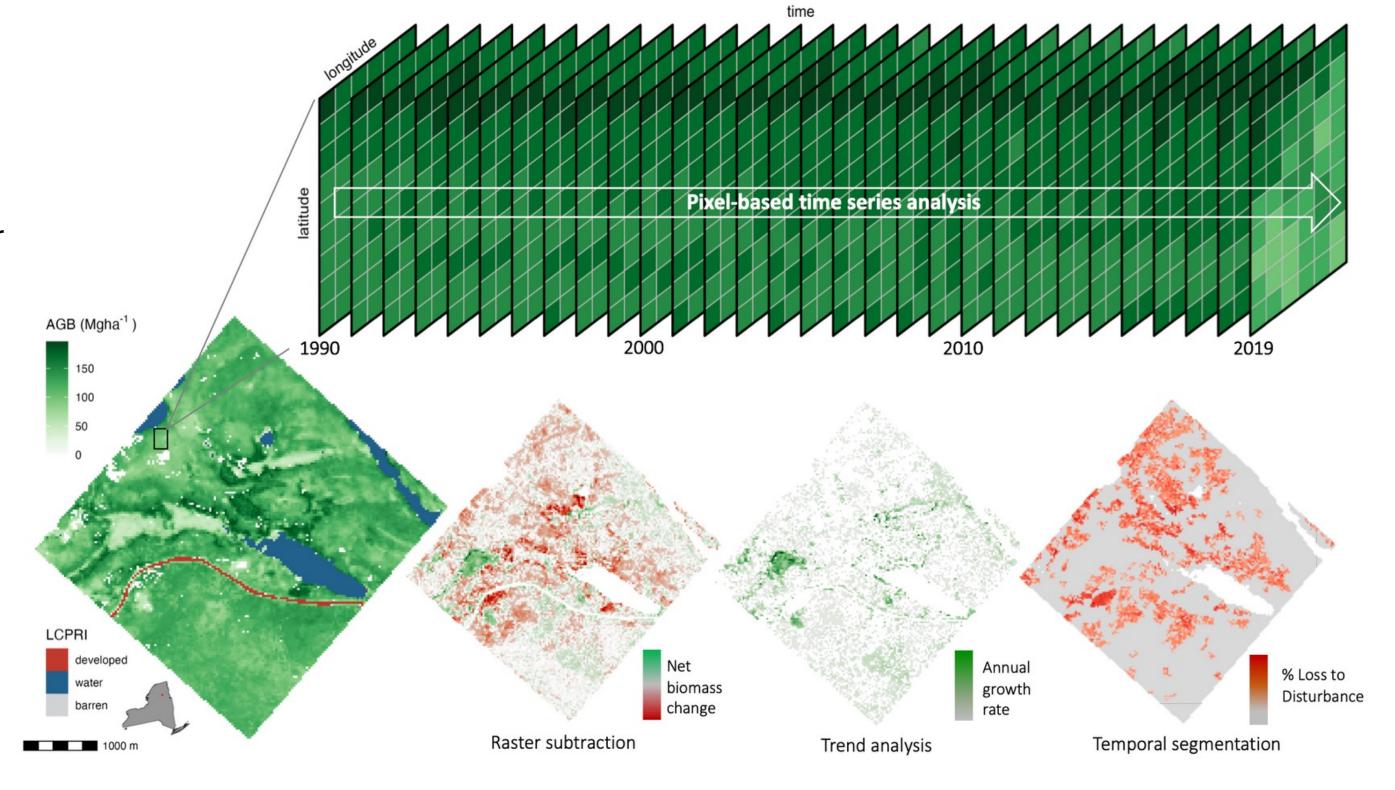


Compiled raster stacks of 30m annual AGB/C maps

Flexible basis for stockchange analysis using raster operations:

- simple differencing
- trend analysis
- temporal segmentation

Next step: spatiotemporal segmentation for object-based (i.e., stand-level) analysis and forecasting



Stock-Change

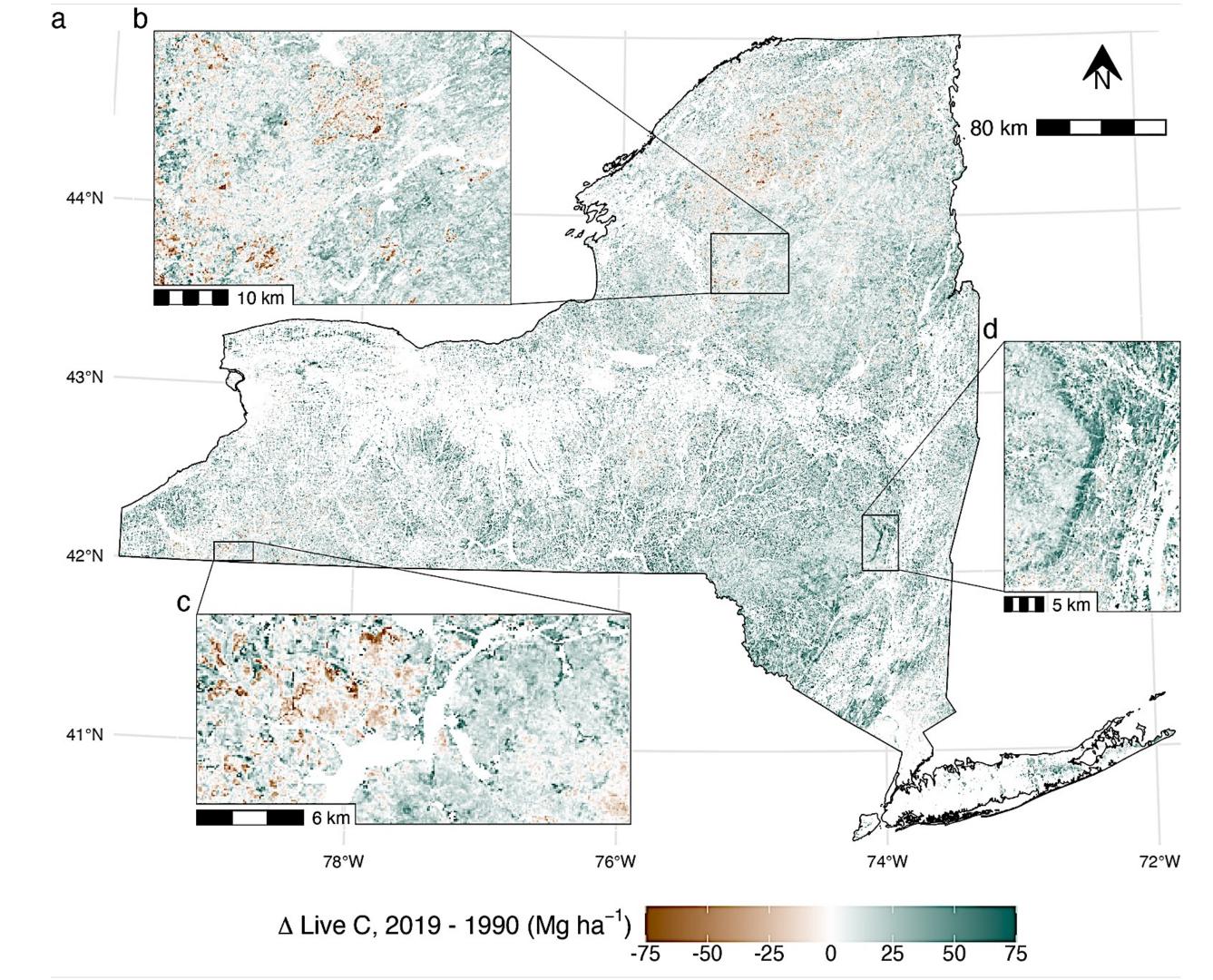
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Stock-Change



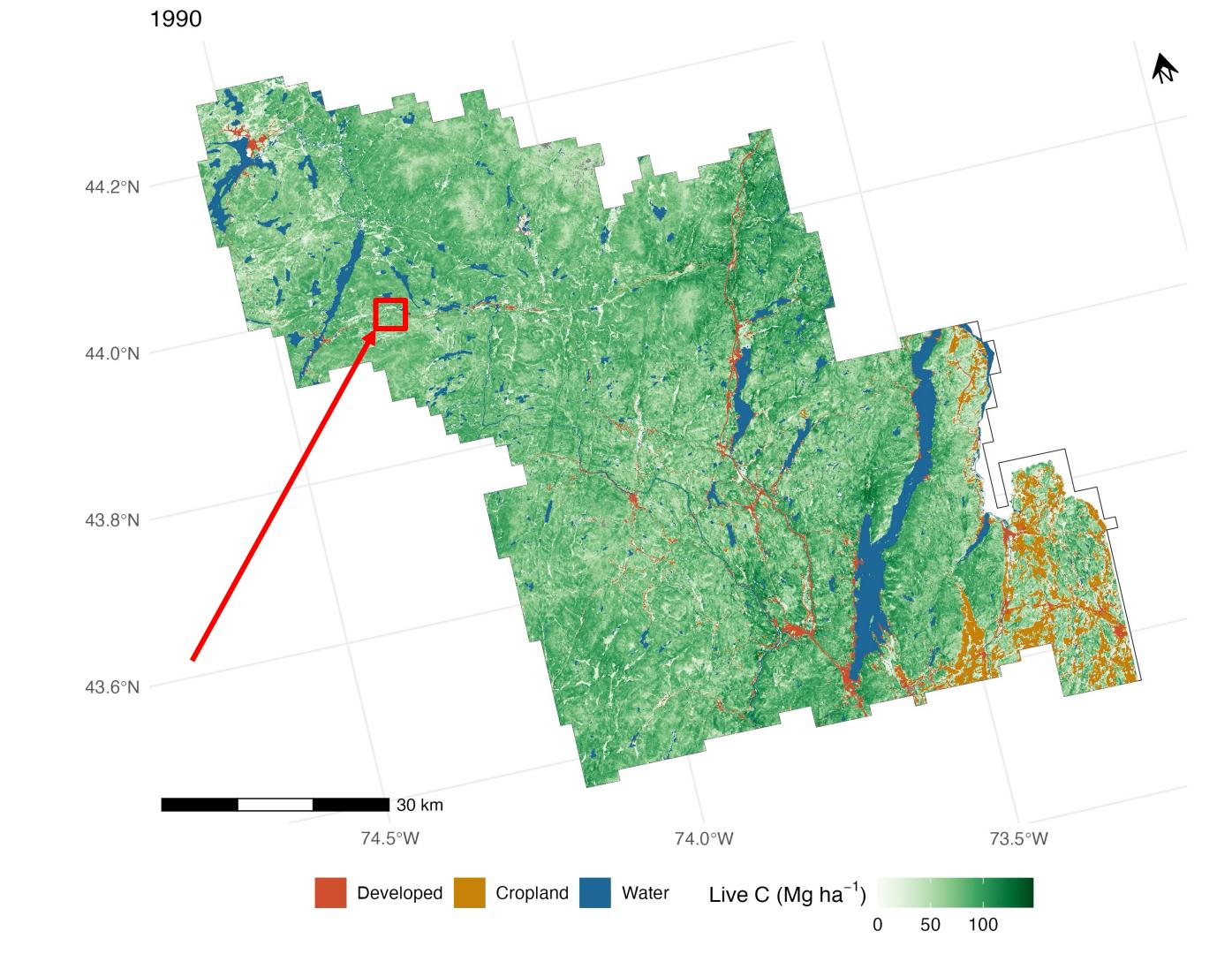
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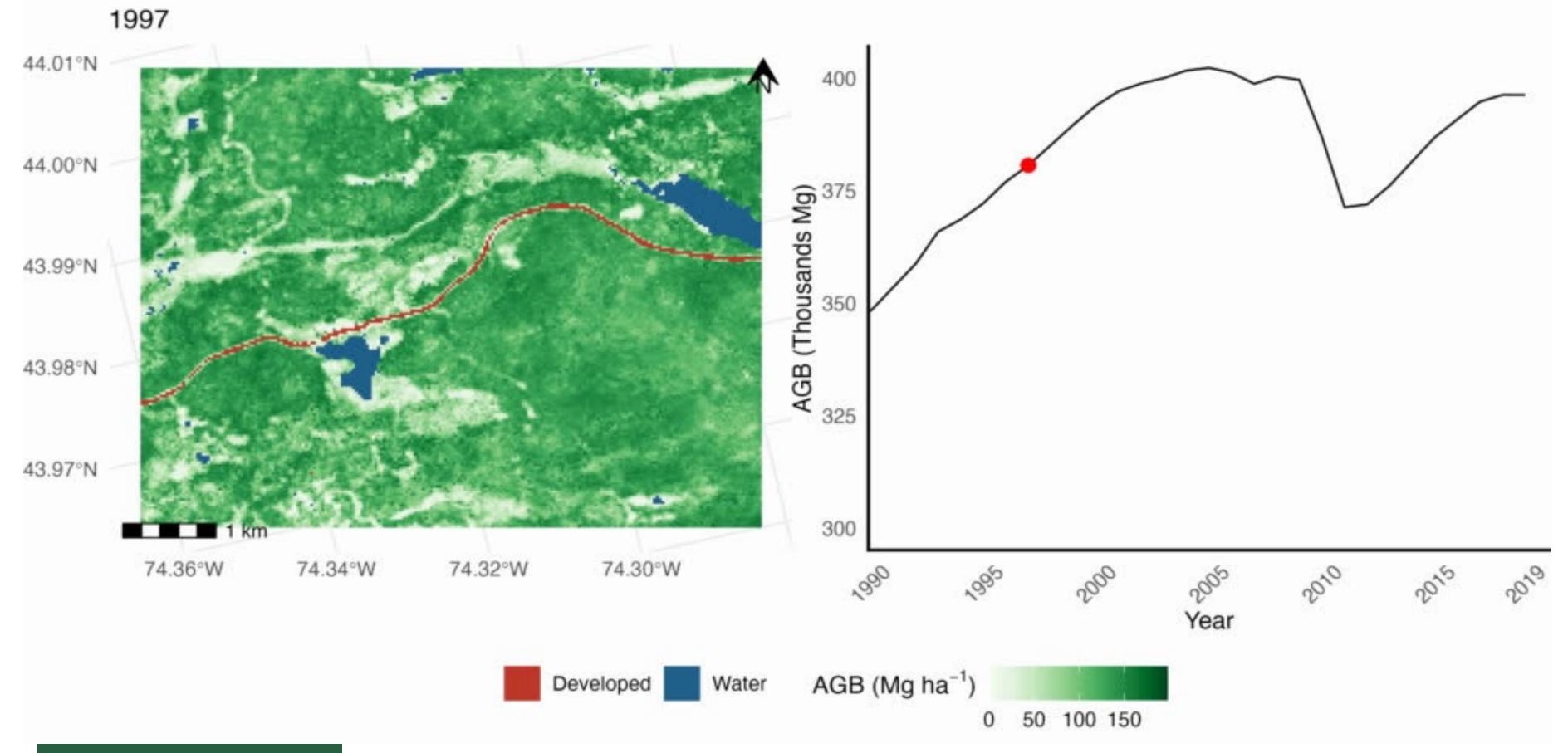
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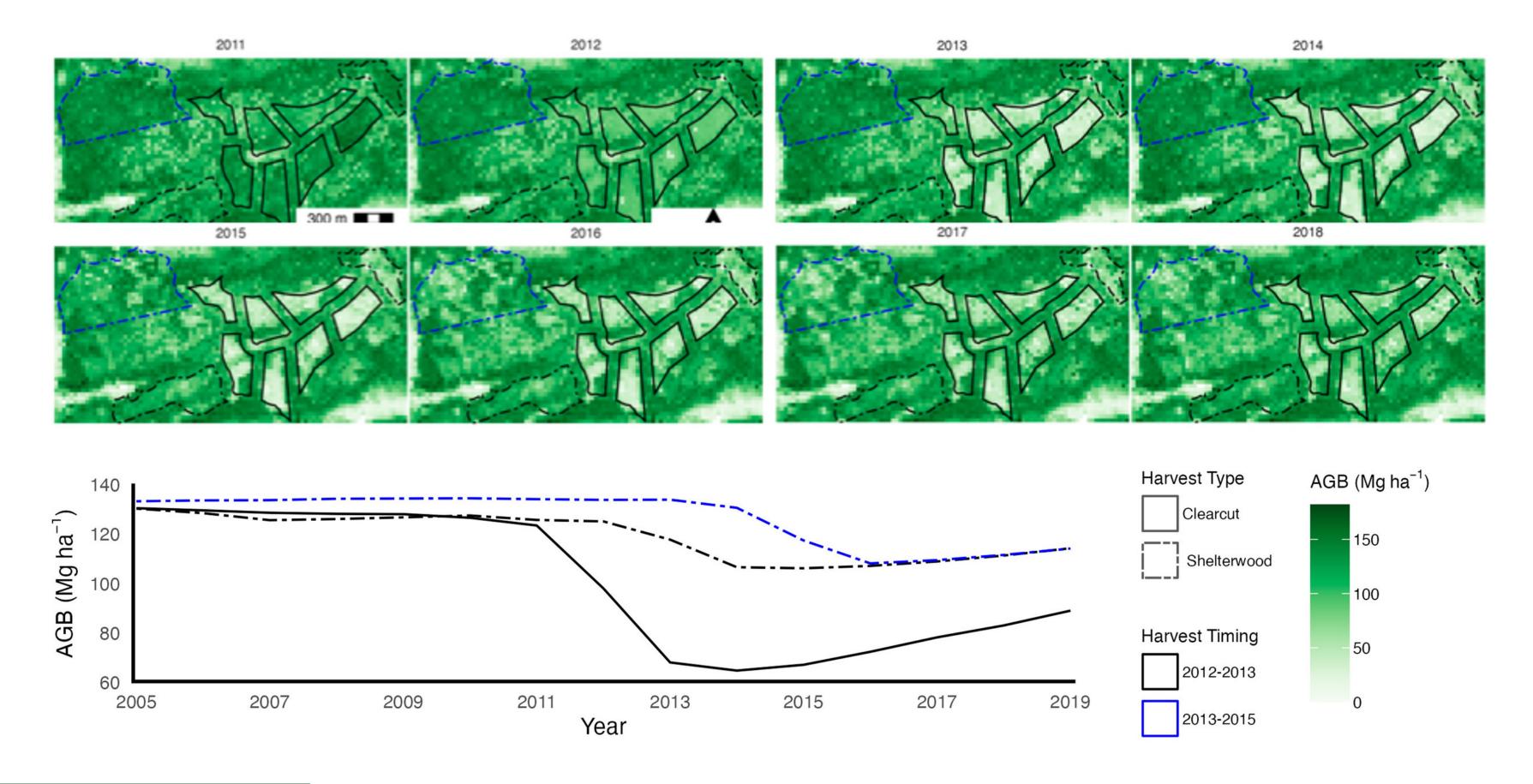






Monitoring

As new Landsat imagery is released, we use the same ensemble models to produce annual 'update' maps of AGB and C stocks. The update maps are added to our raster stacks and stock-change metrics are recalculated. Core models will be re-trained and ensembles updated based on the FIA panel schedule (i.e., every 7 years in NYS).



Monitoring

Time-series maps accurately capture harvest and regeneration dynamics, based on ground-referencing with management records provided by industrial landowners

Models likely underpredict biomass removals and overpredict early regeneration rates

Ground-truthed change detection tools based on Landsat time-series imagery and harvest maps

Landtrendr on Google Earth Engine (LT-GEE) was 'tuned' on ~400K ac of ADK harvest maps/records provided by industrial landowners

Trained a change attribution model based on spatial and temporal attributes of harvests vs. natural disturbances (on Forest Preserve)

Can classify harvest/non-harvest with ~95% accuracy, but cannot classify among 'natural' causes

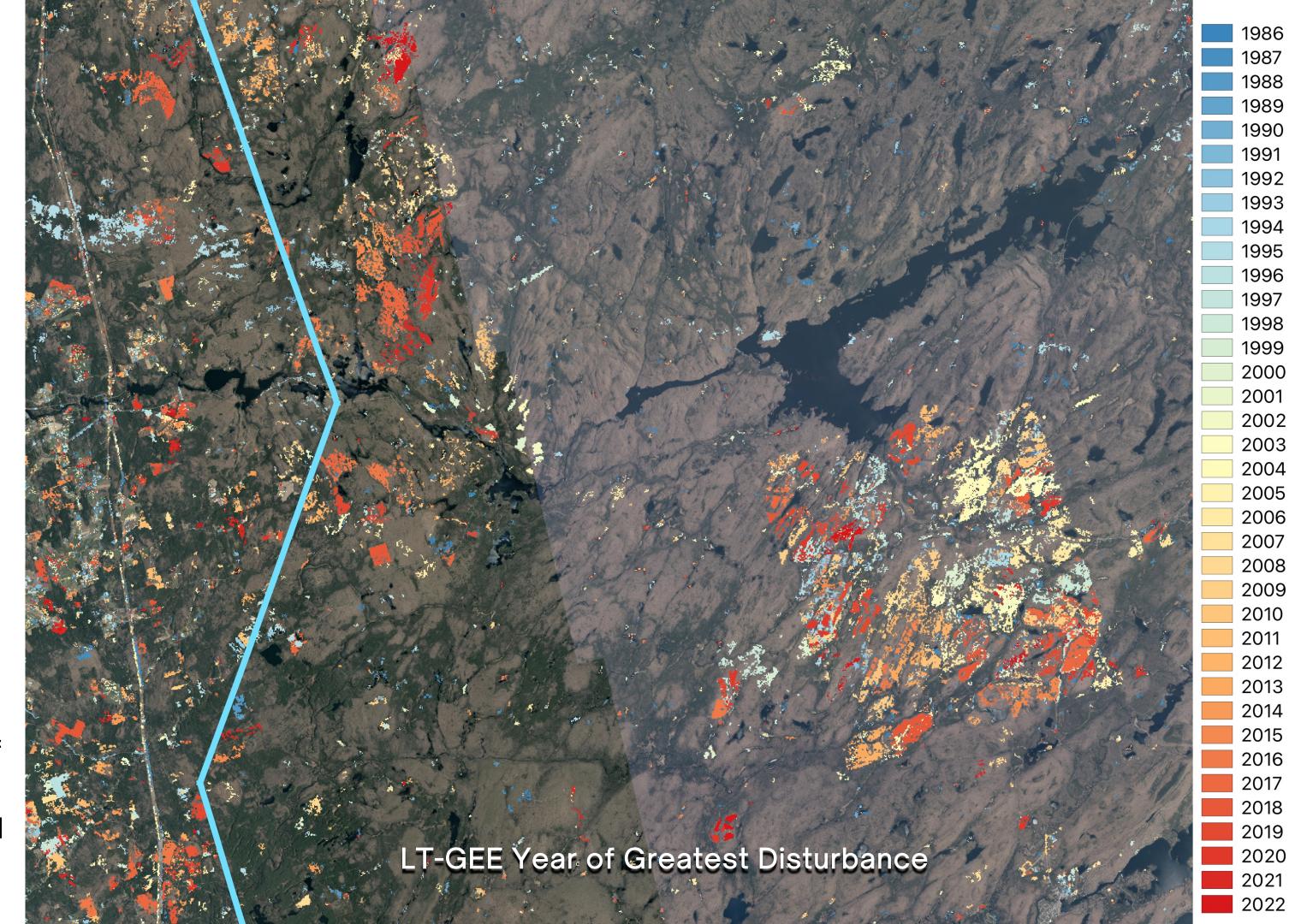
LT-GEE Year of Greatest Disturbance 1990 2010 2020 2000 **NYS Forest** Preserve **ESF** Huntington Wildlife Forest Üpper Hudson Woodlands

Change Detection

Change detection and attribution tools allow mapping of historical land use and disturbance (back to 1990)

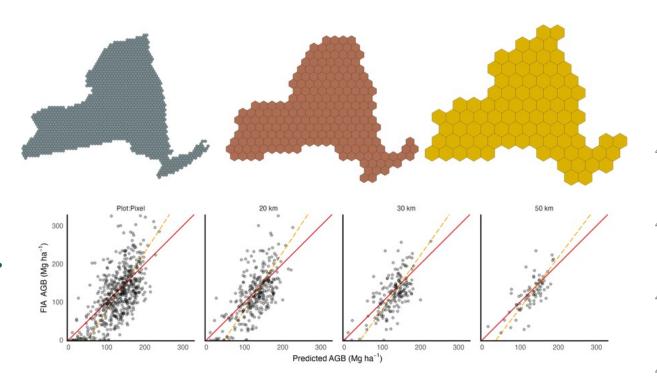
Reconstruct recent patterns of forest management, including areas with multiple entries

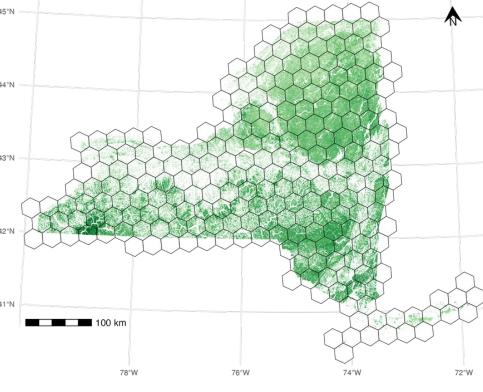
Next step: training models to estimate harvest yields by species & grade, for use in spatial lifecycle assessment of the GHG balance of NYS harvested wood products (HWP)



Model

Prediction accuracy on 70/30 training holdout plots; k-fold cross-validation, hyperparameter tuning; ensembling



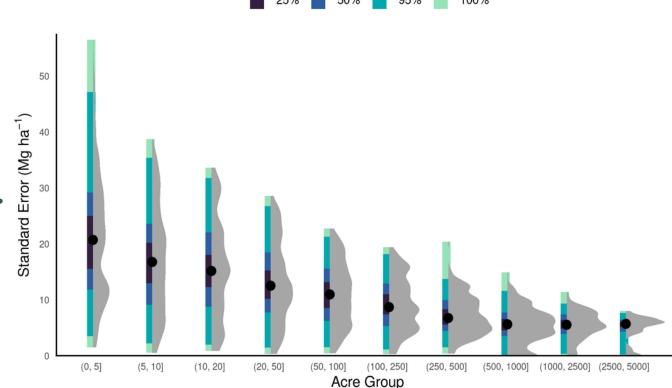


Map

Plot-to-pixel comparisons aggregated to variable scales and LULC-corrected validation against FIA small-area map estimates (60,000 ha hexagons)

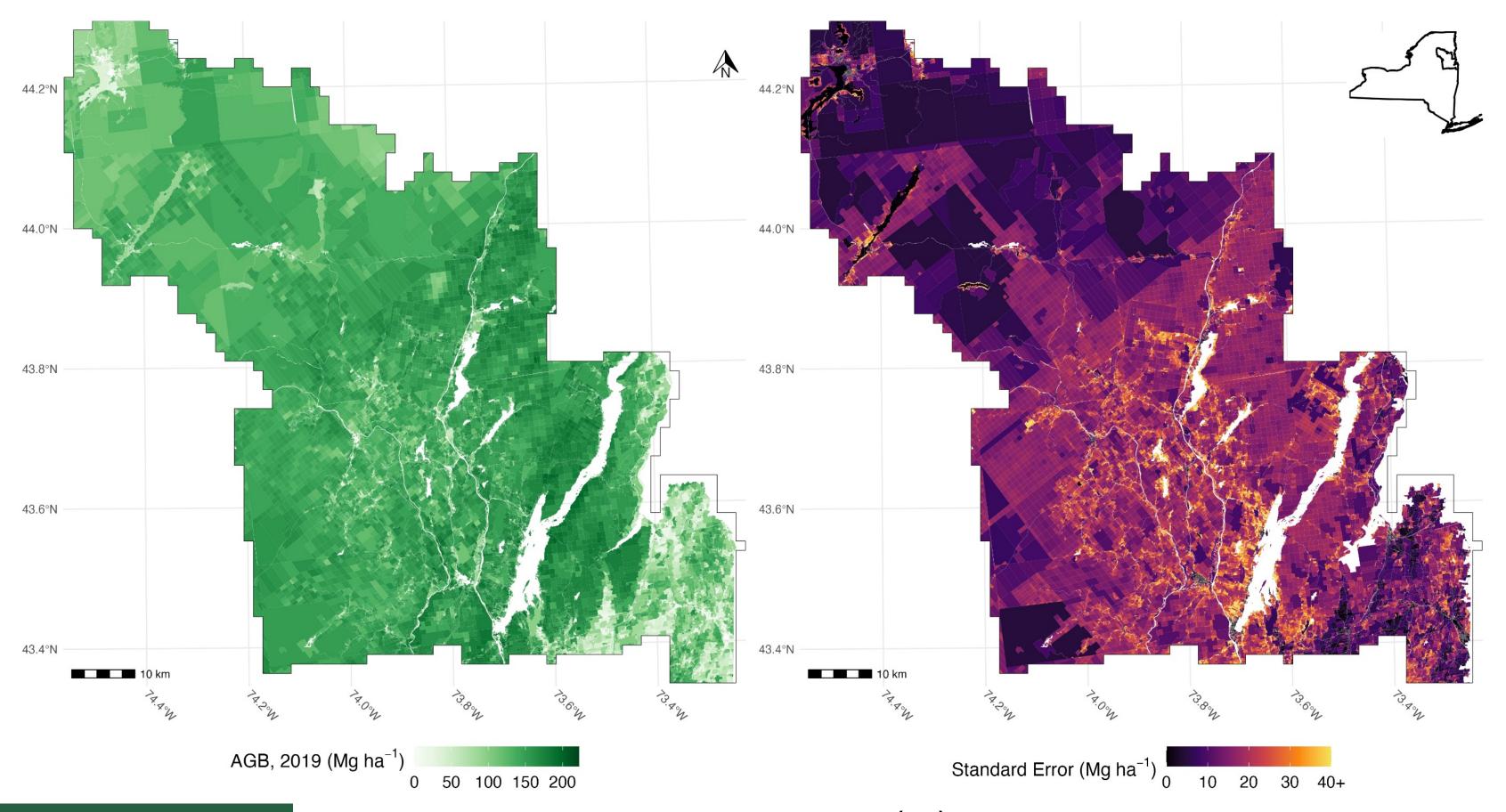


Bootstrapped cumulative uncertainty from reference plots, allometrics, model residuals, spatial autocorrelation, etc;



FIA Estimate Model Prediction

Uncertainty



Uncertainty

We can rapidly estimate the standard error (SE) of our map predictions of biomass and carbon stocks for individual parcels (and any other arbitrary areas within NYS)

Provides easy to understand 'error bars' for decision-support applications

Parcels

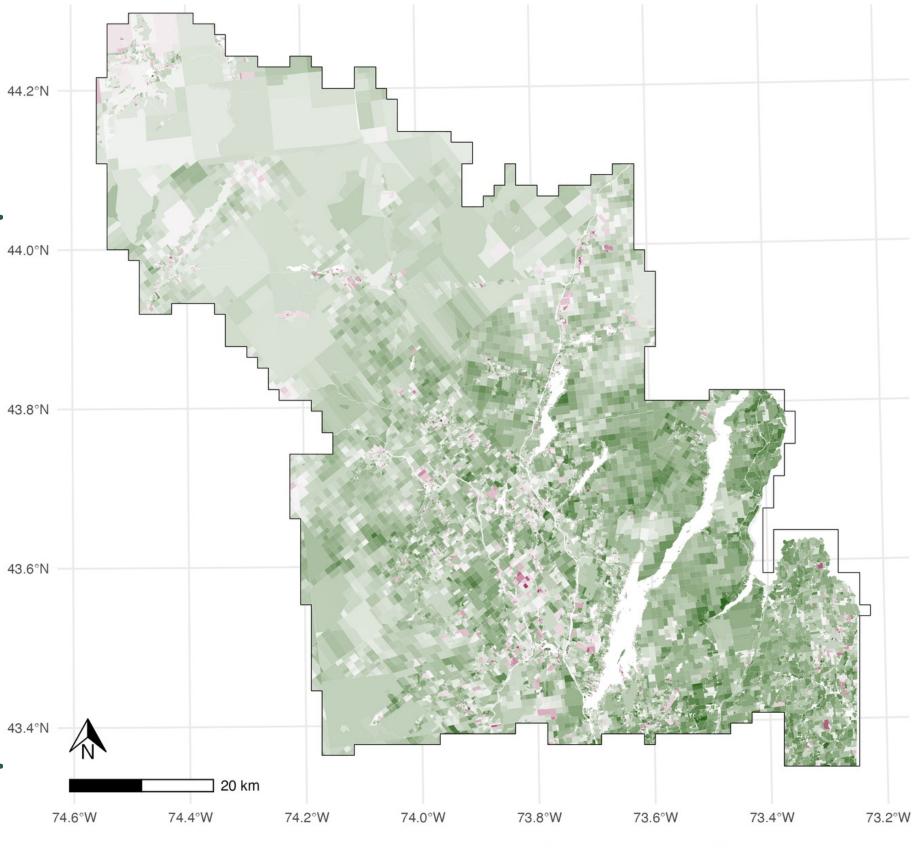
Screening/prioritizing forest parcels for enrollment in voluntary programs, tax incentives, conservation easements, land acquisitions, offsets, etc.

MMRV

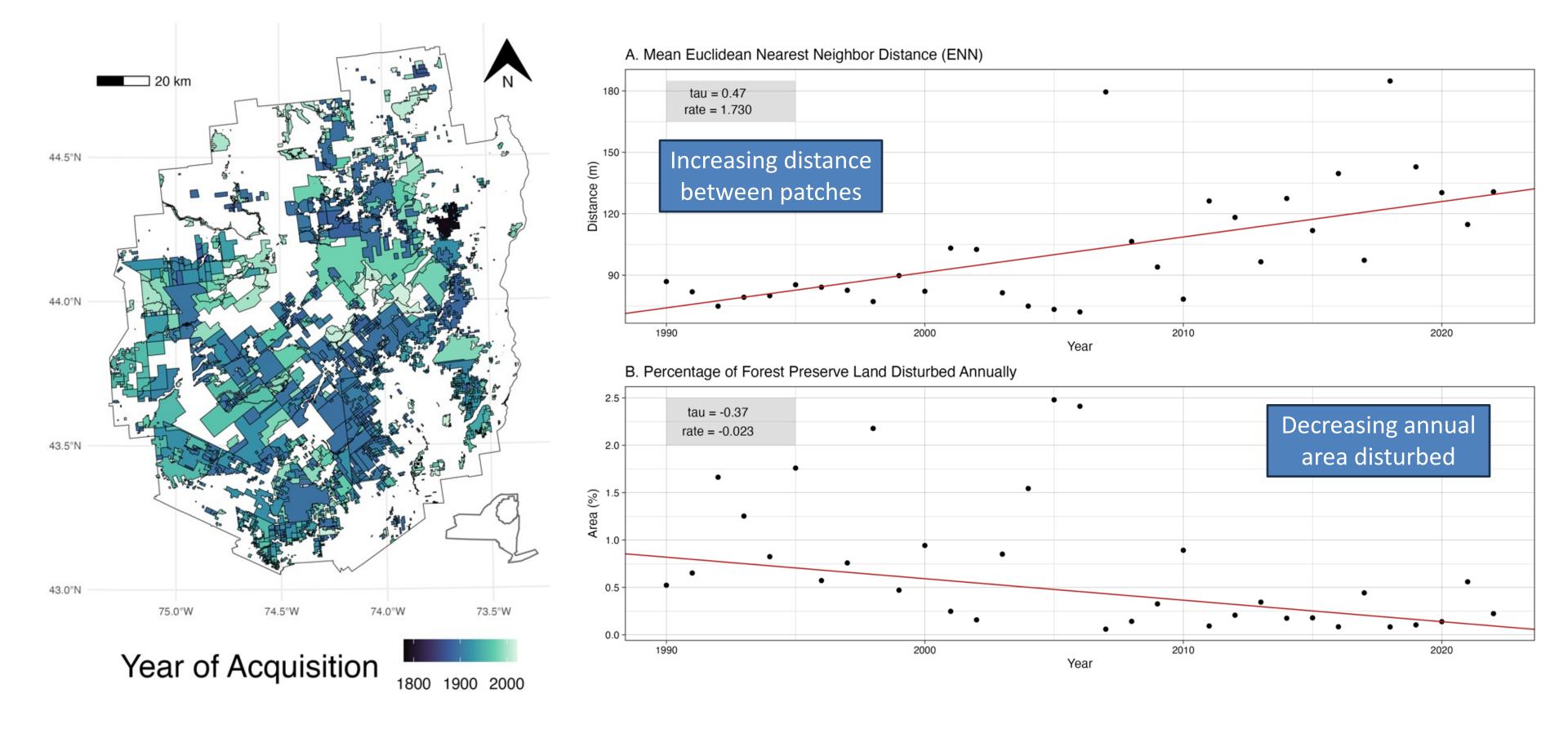
NY Connects: USDA Climate Smart Commodities project will invest \$13M into forestry practices via DEC's RegenerateNY program over next 5 years

Solar

Rapid assessment of proposed solar facilities for NYSERDA's Solar Scorecard, to minimize forest loss, C emissions due to land clearing for solar farms



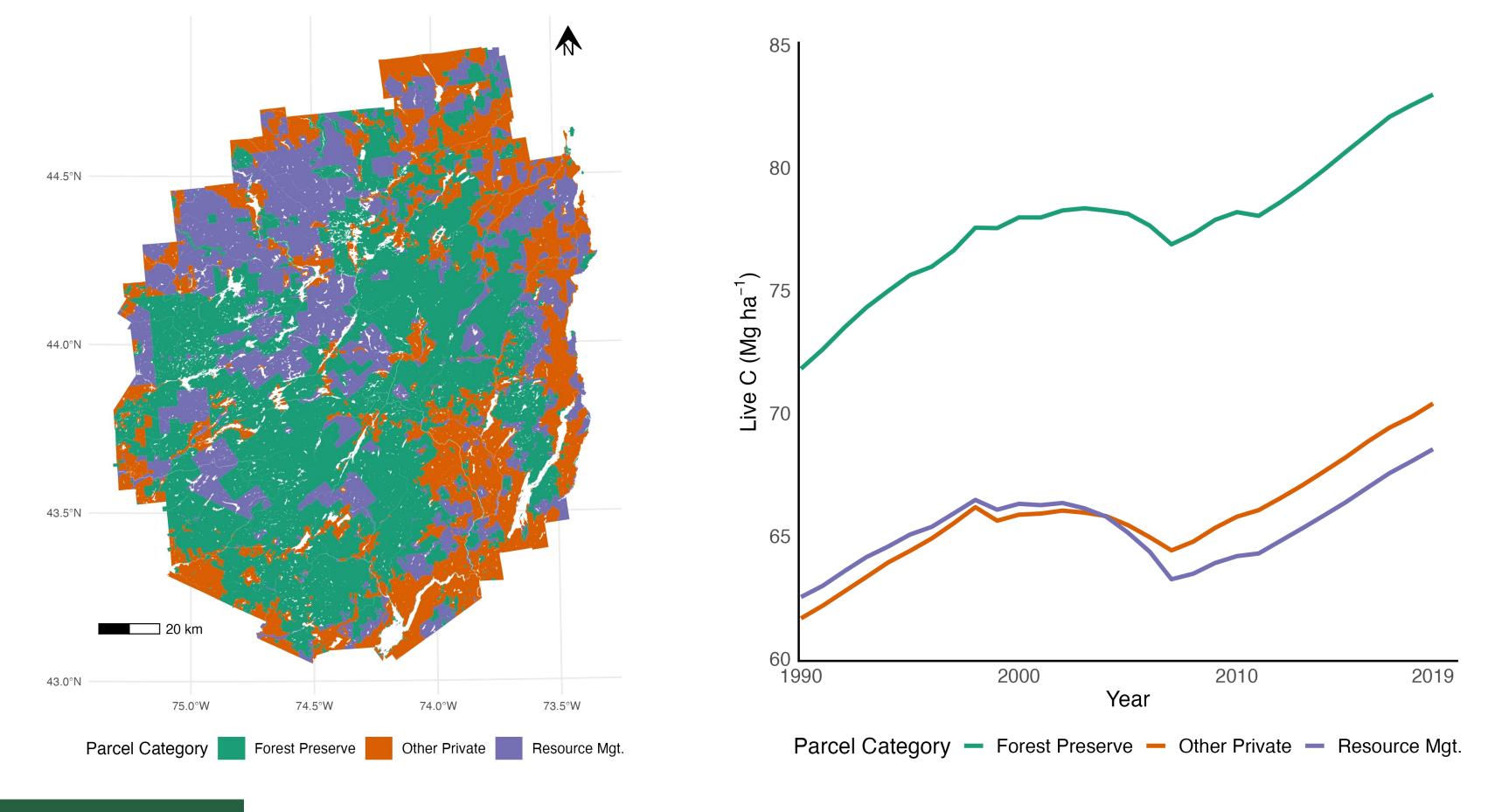
 Δ Live C, 2019 - 1990 (Mg ha⁻¹)_{-50 -25 0 25 50+}





We reconstructed forest disturbance history and assessed trends across the Forest Preserve.

Contrary to expectations, we found no evidence of an intensifying disturbance regime, but observed decreasing frequency of canopy loss (based on annual % area disturbed) since 1990.





On average, Forest Preserve lands have greater live C stocks than privately-owned ADK lands

Based on stock-changes, working and reserve lands across the ADK Park have <u>very similar</u> <u>average rates of C sequestration</u> (net CO2 eq removals) over the last 15 years

Thanks. Questions?

Contributors

Steve Stehman (ESF), Eddie Bevilacqua (ESF), John Battles (Cal), Michelle Brown (TNC-NY), David Saah (UCSF/SIG)

Partners

NYS DEC, US Forest Service, Lyme Adirondack Forest Co, F&W Forestry, NYS GIS Program Office (GPO)

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