



**Laboratory of Forest Geomatics** 

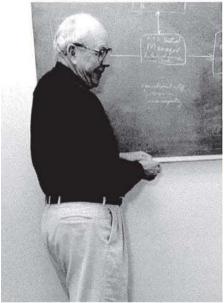
Stima spazialmente esplicita di variabili forestali inventariali tramite telerilevamento



Dr. Walter Mattioli



Dr. Daniel Goodman Horvitz 1921 - 2008



Dr. Donovan J. Thompson 1919 - 1991

#### A GENERALIZATION OF SAMPLING WITHOUT REPLACEMENT FROM A FINITE UNIVERSE\*

D. G. HORVITZ† AND D. J. THOMPSON

Iowa State College

This paper presents a general technique for the treatment of samples drawn without replacement from finite universes when unequal selection probabilities are used. Two sampling schemes are discussed in connection with the problem of determining optimum selection probabilities according to the information available in a supplementary variable. Admittedly, these two schemes have limited application. They

including both the sample design and the method of estimation. One sampling system is said to be more efficient than another if the variance or mean square error of the estimate with the first system is less than that of the second, provided the cost of obtaining the data and results is the same for both. The development of stratified, multi-stage, multiphase, cluster, systematic, and other sample designs beyond

The possibility of using unequal probabilities for selecting the sample elements from the universe as a means of increasing precision perhaps received its first impetus for applied sampling from Hansen and Hurwitz [2] in 1943. They introduced the selection of primary units (in a subsampling scheme) with probabilities proportionate to some measure of their size and presented the appropriate theory. Their sampling scheme was confined (when sampling without replacement) to samples of one primary unit per stratum, however, the theory not having been extended beyond this point. More recently, Midzuno [6] has generalized the Hansen and Hurwitz approach to sampling a combination of n elements of the universe with probability proportionate to some measure of size of the combination. Madow [5] has made some contributions to the theory of the systematic selection of several clusters with probability proportionate to a measure of size.

<sup>\*</sup> Journal Paper No. J 2139 of the Iowa Agricultural Experiment Station, Ames, Iowa, Project 1005. Presented to the Institute of Mathematical Statistics, March 17, 1951.

<sup>†</sup> Now at the University of Pittsburgh.

Tabella 1.3.1 - Valori totali e per unità di superficie del volume del fusto e dei rami grossi per le categorie inventariali Boschi alti, Impianti di arboricoltura da legno e Aree temporaneamente prive di soprassuolo e per la macrocategoria Bosco

Distretto territoriale	Boschi alti				Impianti di arboricoltura da legno				Aree temp. prive di soprassuolo				Totale Bosco			
	Volume	ES	Volume	ES	Volume	ES	Volume	ES	Volume	ES	Volume	ES	Volume	ES	Volume	ES
	(m <sup>3</sup> )	(%)	(m <sup>3</sup> ha <sup>-1</sup> )	(%)	(m <sup>3</sup> )	(%)	(m <sup>3</sup> ha <sup>-1</sup> )	(%)	(m <sup>3</sup> )	(%)	(m <sup>3</sup> ha <sup>-1</sup> )	(%)	(m <sup>3</sup> )	(%)	(m <sup>3</sup> ha <sup>-1</sup> )	(%)
Piemonte	126 821 547	3.1	151.0	2.9	2 947 269	15.7	103.2	12.7	7 613	60.2	3.3	45.9	129 776 430	3.0	149.1	2.8
Valle d'Aosta	15 334 302	7.6	156.0	6.9	0	-	0	-	0	-	0	-	15 334 302	7.6	155.8	6.9
Lombardia	105 423 629	4.2	182.4	3.9	2 613 095	19.6	97.4	17.8	0	-	0	-	108 036 723	4.1	178.3	3.9
Alto Adige	104 721 523	4.6	315.0	4.3	0	-	0	-	467 004	31.4	109.6	14.7	105 188 527	4.6	312.4	4.3
Trentino	105 715 538	4.8	283.5	4.6	0	-	0	-	61 011	97.3	24.2	89.8	105 776 549	4.7	281.8	4.6
Veneto	80 931 420	4.6	204.7	4.2	260 012	45.4	124.4	25.6	4 529	100.0	13.4	-	81 195 960	4.6	204.1	4.2
Friuli V.G.	67 066 949	5.4	212.1	5.1	763 052	26.8	100.3	18.7	0	-	0	-	67 830 001	5.3	209.5	5.0
Liguria	49 379 829	4.5	147.3	4.3	39 233	100.0	107.1	-	19 730	56.8	5.7	49.0	49 438 791	4.5	145.8	4.3
Emilia Romagna	71 063 339	3.9	128.7	3.7	1 274 428	20.6	130.8	13.1	356	72.6	0.3	53.7	72 338 122	3.9	128.4	3.6
Toscana	130 873 621	3.1	129.9	3.0	1 042 114	43.2	189.7	34.7	40 250	66.0	15.6	60.5	131 955 985	3.1	129.9	2.9
Umbria	29 142 004	4.8	79.2	4.6	112 665	55.0	33.3	44.6	0	-	0	-	29 254 669	4.8	78.7	4.6
Marche	24 231 008	6.6	83.5	6.3	62 614	66.5	51.6	41.1	0	-	0	-	24 293 622	6.6	83.4	6.3
Lazio	57 249 600	4.7	107.0	4.4	180 483	66.2	105.9	46.5	80 552	46.9	11.1	41.4	57 510 635	4.6	105.7	4.4
Abruzzo	50 404 587	4.6	129.5	4.4	87 051	71.8	77.5	42.2	1 193	100.0	1.0	-	50 492 831	4.6	129.0	4.4
Molise	14 523 394	9.0	110.5	8.5	106 992	93.1	120.0	54.1	5 598	100.0	22.4	-	14 635 984	8.9	110.4	8.4
Campania	42 353 904	6.0	111.5	5.7	112 595	48.3	97.4	42.7	36 194	59.6	11.2	49.3	42 502 693	6.0	110.6	5.7
Puglia	12 046 337	11.0	84.2	10.5	108 303	95.5	123.5	66.4	5 844	100.0	3.0	-	12 160 485	10.9	83.4	10.4
Basilicata	27 415 389	6.9	106.3	6.5	230 731	46.0	123.8	11.1	15 086	73.0	4.6	64.8	27 661 206	6.9	105.1	6.4
Calabria	86 990 394	4.7	190.0	4.3	706 558	54.8	267.7	39.2	270 501	57.0	35.5	53.1	87 967 454	4.7	187.9	4.3
Sicilia	23 125 002	6.7	91.2	6.1	56 190	95.7	49.4	76.4	1 605	86.9	1.1	70.6	23 182 797	6.7	90.5	6.1
Sardegna	31 286 179	5.8	57.1	5.4	1 543 109	22.2	60.4	20.0	53 445	56.1	5.6	53.1	32 882 733	5.6	56.4	5.2
Italia	1 256 099 493	1.1	146.4	1.0	12 246 493	8.7	100.2	7.4	1 070 512	21.4	19.8	20.0	1 269 416 499	1.1	144.9	1.0

Number of satellites Purpose
3135 Communications
1030 Earth Observation
385 Technology development/demonstration

154 Navigation/positioning

22 Earth science

18 Other purposes



## EU budget: A €16 billion Space Programme to boost EU space leadership beyond 2020

Space sector employs over 231,000 people

Its value is estimated at € 53-62 billion in 2017, 2nd largest in the world



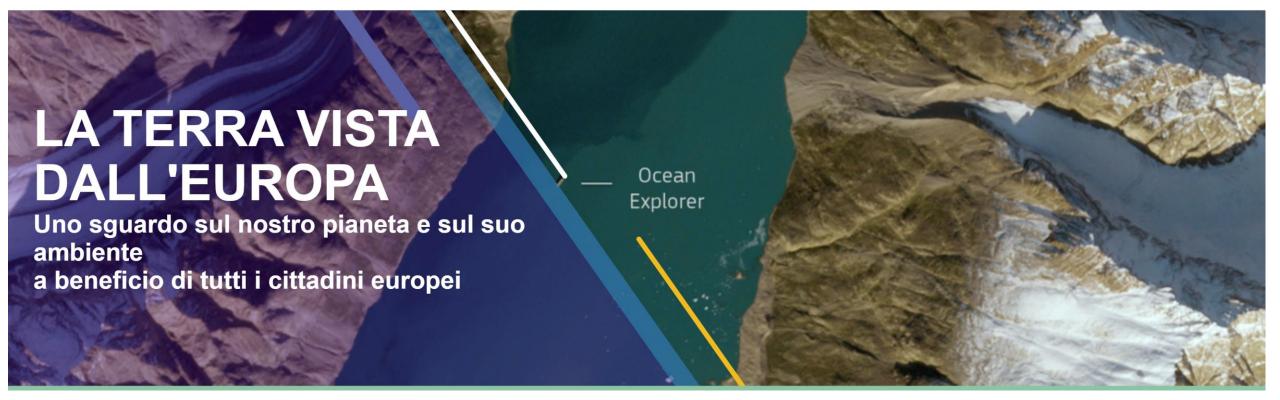
A third of the world's satellites are made in Europe.







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#### Remote sensing support for national forest inventories

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Received 12 August 200 Scandinavian Journal of Forest Research, 2010; 25: 368-38

#### REVIEW ARTICLE

National forest inventory programs are task variety of users and applications. Time, cost, surement and estimation efficiencies and the derived products. Many of the recent innovation using remotely sensed data in lieu of field ob with remotely sensed data obtained from lidar s
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produce and report timely and accurate National forest inventories (NFIs) have a lon DOI: 10.1080/07038992.2016.1207484 sources. The variables for which estim include, but are not limited to, forest area, Forest Resource Assessment: United N

USA, Tel.: +1 651 649 5174; fax: +1 651 6495285.

E-mail address: microberts@fs.fed.us (R.E. McRo)

Strategic national forest inventories

#### Advances and emerging issues in national forest inventories

RONALD E. McROBERTS1, ERKKI O. TOMPPO2 & ERIK NÆSSET

Northern Research Station, US Department of Agriculture Forest Service, 1992 Folwell Avenue, St Paul, Minnesot. MN 55108, USA, 2Finnish Forest Research Institute, Helsinki, Finland, and 3Norwegian University of Life Sciences, As,

the twentieth century. Recent issues such as increased demand for forest data, internation have led to greater inventory efficiencies, bet

2.4 billion hectares of forest, more th Abstract. Forest inventory and mana forested area of the earth (Tomppo et history, advances and emerging issues for international forest inventory communit possible task for a relatively short journa a result, this review focuses on only sel-(1) a brief historical review with emphas issues that have shaped current app implementing NFIs; (2) a summary of orest inventory practices in the next deca structural features of NFIs, albeit w diversity of operational implementations review of international reporting require NFI data with emphasis on approaches

estimation methods that can be enh Correspondence: R. E. McRoberts, Northern Research 5 MN 55108, USA. E-mail: rmcroberts@fs.fed.us

nized estimation; (4) an overview of

(Received 12 March 2010; accepted 24 May 2010) ISSN 0282-7581 print/ISSN 1651-1891 online © 2010 ° DOI: 10.1080/02827581.2010.496739

#### include, but are not limited to, forest area, growth, metallis, removals, trends, and metal-cuts, misred to the sum of more constructions are reported for these variables for categories, convenibles, solviculum and unique to the construction of t

species, ownerships, silvicultural and cut political units such summipratities, ownerships, silvicultural and cut political units such summipratities, or summiprations, counting states. Users of inventory data are many, in planners and managers, forest industry de summipration and industrial states. White states are summitted in the summitted and states are summitted in the summitted and states are summitted in the summitted and states are summitted in the summitted in

Canadian Forest Service (Pacific Forestry Centre), Natural Resources Canada, 506 West Burnsid ments (e.g. United Nations Food and Agric Keywords: estimation, harmonization, inter Road Victoria RC V87 IM5 Canada

> <sup>2</sup>Faculty of Forestry, University of British Columbia, 2424 Main Mall, Vancouver, BC, V6T 1Z4, Canada Department of Forest Sciences, University of Helsinki, FI-00014 Helsinki, Finland

<sup>4</sup>College of Forestry, Oregon State University, Corvallis, OR 97331, USA

economic, environmental, and social polic these escalating information needs and I broader range of information needs. This

which we posit as having the greatest potparticular, has proven to be a transforma

d'objectifs de politique économique, env avancées fournissent des données pour contient des articles qui utilisent une va orestier ou liées à l'inventaire forestier.

s estimons comme ayant le plus grand ressources forestières pour la planifica canning (ALS)», le balavage laser terres et un large éventail de types de forêts. I

Sustainable forest management is a bai the demands of an ever increasing human p

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Corresponding author e-mail: joanne.white@car

Remote Sensing of Environmen

Modelling lidar-derived estimates of forest attributes over space and time: A review of approaches and future trends

Contents lists available at ScienceDire

Nicholas C. Coops \*,\*, Piotr Tompalski \*, Tristan R.H. Goodbody \*, Martin Queinnec \*, Joan E. Luther b, Douglas K. Bolton c, Joanne C. White d, Michael A. Wulder d, Oliver R. van Lier e, Txomin Hermosilla

Light detection and ranging (lidar) data acquired from airborne or spaceborne platforms have revolut flight lines to provide complete coverage of an area of interest, or using transacts to sample a given population

proaches to extend air- or space-based lidar data with the aim of communicating methods, outcomes, and ac puracies, and offering guidance on linking lidar metrics and lidar-derived forest attributes with broad-are uracies, and offetting guidance on linking lidar metrics and lidar-detived forest attributes with broad-are predictors. Modelling approaches are developed for a variety of applications. In some cases, specaration or parially-cahaustric layers may be useful for forest management purpose, driving management and inventor selections over malled rectur areas or regions. In other cases, outputs are designed for monitoring at regional plobal scales, and may be - due to the spatial grain of the structural estimates - insufficiently accurate or reliable r management. From the reviewed studies, we found height, aboveground biomass and volume, derived from either upper proportions of a large-footprint full-waveform lidar profiles, or statistically modelled from discre eturn small-footprint lidar point clouds, to be the most commonly extended forest attributes, followed by canor cover, basil area and stand complexity. Assessment of the accuracy and bias of the extrapolated forest attributes, waited with both independent and model-derived estimate. The coefficient of determination (F.) was the most often reported, followed by absolute and relative (i.e., as a proportion of the mean) root mean square error (GMSE and ADSER's respectively). Compilation of the standa occuration regarded that the variance explained it predictions of forest beight ranged from R<sup>2</sup> = 0.38 to 0.90 (mean = 0.44), RMSE from 2 to 6m and RMSEN from 12 to 34%. For volume, R2 ranged from 0.25 to 0.72 (mean = 0.53) and RMSE from 60 to 87 m2/ha and fo poveground blomass (AGR) R2 ranged from 0.35 to 0.78 (mean = 0.55) and RMSE from 28 to 44 Mg/ha. Then

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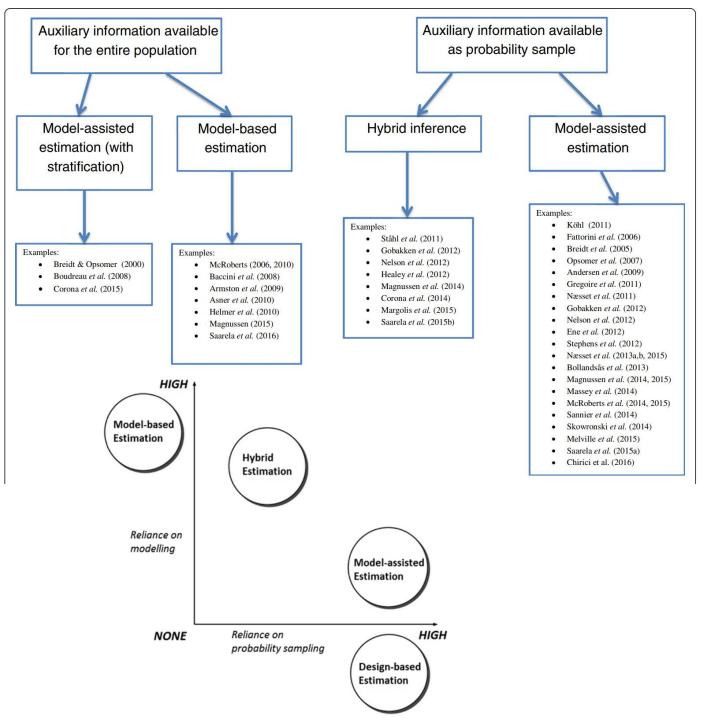


#### In NFIs Remote sensing contributed:

- To the production of more timely, cost efficient, and precise traditional inventory estimates
- To derive **new spatial products** (maps, small area estimates)

**Technologies** that are now on the horizon have the potential to alter radically the ways in which trees are measured, estimates are produced, and products are delivered.

The lack of standardized, spatially exhaustive open access datasets, as well as community consensus on methods and best practices limits the broader uptake and operationalization of these approaches



Ståhl et al. Forest Ecosystems (2016) 3:5 DOI 10.1186/s40663-016-0064-9



#### DISCUSSION

Open Access

## Use of models in large-area forest surveys: comparing model-assisted, model-based and hybrid estimation



Göran Ståhl<sup>1</sup>, Svetlana Saarela<sup>1\*</sup>, Sebastian Schnell<sup>1</sup>, Sören Holm<sup>1</sup>, Johannes Breidenbach<sup>2</sup>, Sean P. Healey<sup>3</sup>, Paul L. Patterson<sup>3</sup>, Steen Magnussen<sup>4</sup>, Erik Næsset<sup>5</sup>, Ronald E. McRoberts<sup>3</sup> and Timothy G. Gregoire<sup>6</sup>

#### Abstract

This paper focuses on the use of models for increasing the precision of estimators in large-area forest surveys. It is motivated by the increasing availability of remotely sensed data, which facilitates the development of models predicting the variables of interest in forest surveys. We present, review and compare three different estimation frameworks where models play a core role: model-assisted, model-based, and hybrid estimation. The first two are well known, whereas the third has only recently been introduced in forest surveys. Hybrid inference mixes design-based and model-based inference, since it relies on a probability sample of auxiliary data and a model predicting the target variable from the auxiliary data. We review studies on large-area forest surveys based on model-assisted, model-based, and hybrid estimation, and discuss advantages and disadvantages of the approaches. We conclude that no general recommendations can be made about whether model-assisted, model-based, or hybrid estimation should be preferred. The choice depends on the objective of the survey and the possibilities to acquire appropriate field and remotely sensed data. We also conclude that modelling approaches can only be successfully applied for estimating target variables such as growing stock volume or biomass, which are adequately related to commonly available remotely sensed data, and thus purely field based surveys remain important for several important forest parameters.

**Keywords:** Design-based inference, Model-assisted estimation, Model-based inference, Hybrid inference, National forest inventory, Remote sensing, Sampling

#### Introduction

Use of models in large-area surveys of forests is attracting increased interest. The reason is the improved availability of auxiliary data from various remote sensing platforms. Aerial photographs (e.g., Næsset 2002a, Bohlin et al. 2012) and optical satellite data (e.g., Reese et al. 2002) have been available and used operationally for many decades, while data from profiling (e.g., Nelson et al. 1984, Nelson et al. 1988) and scanning lasers (e.g., Næsset 1997) and radars (Solberg et al. 2010) have become available for practical applications more recently. Some of the new types of remotely sensed data, such as data from laser scanners, have already become widely applied in forest inventories (e.g., Næsset 2002b). A common application involves

the development of models that are applied wall-to-wall over an area of interest (e.g., Næsset 2004), often for providing data for forest management. However, this type of data is increasingly applied also in connection with large-area forest surveys, such as national-level forest inventories (Tomppo et al. 2010, Asner et al. 2012).

Applications of models in large-area forest surveys often use the model-assisted estimation framework (Särndal et al. 1992) where a model is used to support the estimation following probability sampling within the context of design-based inference (Gregoire 1998). Importantly, an inadequately specified model will not make the estimators biased in this case, but only affect the variance of the estimators. Examples of large-area forest inventory applications include Andersen et al. (2011) who applied the technique in Alaska, Gregoire et al. (2011) and Gobakken et al. (2012), who applied it in Hedmark County, Norway, and Saarela et al. (2015a) who used it in Kuortane, Finland.

Full list of author information is available at the end of the article



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#### Apply models to entire Stand-level management area using inventory wall-to-wall metrics Summarize grid-cell estimates to stand-level Grid Cells Generate wall-to-wall ALS metrics Point Cloud ALS metrics V = 31.2 + 1.23p90 - 4.4p200 Ground plot measurements 0 and point cloud

White et al., 2013. The Utility of Image-Based Point Clouds for Forest Inventory: A Comparison with Airborne Laser Scanning. https://doi.org/10.3390/f4030518

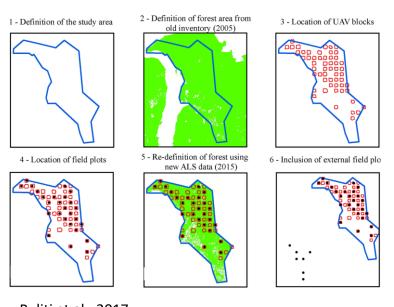
# Municipalities with NFI points Municipalities with NFI points Initialisation Iteration 1 Iteration 2 Small area Partitioning tree (SAETree)

Vega et al., 2021 A new small area estimation algorithm to balance between statistical precision and scale https://doi.org/10.1016/j.jag.2021.102303

#### Data assimilation/data integration

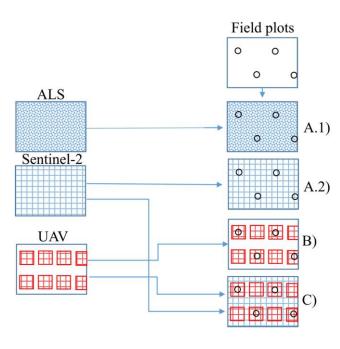
How to integrate multiple input data from multiple sources? We just have to choice!

#### UAV->LiDAR->SATELLITE->NFI



Puliti et al., 2017
Use of partial-coverage UAV data in sampling for large scale forest inventories.

https://doi.org/10.1016/j.rse.2017.03.019



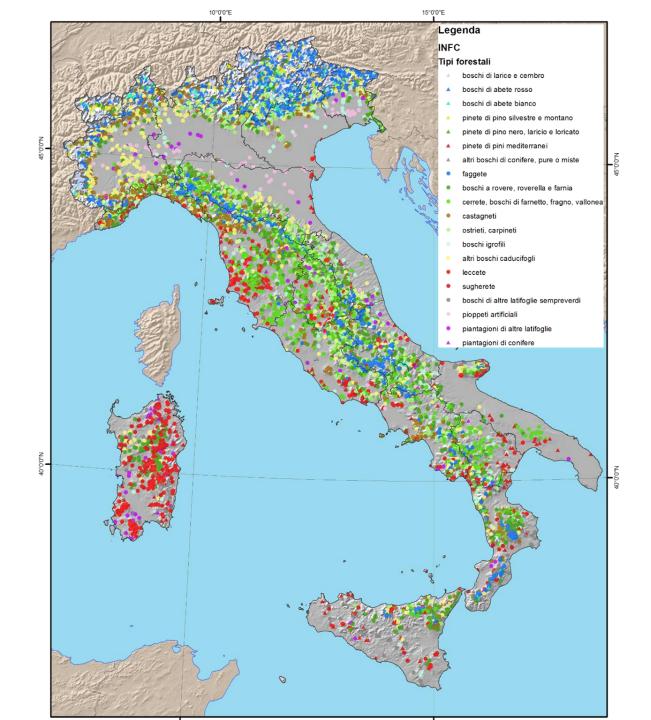
Puliti S, Saarela S, et al., 2018 Combining UAV and Sentinel-2 auxiliary data for forest growing stock volume estimation through hierarchical modelbased inference https://doi.org/10.1016/j.rse.2017.10.007



## I risultati del progetto AGRIDIGIT

dalla ricerca all'applicazione

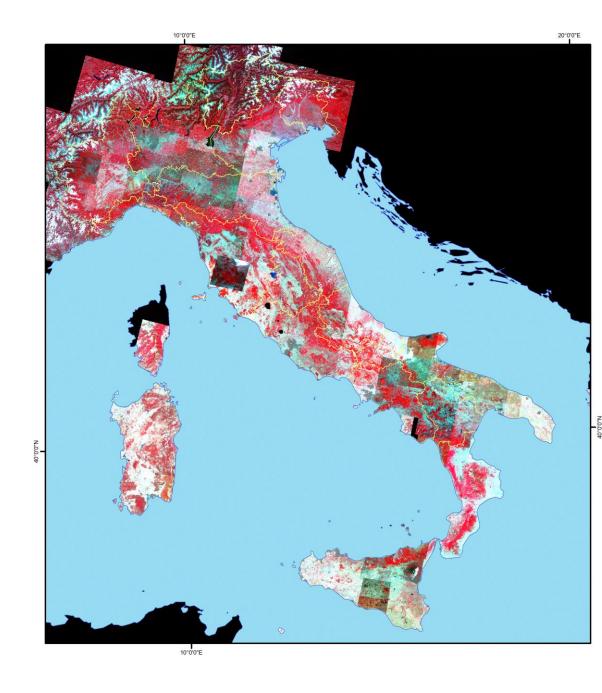
stima spazialmente esplicita di variabili forestali



## Il problema dei dati di input



Con algoritmi di intelligenza artificiale è possibile spazializzare con immagini telerilevate e altri strati informativi geografici il dato osservato nelle aree di saggio a terra INFC



Esempio di applicazione sul dato INFC2005 **National application** LOO  $R^2 = 0.61$  and RMSE  $\cong 52$  m<sup>3</sup>ha<sup>-1</sup> Geographically weighted regression Int J Appl Earth Obs Geoinformation Wall-to-wall spatial prediction of growing stock volume based on Italian National Forest Inventory plots and remotely sensed data Gherardo Chirici<sup>a</sup>, Francesca Giannetti<sup>a</sup>, Ronald E. McRoberts<sup>b,c</sup>, Davide Travaglini<sup>a</sup>, Matteo Pecchi<sup>a</sup>, Fabio Maselli<sup>d</sup>, Marta Chiesi<sup>d</sup>, Piermaria Corona<sup>e</sup> Random Forests (R2=0.61 - RMSE= 52.1 m3ha-1 RMSE<sub>66</sub>=34.2) Legend **Growing Stock Volume** accurate results with a pixel level R<sup>2</sup> = 0.69 and RMSE<sub>8</sub>, = 37.2% against the independent validation datase Model-assisted estimates were more precise than the original design-based estimates provided by the NFI. High: 350 Usually, in the context or international and national programs, that per of data is collected unless apund-bear Mational Portes international (NFs) that are designed to provide aggregated estimates of portantees such as forest area, proving such volume, binansa, increments are national and regional levels (Usundake et al., 2014; Kangari, et al., 2017). The aggregated national event of the context of the cont Forest casts are essential for mitugles purposes including and assessing forest resource distribution (e.g. Kyoto protocol) (Coronator et al., 2011; FAO, 2010), monitoring biodiversity (Chirel et al., 2012; FOREST EUROPE, 2015), improving restoration programs (FAO and UNCCD, 2015), mitugles and application of the control of the coronatory of the corona as large sub-national regions. In these traditional NFIs, remote sensing is used for purposes such as initial stratification of sampling units ntiper/astorg/10/1016/j/kg.2019.101959

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Artic

### Wall-to-Wall Mapping of Forest Biomass and Wood Volume Increment in Italy

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- Correspondence: elia.vangi@unifi.it

Abstract: Several political initiatives aim to achieve net-zero emissions by the middle of the twentyfirst century. In this context, forests are crucial as a carbon sink to store unavoidable emissions. Assessing the carbon sequestration potential of forest ecosystems is pivotal to the availability of accurate forest variable estimates for supporting international reporting and appropriate forest management strategies. Spatially explicit estimates are even more important for Mediterranean countries such as Italy, where the capacity of forests to act as sinks is decreasing due to climate change. This study aimed to develop a spatial approach to obtain high-resolution maps of Italian forest above-ground biomass (ITA-BIO) and current annual volume increment (ITA-CAI), based on remotely sensed and meteorological data. The ITA-BIO estimates were compared with those obtained with two available biomass maps developed in the framework of two international projects (i.e., the Joint Research Center and the European Space Agency biomass maps, namely, JRC-BIO and ESA-BIO). The estimates from ITA-BIO, JRC-BIO, ESA-BIO, and ITA-CAI were compared with the 2nd Italian NFI (INFC) official estimates at regional level (NUT2). The estimates from ITA-BIO are in good agreement with the INFC estimates ( $R^2 = 0.95$ , mean difference = 3.8 t ha-1), while for JRC-BIO and ESA-BIO, the estimates show R2 of 0.90 and 0.70, respectively, and mean differences of 13.5 and of 21.8 t ha-1 with respect to the INFC estimates. ITA-CAI estimates are also in good agreement with the INFC estimates ( $R^2$ =0.93), even if they tend to be slightly biased. The produced maps are hosted on a web-based forest resources management Decision Support System developed under the project AGRIDIGIT (ForestView) and represent a key element in supporting the new Green Deal in Italy, the European Forest Strategy 2030 and the Italian Forest

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Chiesi, M.; D'Amico, G.; Puletti, N.

Increment in Italy. Forests 2022, 13,

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Biomass and Wood Volume

1989. https://doi.org/10.3390/

Received: 30 September 2022

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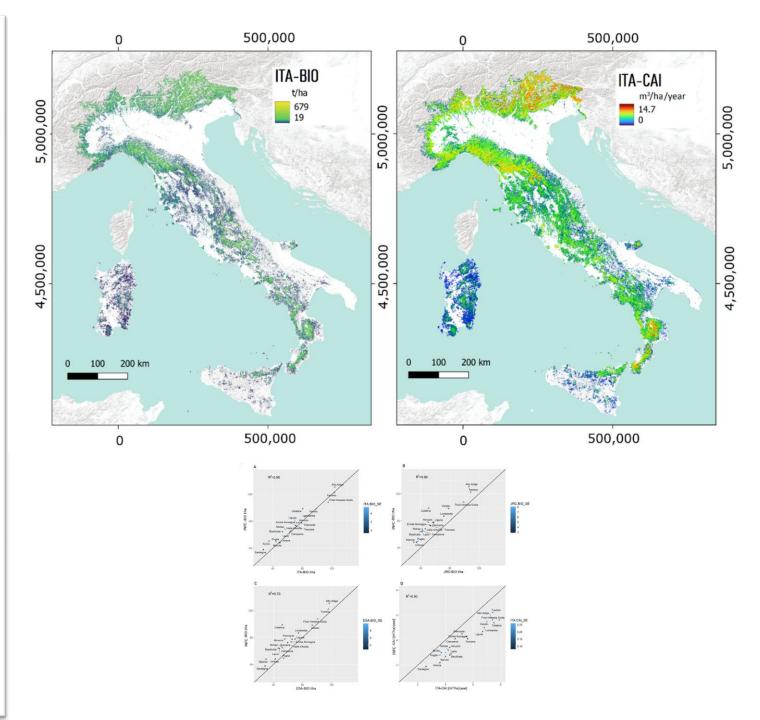
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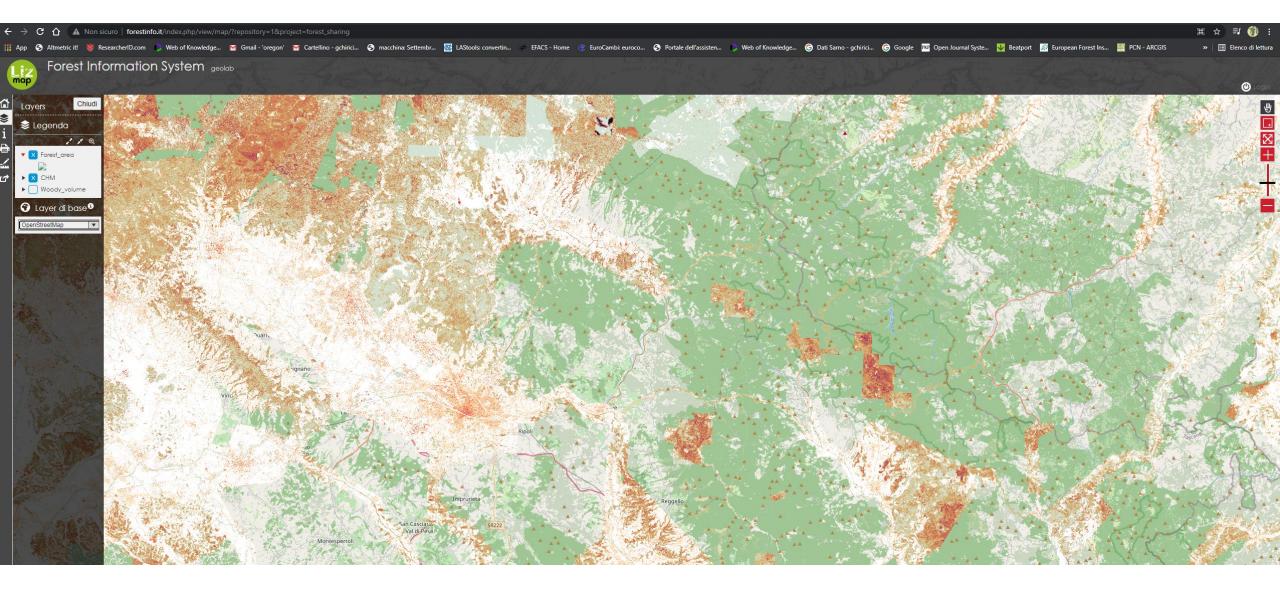
Copyright: © 2022 by the authors. Licensee MDPI, Basel, Switzerland. This article is an open access article distributed under the terms and conditions of the Creative Commons Attribution (CC BY) license (https://creative.commons.org/license s/by/4.0/). Keywords: forest biomass; National Forest Inventories; remote sensing; Mediterranean forest; forest increment

#### 1 Introduction

Measuring the amount of CO<sub>2</sub> stocked in forest ecosystems is mandatory to support the new European (EU) Forest Strategy for 2030, a flagship initiative of the European Green Deal, in sight of achieving neutrality with respect to greenhouse gas emission in



Le mappe derivanti dalla spazializzazione possono essere facilmente condivise on line Senza dover condividere la posizione geografica precisa delle aree di saggio



Contents lists available at ScienceDirect

#### Remote Sensing of Environment

journal homepage: www.elsevier.com/locate/rse



Large-area mapping of Canadian boreal forest cover, height, biomass and other structural attributes using Landsat composites and lidar plots



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#### ARTICLE INFO

Keywords: Lidar Landsat Forest structure Monitoring Imputation Random Fores

#### ABSTRACT

Passive optical remotely sensed images such as those from the L spatially comprehensive, well-calibrated reflectance measures that as an alternative to field plot data, the use of Light Detection and Ra validation purposes in combination with such satellite reflectance response variables has become well established. In this research, we forest structural attributes over the ~552 million ha boreal fores dependent validation we utilize airborne lidar-derived measuremen plots) obtained in 2010 via a > 25,000 km transect-based national lidar plot structural variables to wall-to-wall 30-m spatial resolution

ree Crown Closure in Canada



Proportion of area covered by trees



DOI: 10.1080/07038992.2016.1207484

#### **Remote Sensing Technologies for Enhancing Forest Inventories: A Review**

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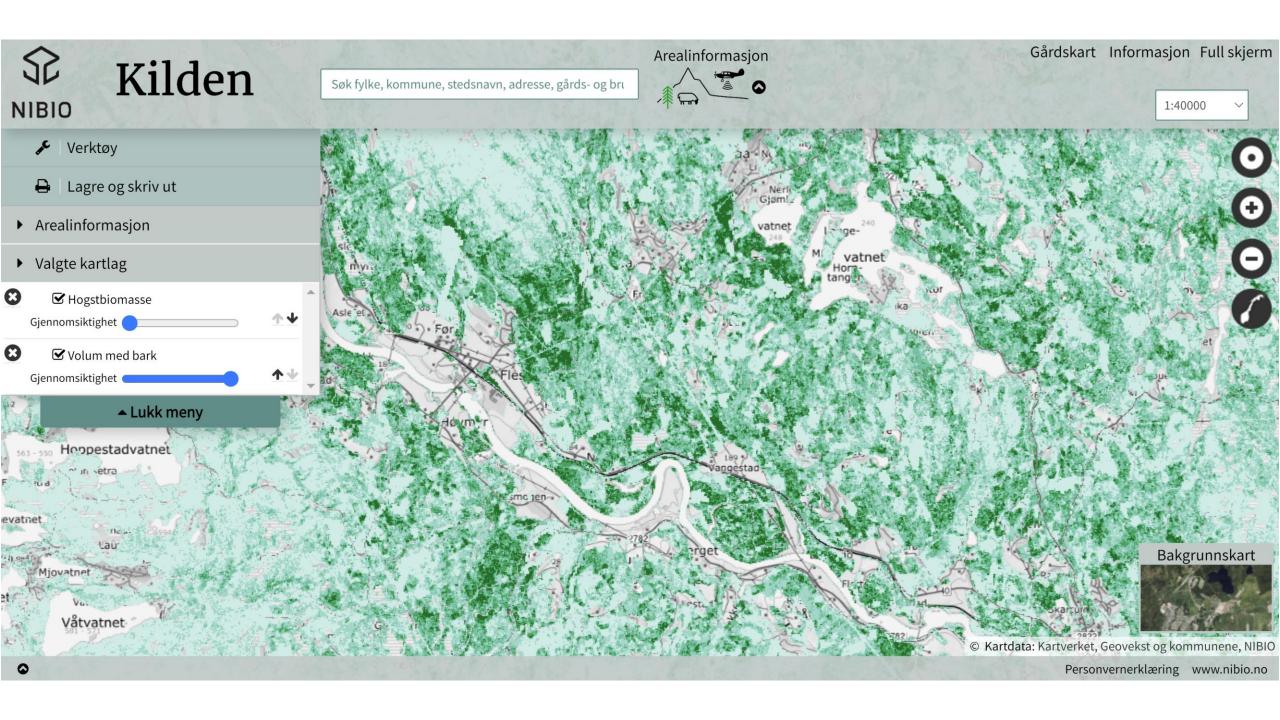
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Abstract. Forest inventory and management requirements are changing rapidly in the context of an increasingly complex set of economic, environmental, and social policy objectives. Advanced remote sensing technologies provide data to assist in addressing these escalating information needs and to support the subsequent development and parameterization of models for an even broader range of information needs. This special issue contains papers that use a variety of remote sensing technologies to derive forest inventory or inventory-related information. Herein, we review the potential of 4 advanced remote sensing technologies, which we need as having the greatest notantial to influence forest inventories decianed to characterize forest resource information

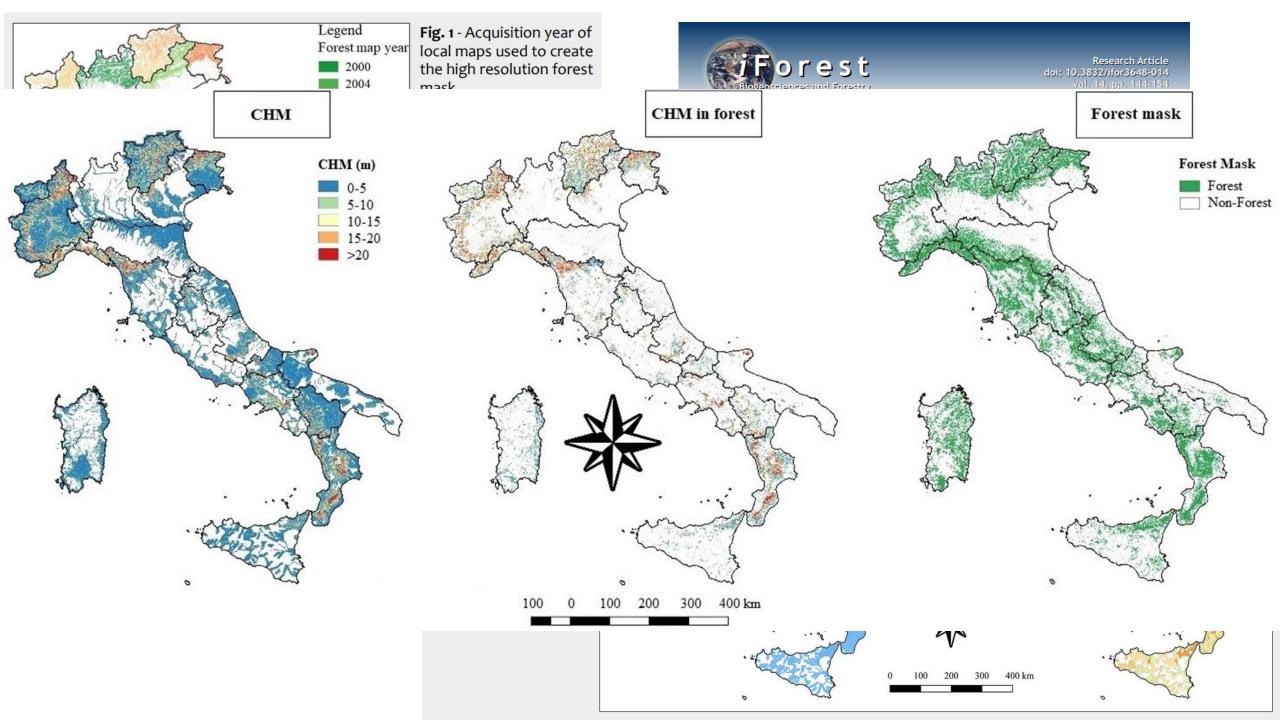
b Canadian Forest Service (Pacific Forestry Centre), Natural Resources Canada, 506 West Burnside Road, Victoria, BC, V8Z 1M5, Canada

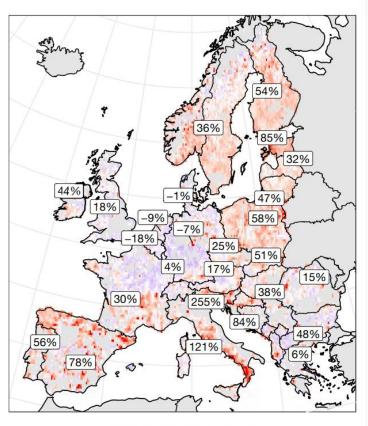
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Change in harvested forest area 2016–2018 versus 2004–2015 (%)



#### **Article**

## Abrupt increase in harvested forest area over Europe after 2015

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Check for updates

Guido Ceccherini<sup>1⊠</sup>, Gregory Duveiller<sup>1</sup>, Giacomo Grassi<sup>1</sup>, Guido Lemoine<sup>2</sup>, Valerio Avitabile<sup>1</sup>, Roberto Pilli<sup>1</sup> & Alessandro Cescatti<sup>1</sup>

Forests provide a series of ecosystem services that are crucial to our society. In the European Union (EU), forests account for approximately 38% of the total land surface<sup>1</sup>. These forests are important carbon sinks, and their conservation efforts are vital for the EU's vision of achieving climate neutrality by 2050<sup>2</sup>. However, the increasing demand for forest services and products, driven by the bioeconomy, poses challenges for sustainable forest management. Here we use fine-scale satellite data to observe an increase in the harvested forest area (49 per cent) and an increase in biomass loss (69 per cent) over Europe for the period of 2016–2018 relative to 2011–2015, with large losses occurring on the Iberian Peninsula and in the Nordic and Baltic countries. Satellite imagery further reveals that the average patch size of harvested area increased by 34 per cent across Europe, with potential effects on biodiversity, soil erosion and water regulation. The increase in the rate of forest harvest is the result of the recent expansion of wood markets, as suggested by econometric indicators on forestry, wood-based bioenergy and international trade. If such a high rate of forest harvest continues, the post-2020 EU vision of forest-based climate mitigation may be hampered, and the additional carbon losses from forests would require extra emission reductions in other sectors in order to reach climate neutrality by 2050<sup>3</sup>.

#### **Matters arising**

## Quantifying forest change in the European Union

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Iddo K. Wernick<sup>™</sup>, Philippe Ciais², Jonas Fridman³, Peter Högberg⁴, Kari T. Korhonen⁵, Annika Nordin⁴ & Pekka E. Kauppi⁴.6

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Published online: 28 April 2021

Check for updates

Check for updates

Breidenbach et al. Annals of Forest Science https://doi.org/10.1186/s13595-022-01120-4

(2022) 79:2





#### OPINION PAPER

**Open Access** 

Harvested area did not increase abruptly—how advancements in satellite-based mapping led to erroneous conclusions



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#### **Abstract**

**Key message:** Using satellite-based maps, Ceccherini et al. (Nature 583:72-77, 2020) report abruptly increasing harvested area estimates in several EU countries beginning in 2015. Using more than 120,000 National Forest Inventory observations to analyze the satellite-based map, we show that it is not harvested area but the map's ability to detect harvested areas that abruptly increases after 2015 in Finland and Sweden.

Keywords: Global Forest Watch, Landsat, Remote sensing, National Forest Inventory, Greenhouse Gas Inventory

#### **Matters arising**

## Concerns about reported harvests in European forests

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ARISING FROM G. Ceccherini et al. Nature https://doi.org/10.1038/s41586-020-2438-y (2020)

## Conclusioni





- Integrazione INFC e Sistema Informativo Forestale Nazionale -> produzione di cartografie con telerilevamento
- Mappe UFFICIALI a 23 m di risoluzione da INFC2005 e INFC2015 saranno rilasciate a breve (collaborazione CUFA-AISF)
- Passaggio al nuovo programma permanente IFNI2025
- Maggiore collaborazione con enti di ricerca e accesso ai dati grezzi
- E' necessario il completamento delle informazioni di base: LiDAR e carta forestale

