

# **Theoretical and practical aspects of using satellite observations for crop area estimation**

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# Problem statement: Area estimation

- Maps imperfect
  - **Wall-to-wall** coverage: information on the population (though inaccurate)
  - Omission and commission **errors**
  - Direct use: a ***pixel counting estimator*** is biased [*Gallego, 2004; Gallego et al., 2010; McRoberts, 2011*]
- Statistical frameworks
  - **Sampling-based** approach and a **design-based inference** framework [*Olofsson et al., 2014*]
    - Maps used for strata
    - Map **inaccuracies will not impact the bias**: depends on the sampling design and estimator
    - Map **inaccuracies will impact the efficiency of stratification**
    - **Model-assisted** estimation to further improve the precision of the estimator [*Carfagna & Gallego, 2005; Gallego et al., 2010; McRoberts et al., 2024*]
  - Alternative to the design-based inference framework
    - **Model-based** inference [*Ståhl et al., 2016; McRoberts et al., 2022*]
    - Requires specification of an adequate model that will impact the bias and precision of the estimator

# Efficiency of stratification

- **Assumptions**

- Sampling design: **stratified random sampling**
- Map's **classes** used as **strata**
- Two maps: *A* and *B* used for stratification when estimating an area proportion  $\hat{y}$

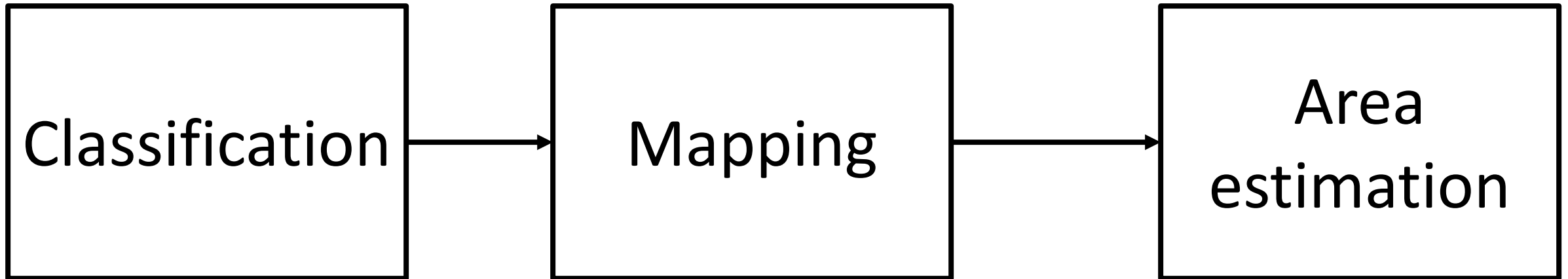
- **Efficiency** of stratification [*Gallego, 2007*]

$$\eta_{A,B} = \frac{\text{var}(\hat{y}_B) \times n_B}{\text{var}(\hat{y}_A) \times n_A}$$

$n_B$  and  $n_A$  are sample size, and  $\text{var}(\hat{y}_B)$  and  $\text{var}(\hat{y}_A)$  are estimated variances of  $\hat{y}$

- **Hypothesis:** a more accurate map will be more efficient
- **Question:** how much?

# Land cover / land use mapping and area estimation



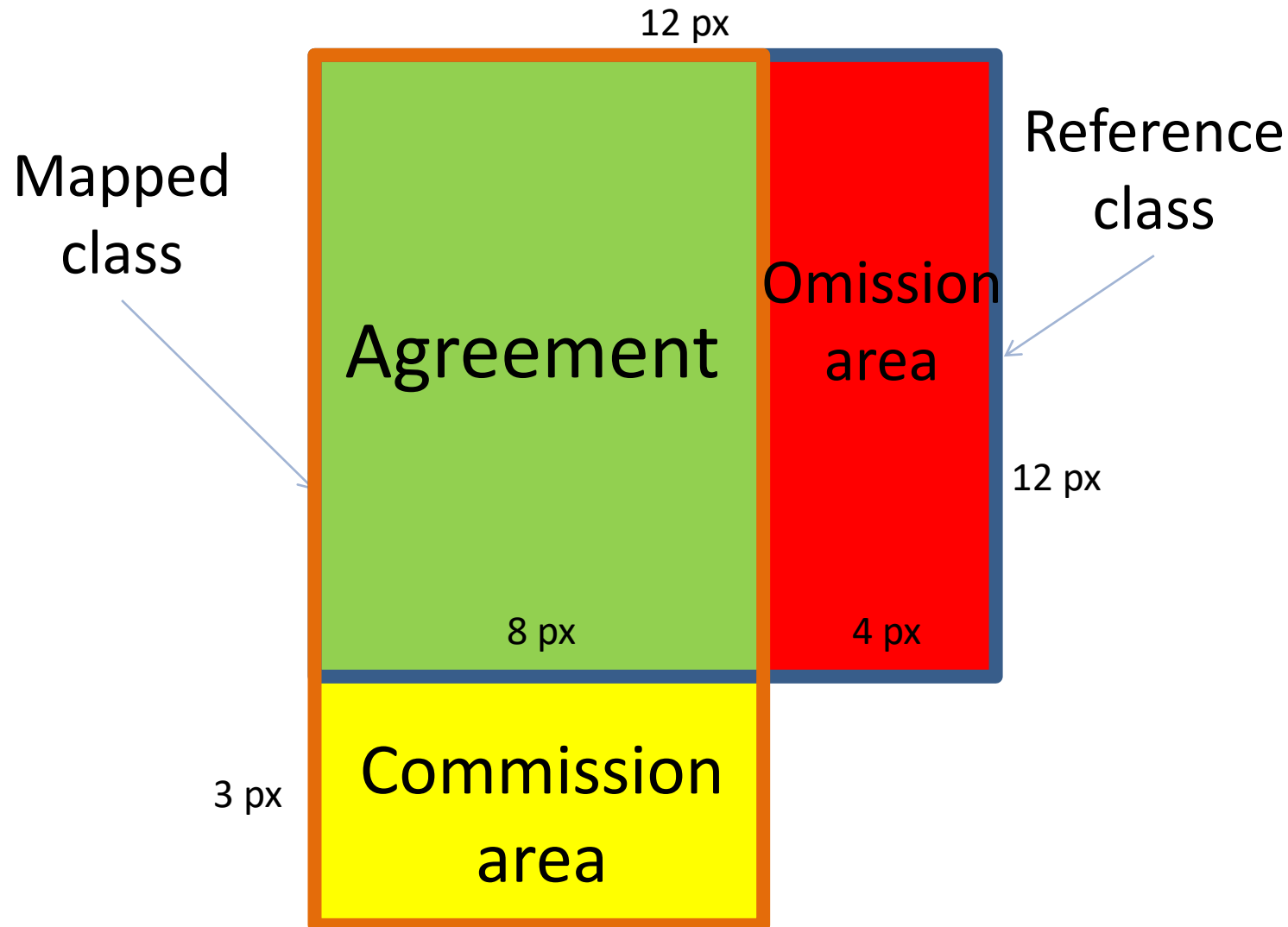
- Data
- Features
- Algorithms

- Spatial context

- Unbiased estimators with uncertainties

# Land cover / land use mapping and area estimation

- Pixel counting is a biased estimator



Reference area:

$$12 \times 12 = 144 \text{ px}$$

Mapped area:

$$8 \times 15 = 120 \text{ px (bias -17\%)}$$

$$PA = 8 \times 12 / (12 \times 12) = 66.7\%$$

$$UA = 8 \times 12 / (8 \times 15) = 80\%$$

$$\underline{\text{Rel. bias} = PA/UA - 1}$$

# Land cover / land use mapping and area estimation

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## Remote Sensing of Environment

journal homepage: [www.elsevier.com/locate/rse](http://www.elsevier.com/locate/rse)



Review

## Good practices for estimating area and assessing accuracy of land change

Pontus Olofsson<sup>a,\*</sup>, Giles M. Foody<sup>b</sup>, Martin Herold<sup>c</sup>, Stephen V. Stehman<sup>d</sup>,  
Curtis E. Woodcock<sup>a</sup>, Michael A. Wulder<sup>e</sup>



# General steps

## ▪ Sampling design

- The sampling design is the protocol for selecting the **subset of spatial units** (e.g., pixels or polygons) that will form the **basis** of the **accuracy assessment**

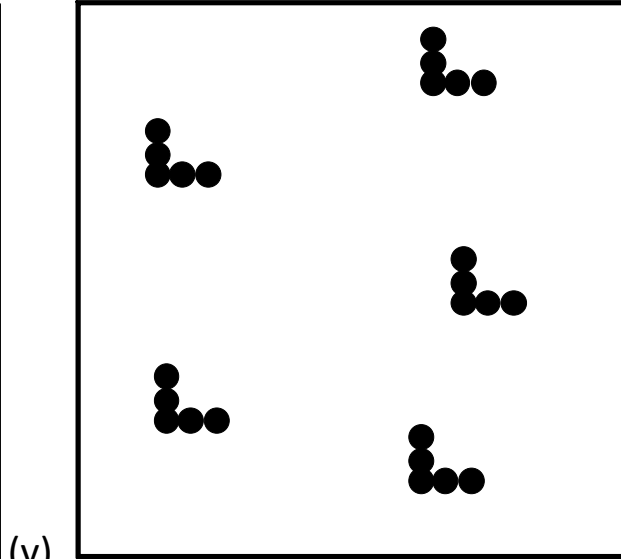
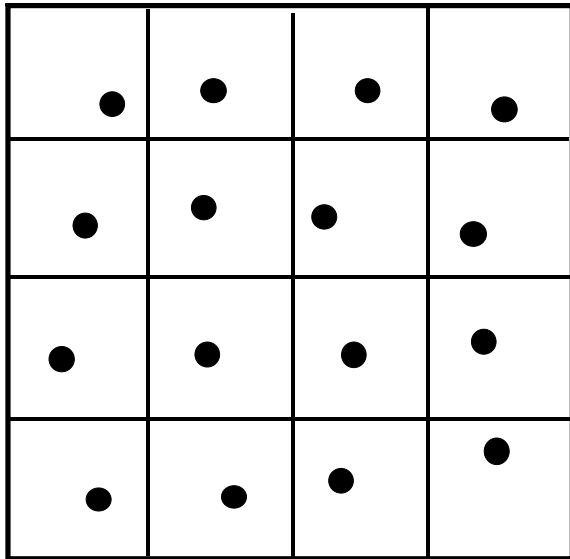
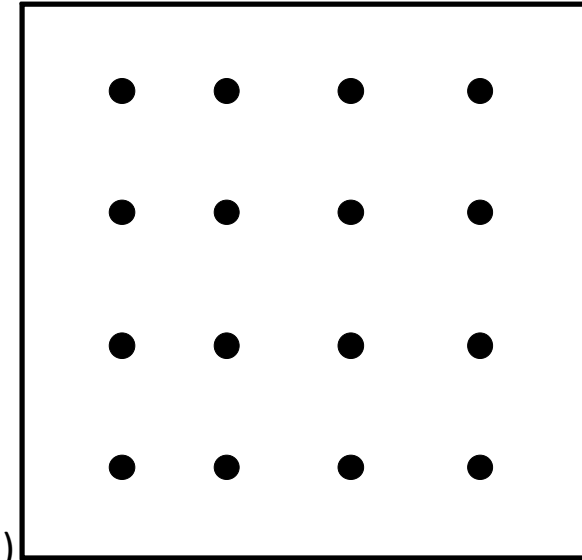
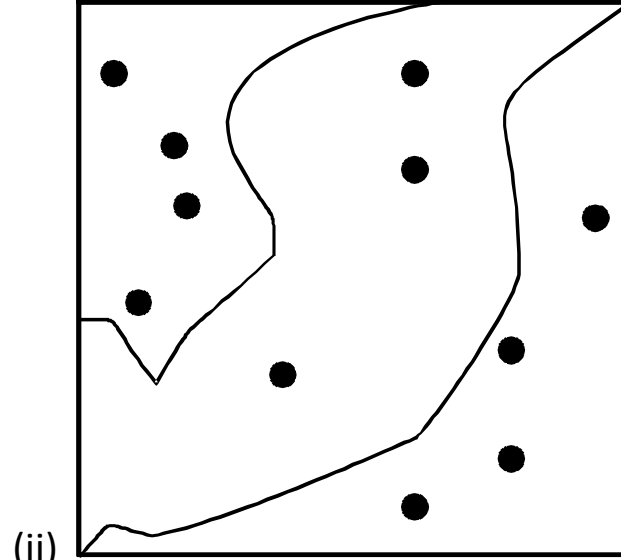
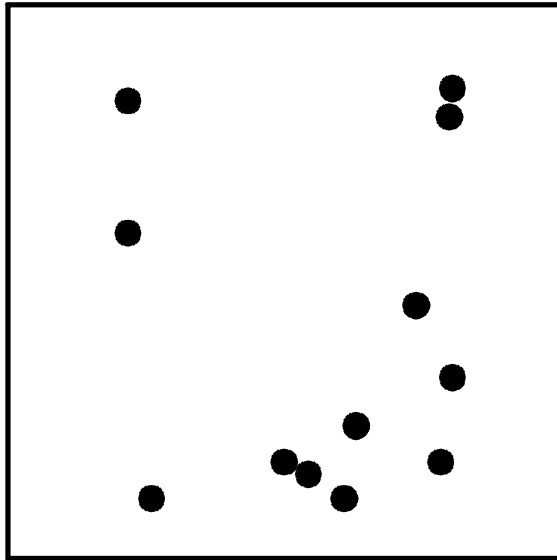
## ▪ Response design

- Encompasses all aspects of the protocol that lead to determining whether the **map** and **reference** classifications are **in agreement**

## ▪ Analysis

- Includes protocols for defining how to quantify **accuracy** along with the formulas and inference framework for estimating **accuracy** and area and quantifying **uncertainty** of these estimates

# Sampling design



- (i) Random
- (ii) Stratified random
- (iii) Systematic
- (iv) Unaligned systematic
- (v) Cluster



# Error matrix

- Population matrix (in terms of area proportions)

		Reference				
		Class 1	Class 2	...	Class K	Total
Map	Class 1	$f_{11}$	$f_{12}$	...	$f_{1K}$	$f_{1\cdot}$
	Class 2	$f_{21}$	$f_{22}$	...	$f_{2K}$	$f_{2\cdot}$
	...	...	...	...	...	...
	Class k	$f_{k1}$	$f_{k2}$	...	$f_{kK}$	$f_{k\cdot}$
	Total	$f_{\cdot 1}$	$f_{\cdot 2}$	...	$f_{\cdot K}$	1

$f_{k\cdot}$  is the **mapped** area proportion

$f_{\cdot k}$  is the **reference** area proportion

Producer's accuracy (PA):  $v_k = f_{kk} / f_{k\cdot}$

Omission error (OE):  $\psi_k = (1 - v_k)$

User's accuracy (UA):  $u_k = f_{kk} / f_{\cdot k}$

Commission error (CE):  $\phi_k = (1 - u_k)$

# Error matrix

- Population matrix (in terms of area proportions)

		Reference				
		Class 1	Class 2	...	Class K	Total
Map	Class 1	$f_{11}$	$f_{12}$	...	$f_{1K}$	$f_{1\cdot}$
	Class 2	$f_{21}$	$f_{22}$	...	$f_{2K}$	$f_{2\cdot}$
	...	...	...	...	...	...
	Class k	$f_{k1}$	$f_{k2}$	...	$f_{kK}$	$f_{k\cdot}$
	Total	$f_{\cdot 1}$	$f_{\cdot 2}$	...	$f_{\cdot K}$	1

$f_{k\cdot}$  is the **mapped** area proportion

$f_{\cdot k}$  is the **reference** area proportion

- A bias of pixel counting estimator

*Gallego et al. (2010)*: through OE/CE

$$b_k = f_{k\cdot} - f_{\cdot k} = \phi_k f_{k\cdot} - \psi_k f_{\cdot k}$$

this work: through PA/UA

$$b_k = f_{\cdot k} \underbrace{\left( \frac{v_k}{u_k} - 1 \right)}_{\text{Relative bias}}$$

# Analysis

- **Elements of the population error matrix should be estimated from sample!**
- Suppose sample-based estimator of  $p_{ij}$  is  $\hat{p}_{ij}$
- The error matrix should be reported in terms of estimated area proportions!
- For **equal probability sampling designs** (e.g., **simple random** and **systematic sampling**) and for **stratified random sampling in which the strata correspond to the map classes**,

$$\hat{p}_{ij} = W_i \frac{n_{ij}}{n_{i.}}$$

$n_{ij}$  – number of MMU mapped as class  $i$  but reference is class  $j$

$n_{i.}$  – number of MMU mapped as class  $i$

$W_i$  – the proportion of area mapped as class  $i$

# Relative efficiency

- Elements of the confusion matrix estimated from the sample
- Sample size [Olofsson et al., 2014; Cochran, 1977, Eq. 5.25]

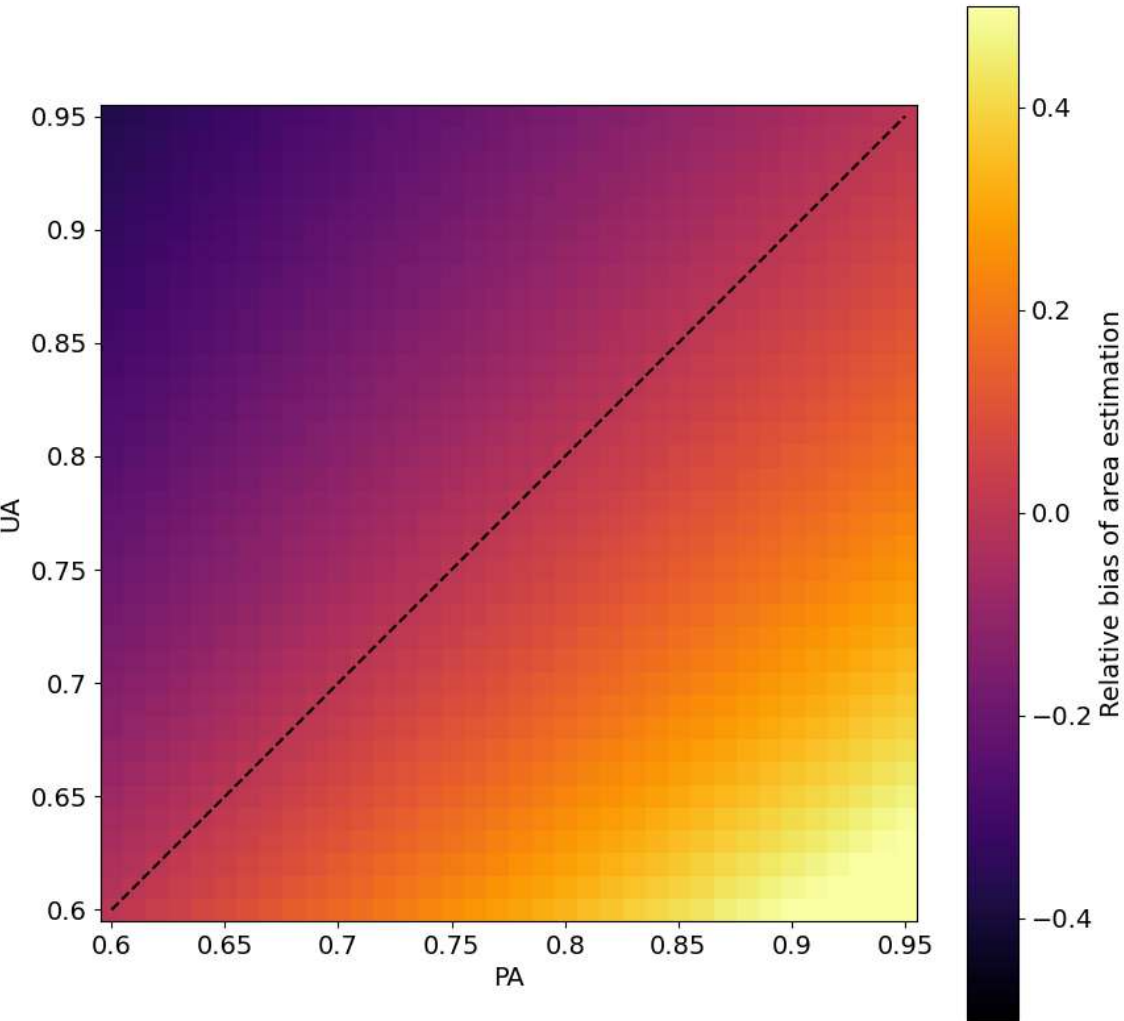
$$n = \frac{\left(\sum_{k=1}^K w_k s_k\right)^2}{V_{target} + 1/N \sum_{k=1}^K w_k s_k^2} \approx \frac{\left(\sum_{k=1}^K w_k s_k\right)^2}{V_{target}}$$

where  $V_{target}$  is the target variance,  $w_k \equiv f_k$  is the stratum weight,  $N$  is population size, and  $s_k$  is the standard deviation of stratum  $k$

(for area estimation  $s_i(\hat{f}_{\cdot k}) = \sqrt{\frac{n_{ik}}{n_{i\cdot}} \left(1 - \frac{n_{ik}}{n_{i\cdot}}\right)}$ )

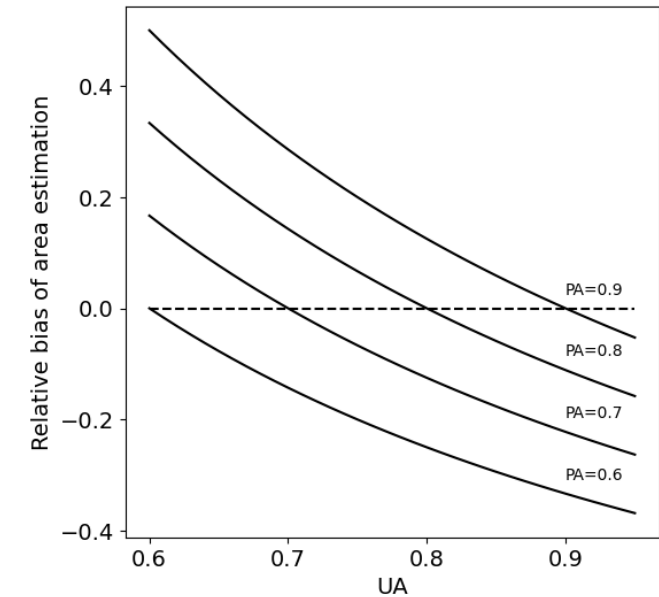
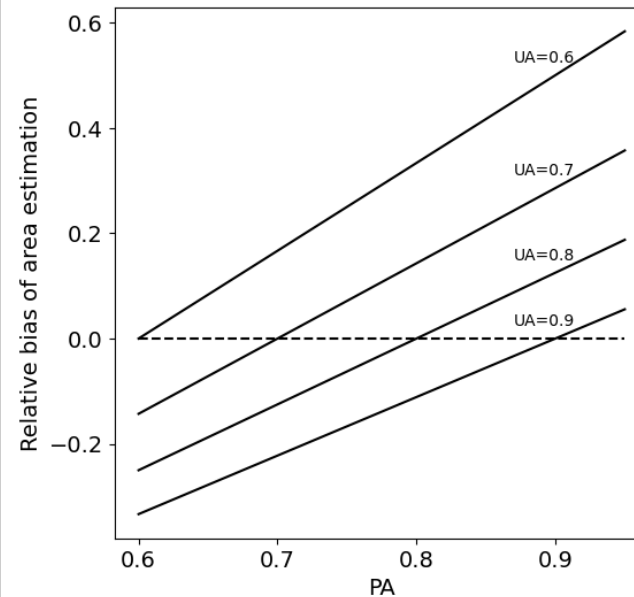
- Relative efficiency
 
$$\eta_{A,B} = \frac{\text{var}(\hat{y}_B) \times n_B}{\text{var}(\hat{y}_A) \times n_A} = \frac{\left(\sum_{i=1}^K f_{\cdot i} \frac{v_i^B}{u_i^B} s_i^B(\hat{f}_{\cdot k})\right)^2}{\left(\sum_{i=1}^K f_{\cdot i} \frac{v_i^A}{u_i^A} s_i^A(\hat{f}_{\cdot k})\right)^2}$$

# Pixel counting area estimator



The relative bias of area estimation with pixel counting depending on PA and UA

$$\text{Relative bias} = \left( \frac{PA}{UA} - 1 \right)$$



The relative bias of area estimation with pixel counting under fixed UA (left) and fixed PA (right)

# Relative efficiency of stratification

- Two classes  $K=2$ , true area of the target class  $f \equiv f_{.1} = 0.15$  and target CV=5%.

	<u>A</u> PA=0.9 UA=0.7	<u>B</u> PA=0.7 UA=0.9	<u>C</u> PA=0.8 UA=0.8	<u>D</u> PA=0.6 UA=0.6
	$f = 0.15$			
<b>Overall</b>	0.927	0.943	0.940	0.880
<b>F-score</b> = $\frac{2v_1u_1}{v_1+u_1}$	0.788	0.788	0.800	0.600
<b>Sample size <math>n</math></b>				1508
<b>Rel. eff. strat.</b>				1.50

Simple random sampling, sample size  $n=2,262$

# Relative efficiency of stratification

- Two classes  $K=2$ , true area of the target class  $f \equiv f_{.1} = 0.15$  and target CV=5%.

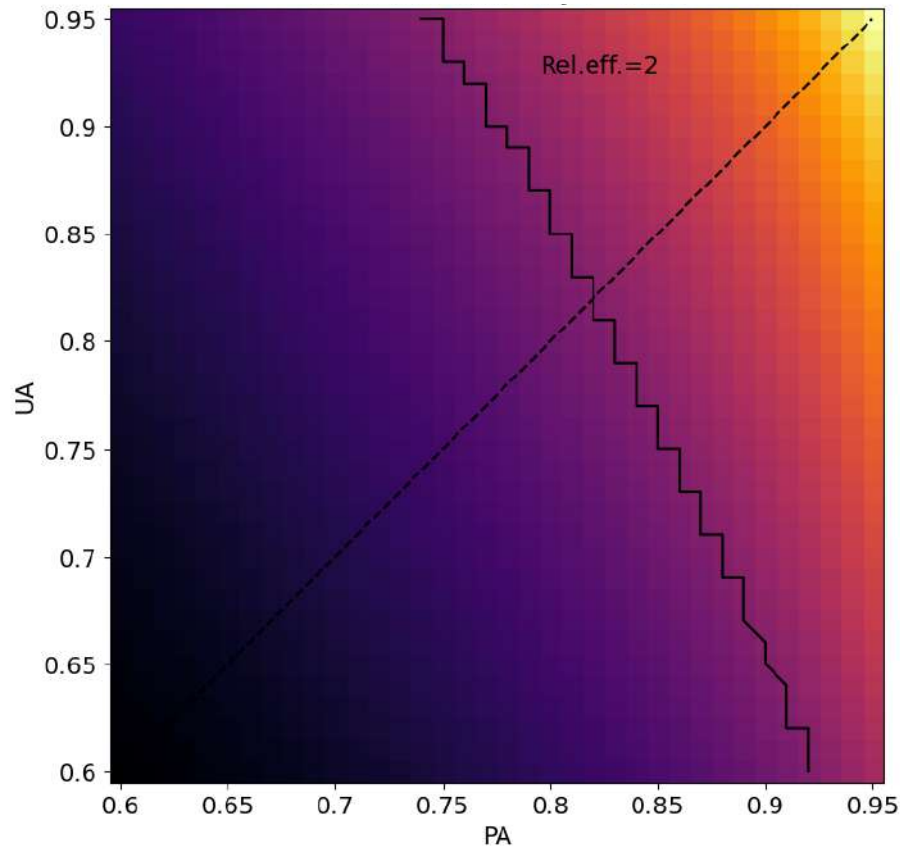
	<u>A</u> PA=0.9 UA=0.7	<u>B</u> PA=0.7 UA=0.9	<u>C</u> PA=0.8 UA=0.8	<u>D</u> PA=0.6 UA=0.6
	$f = 0.15$			
<b>Overall</b>	0.927	0.943	0.940	0.880
<b>F-score</b> = $\frac{2v_1u_1}{v_1+u_1}$	0.788	0.788	0.800	0.600
<b>Sample size <math>n</math></b>	693	934	836	1508
<b>Rel. eff. strat.</b>	3.27	2.43	2.71	1.50

Simple random sampling, sample size  $n=2,262$

# Relative efficiency of stratification

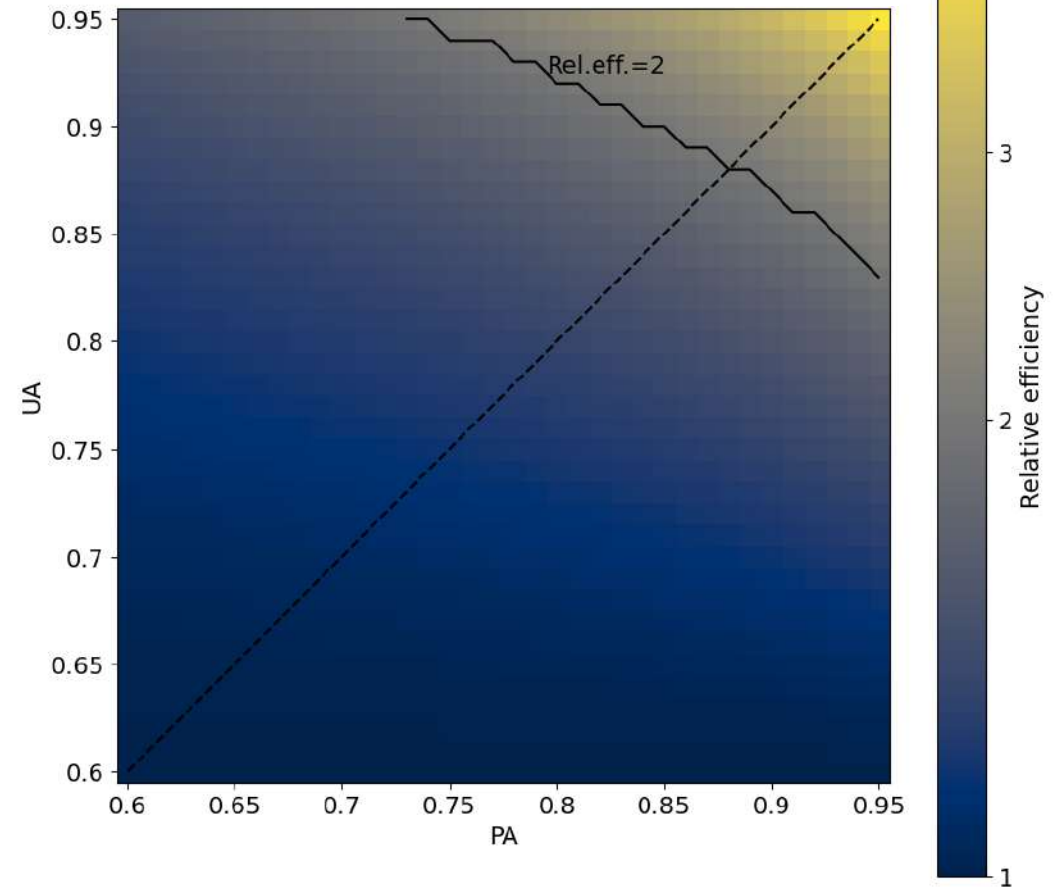
True area of the target class

$$f \equiv f_{.1} = 0.15$$



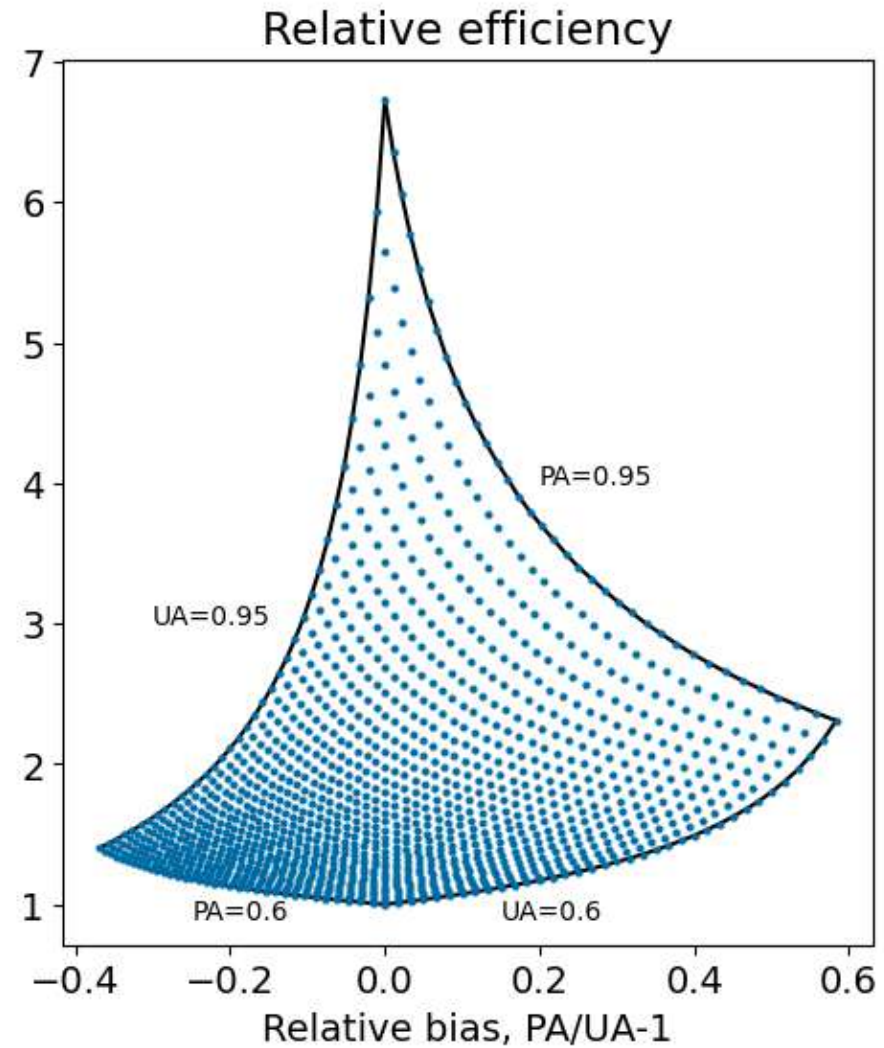
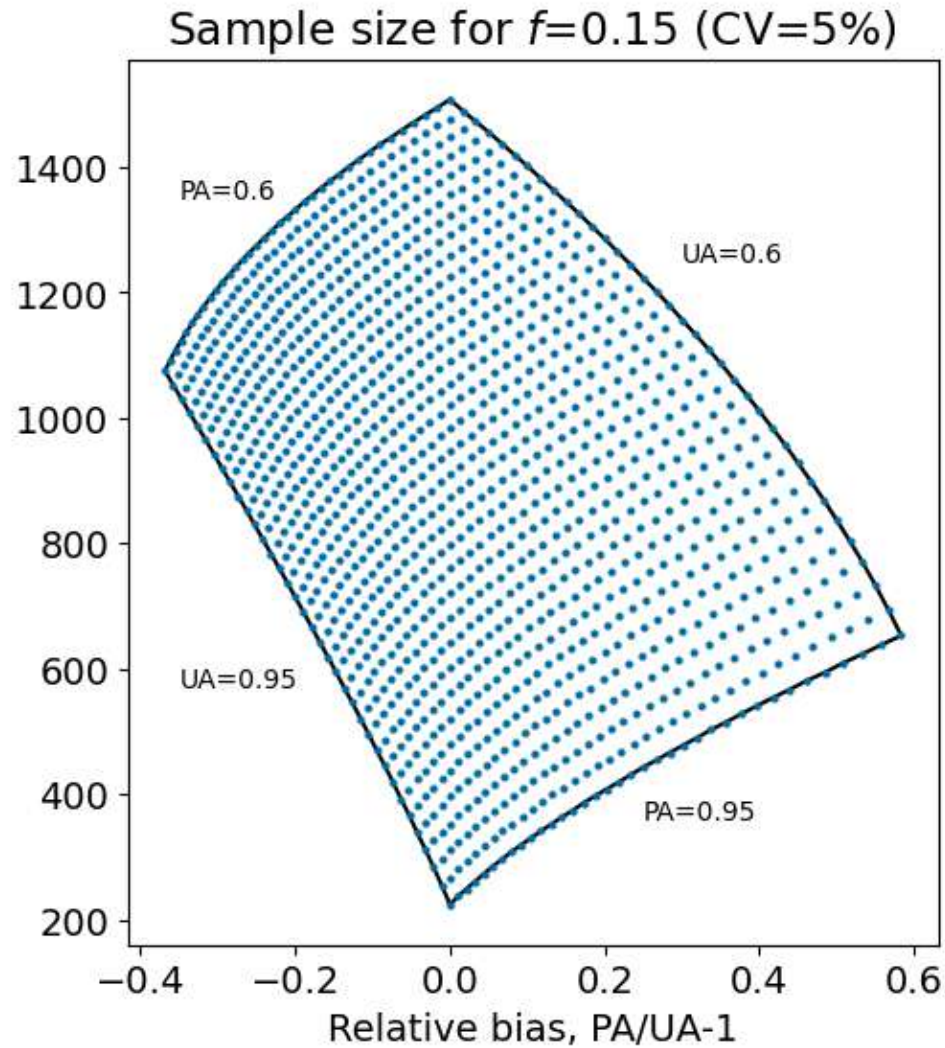
True area of the target class

$$f \equiv f_{.1} = 0.6$$



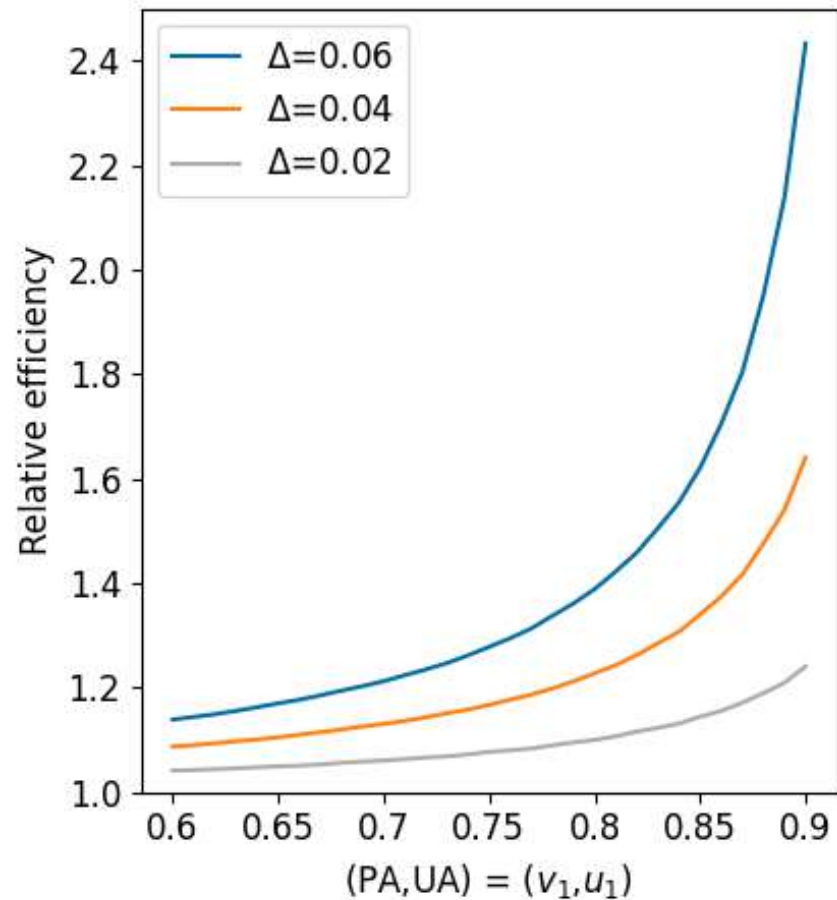
Relative efficiency  $\eta_{A,B}$  calculated for each pair  $(v_1, u_1)$  (map A) in relation to the map at  $(v_1, u_1) = (0.6, 0.6)$  (map B):  $f = 0.15$  (left) and  $f = 0.60$  (right). In each figure the solid line shows multiple solutions in the PA/UA space when a relative efficiency of 2 can be reached.



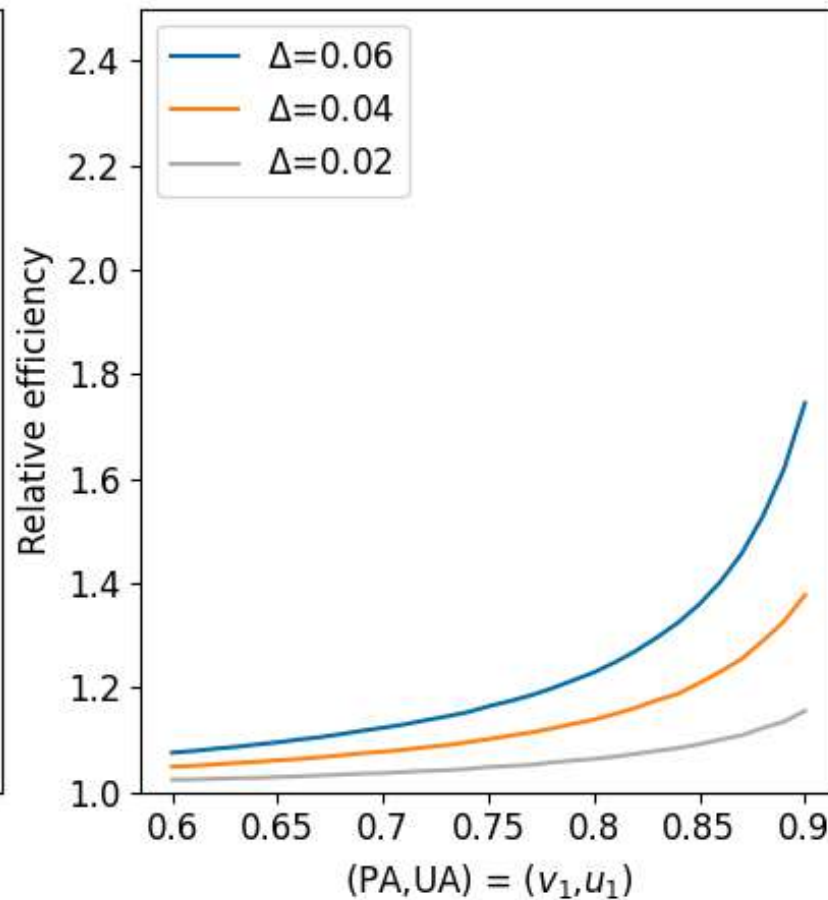


Dependence of the total sample size and relative efficiency (in relation to  $(v_1, u_1) = (0.6, 0.6)$ ) on the relative bias.  
Case  $f = 0.15$ .

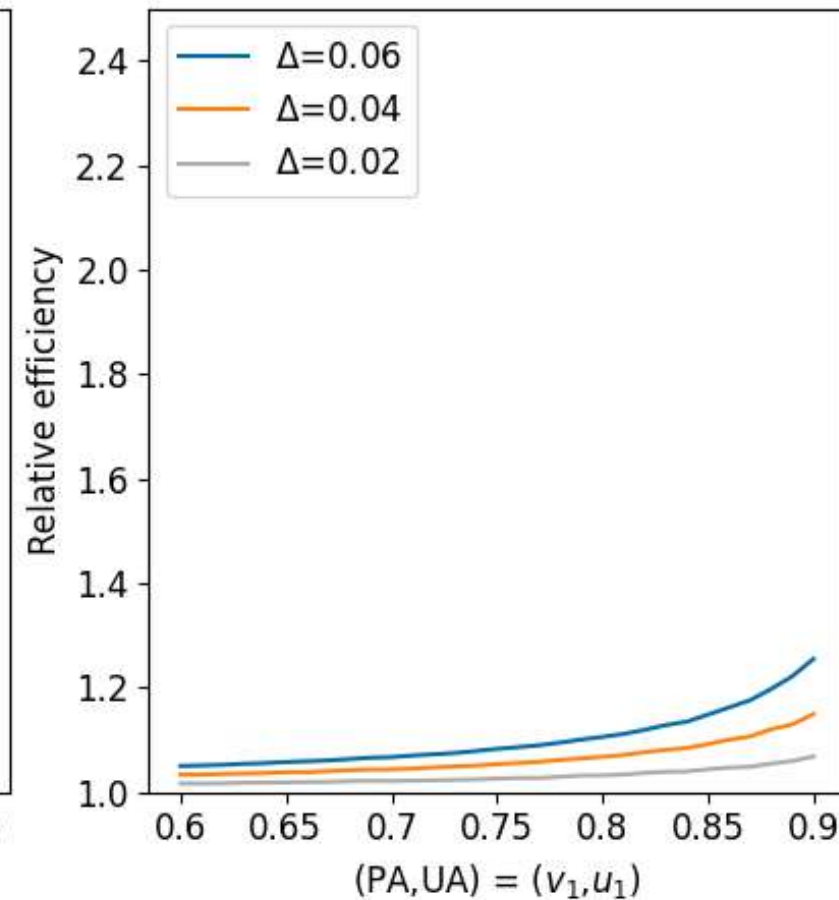
Efficiency on PA/UA increase ( $f=0.15$ ):  
 $(v_1, u_1) \rightarrow (v_1 + \Delta, u_1 + \Delta)$



Efficiency on PA increase ( $f=0.15$ ):  
 $(v_1, u_1) \rightarrow (v_1 + \Delta, u_1)$

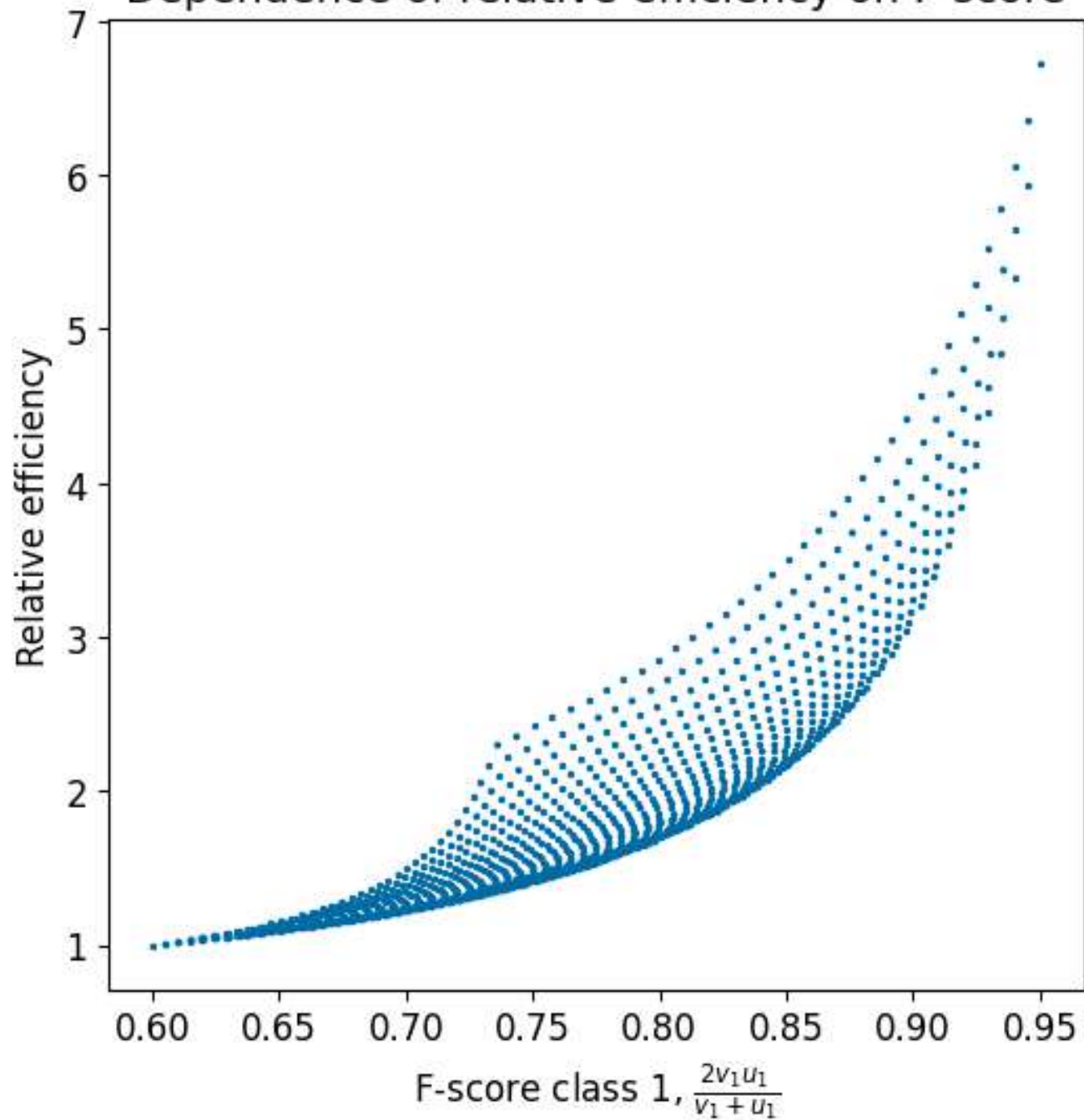


Efficiency on UA increase ( $f=0.15$ ):  
 $(v_1, u_1) \rightarrow (v_1, u_1 + \Delta)$

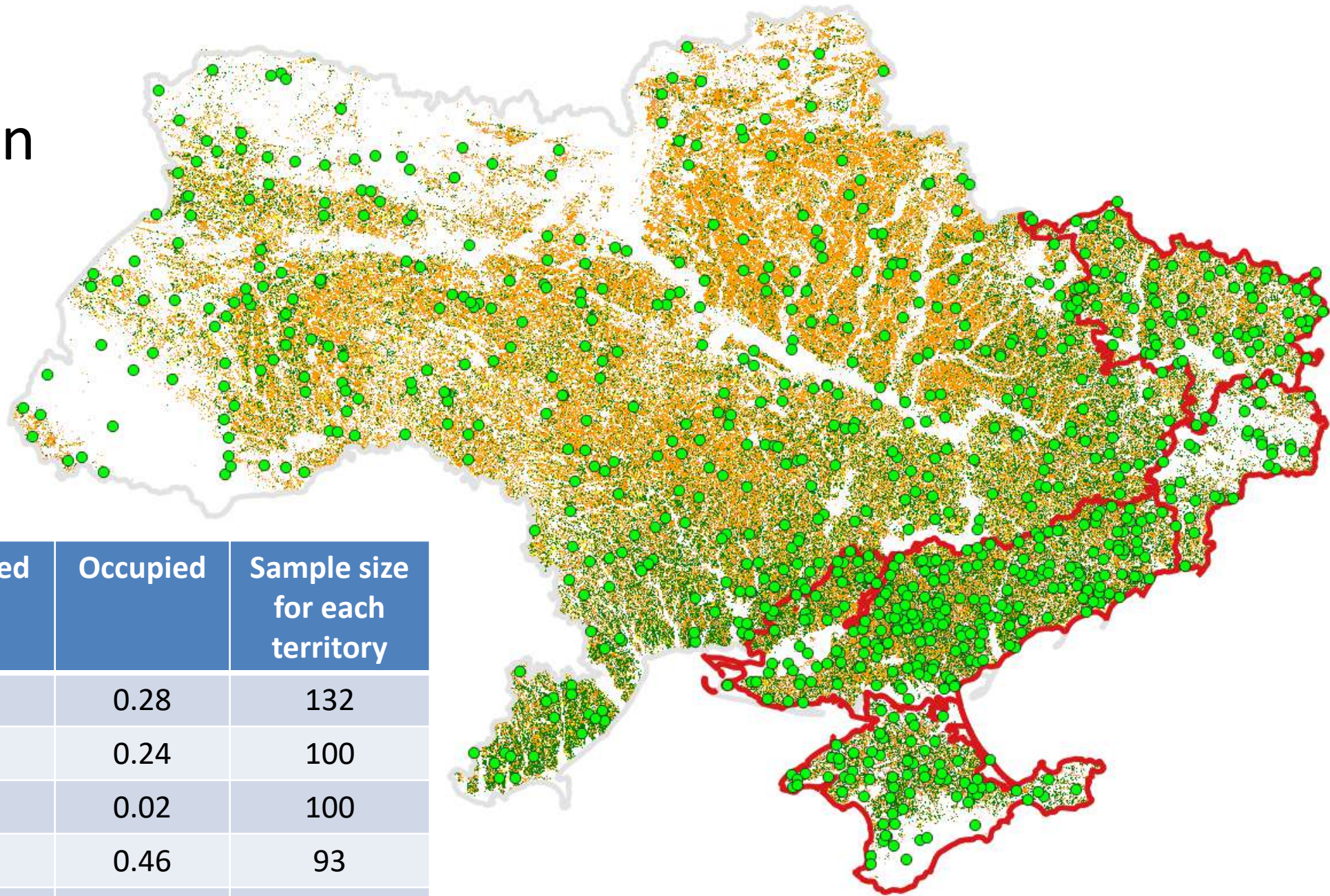


Dependence of improvements of relative efficiency when PA and UA increase by  $\Delta = +0.02$ ,  $+0.04$  and  $+0.06$ . Efficiency increase when both PA and UA increase by  $\Delta$  is shown in left figure; PA increase only is center figure; and UA increase only is right figure.

Dependence of relative efficiency on F-score



# Winter cereal area estimation in 2022



Stratum	Unoccupied	Occupied	Sample size for each territory
Summer crops	0.31	0.28	132
Winter cereal	0.13	0.24	100
Rapeseed	0.02	0.02	100
Non-cropland	0.53	0.46	93
Total	1	1	425

# Example

Description of sample data as an error matrix of sample counts,  $n_{ij}$

		Reference						
Map		1. Summer	2. Winter cereals	3. Rapeseed	4. Non-cropland	Total	Total mapped [ha]	$W_i$
	1. Summer	108	2	0	22			
	2. Winter cereals	24	70	0	6			
	3. Rapeseed	4	5	87	4			
	4. Non-cropland	10	4	0	79			
	Total							

Estimate of the proportion  $\hat{p}_{22}$  - mapped as "2. Winter cereals" and reference "2. Winter cereals"

$$\hat{p}_{22} = W_2 \frac{n_{22}}{n_{2\cdot}} = 0.240 \frac{70}{100} = 0.168$$

# Example

The error matrix populated by estimated proportions of area

		Reference				
Map		<i>1. Summer</i>	<i>2. Winter cereals</i>	<i>3. Rapeseed</i>	<i>4. Non-cropland</i>	Total $\hat{p}_i$
	<i>1. Summer</i>	0.2283	0.0042	0.0000	0.0465	
	<i>2. Winter cereals</i>	0.0576	0.1680	0.0000	0.0144	
	<i>3. Rapeseed</i>	0.0006	0.0008	0.0140	0.0006	
	<i>4. Non-cropland</i>	0.0500	0.0200	0.0000	0.3949	
	Total $\hat{p}_j$					

# Example

The error matrix populated by estimated proportions of area

		Reference					
Map		1. Summer	2. Winter cereals	3. Rapeseed	4. Non-cropland	Total $\hat{p}_{i.}$	$\hat{U}_i$
	1. Summer	0.2283	0.0042	0.0000	0.0465		
	2. Winter cereals	0.0576	0.1680	0.0000	0.0144		
	3. Rapeseed	0.0006	0.0008	0.0140	0.0006		
	4. Non-cropland	0.0500	0.0200	0.0000	0.3949		
	Total $\hat{p}_{.j}$						
	$\hat{P}_j$						

**Overall accuracy**  $\hat{O} = \sum_{j=1}^4 \hat{p}_{jj} = 0.2283 + 0.1680 + 0.014 + 0.3949 = 0.8052$

Estimate of **user's accuracy** of class 2 (winter cereals)

$$\hat{U}_2 = \hat{p}_{22} / \hat{p}_{2.} = 0.168 / 0.24 = 0.70, \text{ or } \text{commission error } 1 - \hat{U}_2 = 1 - 0.70 = 0.3$$

Estimate of **producer's accuracy** of class 2

$$\hat{P}_2 = \hat{p}_{22} / \hat{p}_{.2} = 0.168 / 0.1931 = 0.87, \text{ or } \text{omission error } 1 - \hat{P}_2 = 1 - 0.87 = 0.13$$

# Example

## Area estimation

		Reference						
Map		1. Summer	2. Winter cereals	3. Rapeseed	4. Non-cropland	Total $\hat{p}_i$	$\hat{U}_i$	Total mapped [ha]
	1. Summer	0.2283	0.0042	0.0000	0.0465	<b>0.279</b>	<b>0.82</b>	<b>3,470,363</b>
	2. Winter cereals	0.0576	0.1680	0.0000	0.0144	<b>0.240</b>	<b>0.70</b>	<b>2,985,323</b>
	3. Rapeseed	0.0006	0.0008	0.0140	0.0006	<b>0.016</b>	<b>0.87</b>	<b>199,502</b>
	4. Non-cropland	0.0500	0.0200	0.0000	0.3949	<b>0.465</b>	<b>0.85</b>	<b>5,781,756</b>
	Total $\hat{p}_j$	<b>0.3365</b>	<b>0.1931</b>	<b>0.0140</b>	<b>0.4565</b>	<b>1</b>		<b>12,436,944</b>
	$\hat{P}_j$	<b>0.68</b>	<b>0.87</b>	<b>1.00</b>	<b>0.87</b>		<b>0.81</b>	

Estimate of areas:

**Option 1** (directly from the map with pixel counting): **class 2 Winter cereals**

$$A_{m,1} = 2,985,323 \text{ ha}$$

**Option 2** (from confusion matrix): **class 2 Winter cereals**

$$\hat{A}_1 = \hat{p}_{.1} \times A_{\text{total}} = 0.1931 \times 12,436,944 \text{ ha} = 2,401,574 \text{ ha}$$

Pixel counting overestimates **winter cereals** by 583,749 ha!



# Example

## Area estimation

		Reference						
Map		1. Summer	2. Winter cereals	3. Rapeseed	4. Non-cropland	Total $\hat{p}_i$	$\hat{U}_i$	Total mapped [ha]
	1. Summer	0.2283	0.0042	0.0000	0.0465	<b>0.279</b>	<b>0.82</b>	<b>3,470,363</b>
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	Total $\hat{p}_j$	<b>0.3365</b>	<b>0.1931</b>	<b>0.0140</b>	<b>0.4565</b>	<b>1</b>		<b>12,436,944</b>
	$\hat{P}_j$	<b>0.68</b>	<b>0.87</b>	<b>1.00</b>	<b>0.87</b>		<b>0.81</b>	

Why is it happening?

Since  $\hat{P}_2 = 0.87$  meaning we miss winter cereal areas on average 13% (omission error), while  $\hat{U}_2 = 0.70$ , meaning that we commit on average 30% (commission error). Misbalance in omission and commission errors leads to a bias ( $PA/UA-1 = 0.87/0.7-1 = 24\%$ ) if using pixel counting!

Pixel counting overestimates **winter cereals** by 583,749 ha!

# Final areas: temporary occupied territories (as of July 2022)

	<b>Summer</b>	<b>Winter cereals</b>	<b>Winter rapeseed</b>	<b>Non-cropland</b>
Estimated area [ha]	4,185,540	2,400,960	173,567	5,676,877
95% CI of area [ha]	±499,631	±367,998	±13,217	±497,106

# Conclusions

- When **maps** are used for **area estimation** using a **sample-based** approach within the **design-based** inference framework:
  - **Map quality** impacts stratification **efficiency**
    - A more accurate map → **smaller sample size** to reach the desired precision of the estimate or
    - A more accurate map → **higher precision** when the sample size is fixed
  - A criterion (**relative efficiency**) ← **implications** of **accuracy** increase
    - Depends on class-specific PA and UA, and target class area
    - The impact is not linear and contributions of PA, UA, and  $f$  are not equal
    - F-score is not an adequate metric
    - Costs associated with reference data collection (ground surveys, photo-interpretation)
      - E.g, LUCAS [d'Andrimont et al., 2020], sample unit cost=35 Euro
      - Target  $f = 0.05$  (CV=5%), relative efficiency of 2 → 140,000 Euro (4,000 sample units) to 70,000 Euro (2,000)
      - Target  $f = 0.3$  (CV=5%), relative efficiency of 2 → 26,250 Euro (750) to 13,125 Euro (325).

# Conclusions (cont')

- Re-emphasized

- **Rigorous map quality assessment**, e.g., [Olofsson et al., 2014], otherwise, a map is a “pretty picture” [McRoberts, 2011])
- Full **area-based confusion matrix** to be reported → critical for further map use (esp., area)
- Pixel counting estimator → biased → discouraged and should not be used

- Future directions

- **Costs of map generation**, e.g., data costs & compute/storage costs, not considered

$$\eta_{A,B} = \frac{c_1^B + c_0 n^B}{c_1^A + c_0 n^A}$$

- Sample unit is a block (segment)