

# Privacy and data confidentiality for Official Statistics: new challenges and new tools

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**EUROSTAT - Unit B1** 

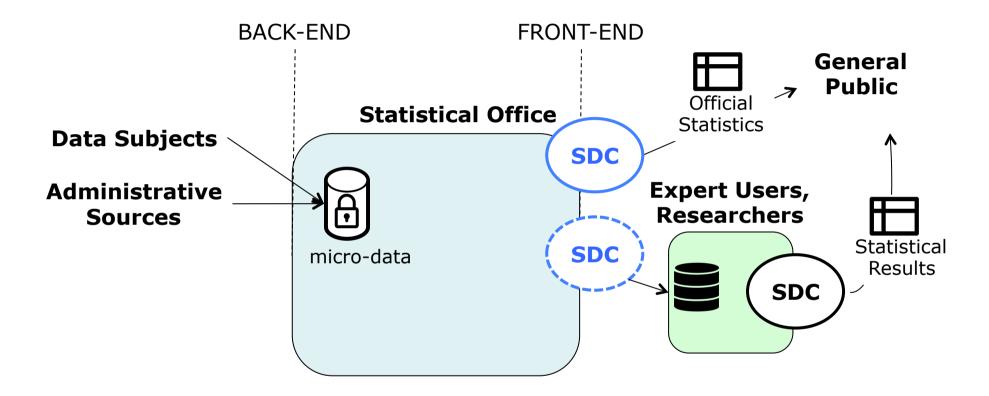
NTTS 2019, Brussels 14.3.2019

# **Statistical Disclosure Control (SDC)**

- Suppression (e.g. cell deletion, column removal)
- Add noise, perturbation, rounding

								Town	Count all	Count sick	Averag Income
Town	Count all	Count sick	Average Income		SDC		$\mathbb{H}$	Town	Count all	Count sick	Avera Incon
							Ш				
Smallville	5	1	51					Smallville	6	2	59
Midpoli	85	7	40678		_			Midpoli	88	7	4040
Largetown		45	89					Largetown	5773	44	89
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			er In		Town  Smallville	Sick 1	S	DC /		 1	
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Eva Fabio Elisa Oscar	23 38 78 32	 F M	7	 10 30 100 23	Smallville Largetown Largetown Midpoli	1 0	S	100 23 40000 30	Largetown Largetown Smallville Midpoli Midpoli Largetown	 1 0 1 0 0 0	
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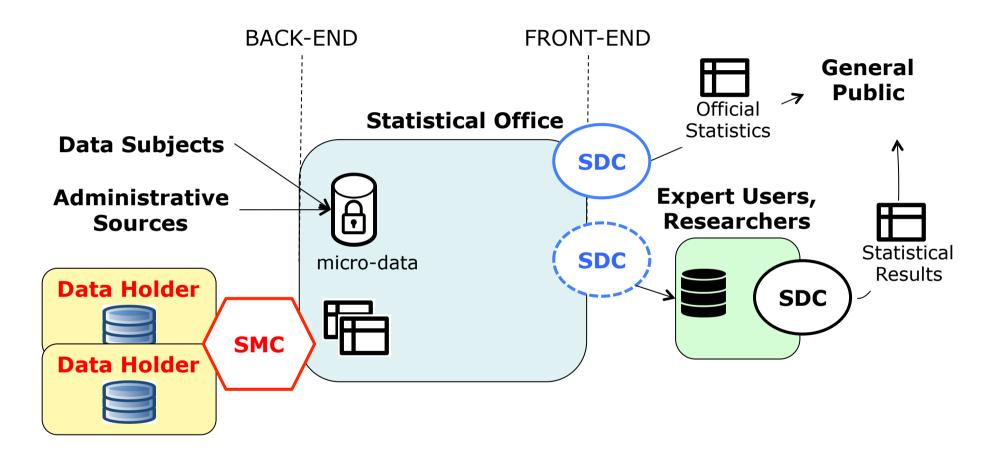
#### SDC on the front-end



SDC: Statistical Disclosure Control



#### SMC on the back-end



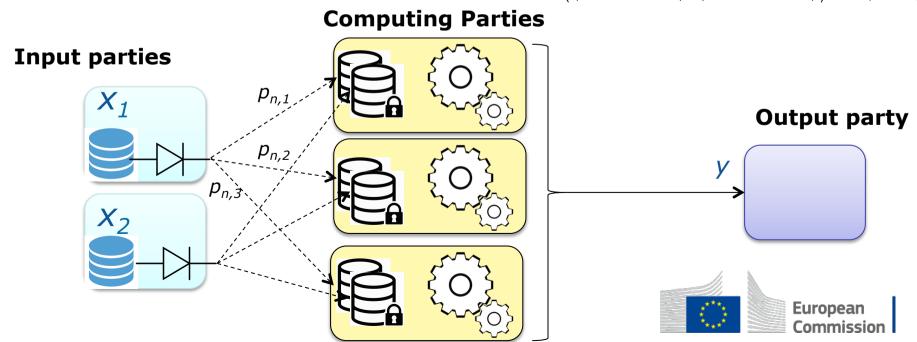
SDC: Statistical Disclosure Control

SMC: Secure Multi-Party Computation



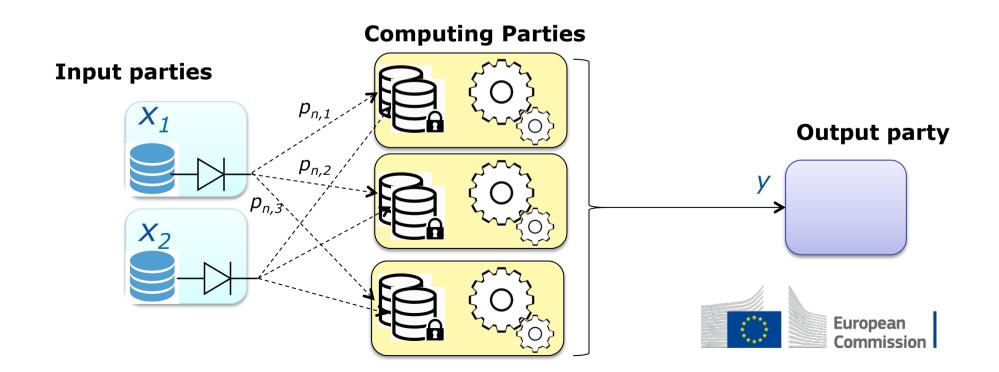
# **Secure Multi-Party Computation (SMC)**

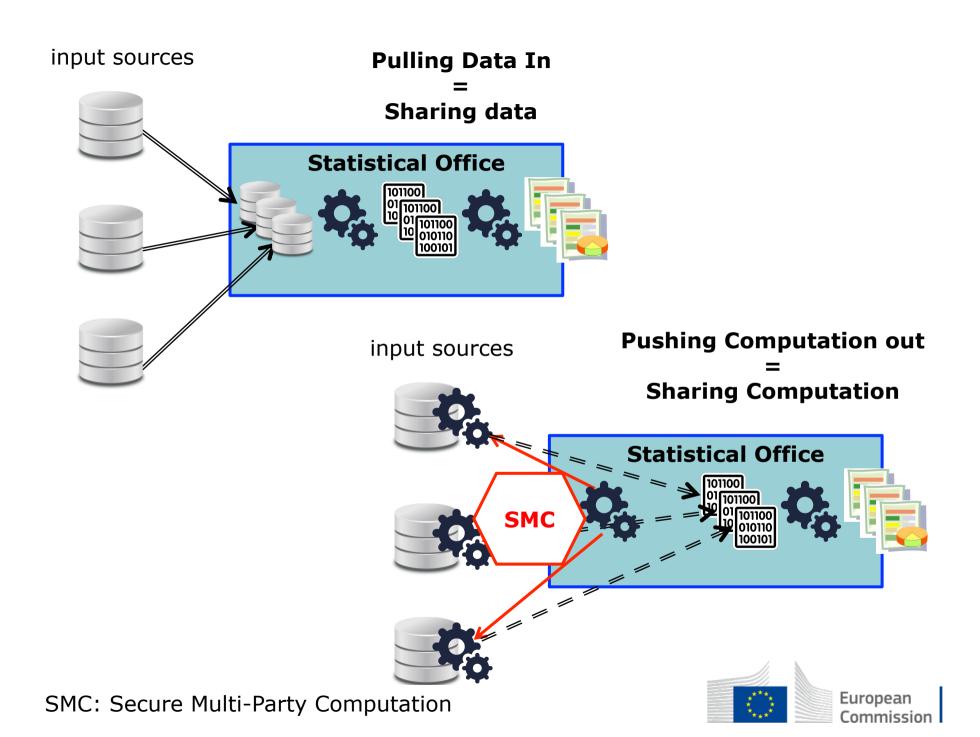
- Each element of secret input  $x_n$  is transformed into K "shares"  $p_{n,1}, p_{n,2} \dots p_{n,k}$  that are distributed to different computing parties.
- The computation on secret shares
  - is distributed (shared) among the computing parties
  - returns the same output value that would be obtained from the input data (homomorfism)  $y = f_s \left( \langle p_{1,1}, p_{1,2}, p_{1,3} \rangle, \langle p_{2,1}, p_{2,2}, p_{2,3} \rangle \right) = f\left(x_1, x_2\right)$



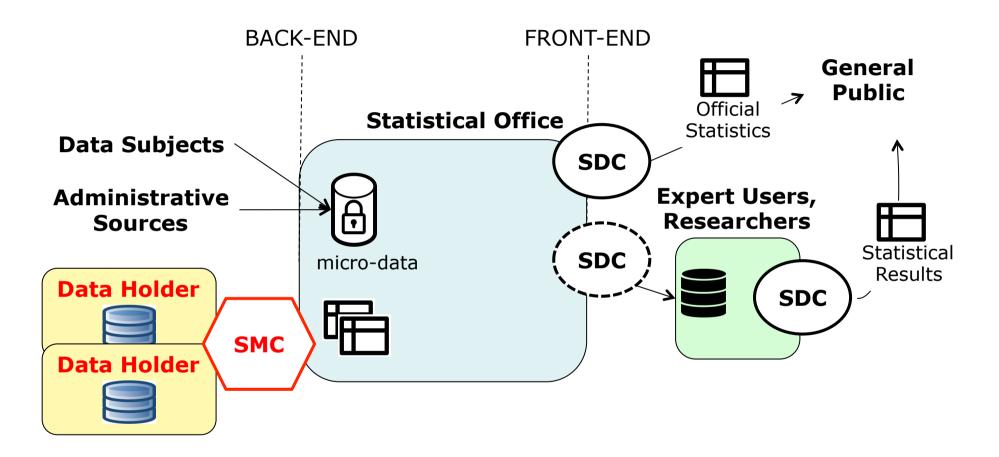
## **Secure Multi-Party Computation (SMC)**

- Individual shares do not reveal nothing about the secret input
  - → no single party holds "data"
    - → "passing shares" ≠ "sharing data"
- Computing parties need to be trusted collectively, not individually





#### SMC on the back-end

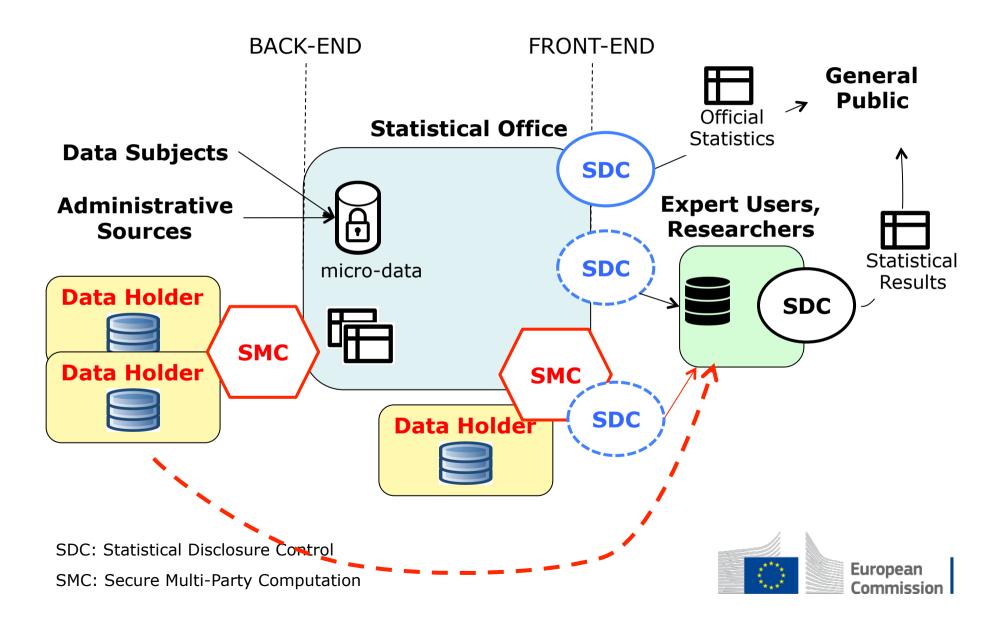


SDC: Statistical Disclosure Control

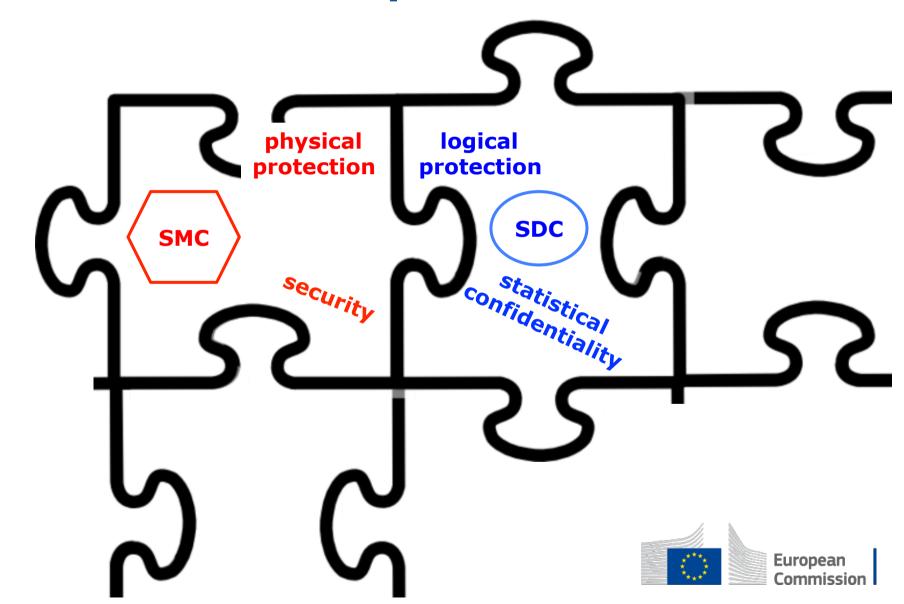
SMC: Secure Multi-Party Computation



# **Combining SMC+SDC on the front-end?**



# SMC & SDC as complementary but distinct components



### Take-home messages

Confidentiality in Official Statistics need to evolve towards more articulated solutions

Evolution of SDC solutions from traditional **static tools** solutions towards **dynamic SDC** <sub>Table Builder, on-the-fly anonymization</sub> is part of the story

SMC can complement (not replace!) SDC in multi-source scenarios

Towards a system-level view of "confidentiality engineering"

- learn to compose multiple elements/layers/components in a consistent design (technology, legal, organizational)
- centrality of feasible attack models, analysis & minimization of risks





# Thanks for your attention

For follow-up:

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