

# DuraClone, CytoFLEX LS, DxFLEX, NAVIOS

KAI

# COVID 19





Μαρία Γεωργίου LERIVA DIAGNOSTICS







#### PERSPECTIVE





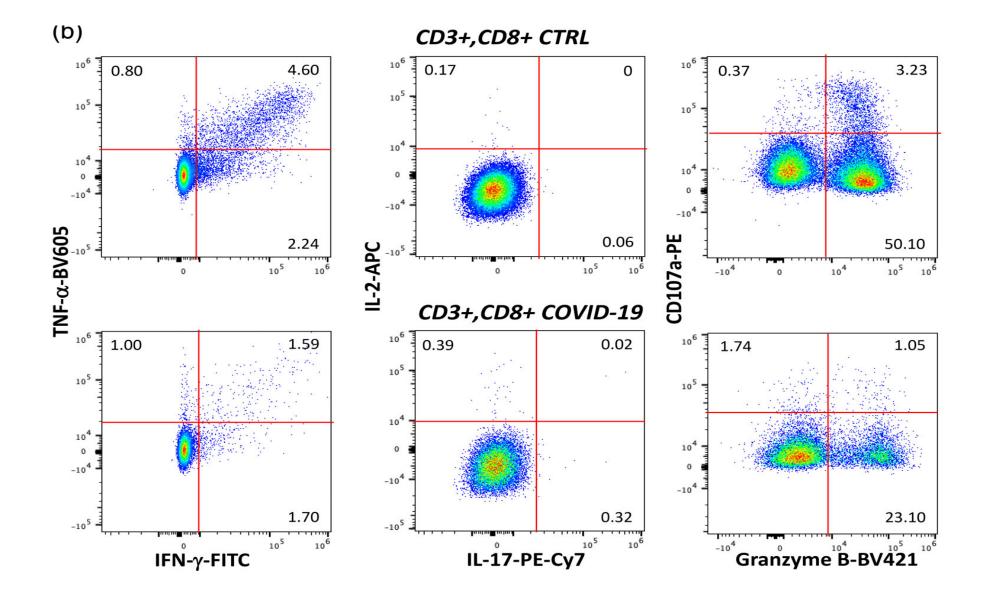
## SARS-CoV-2, the Virus that Causes COVID-19: Cytometry and the New Challenge for Global Health

Andrea Cossarizza,<sup>1\*</sup> Sara De Biasi,<sup>1†</sup> Giovanni Guaraldi,<sup>2</sup> Massimo Girardis,<sup>3</sup> Cristina Mussini,<sup>2</sup> for the Modena Covid-19 Working Group (MoCo19)<sup>#</sup>

• Key terms SARS-CoV-2; Covid-19; coronavirus; cytometry; CD4; CD8; T cells







(a) CD3+,CD4+ CTRL 106 106 106 0.39 0.30 0.29 1.77 0.05 0.02 105 105 105 104 104 10 0 -104 -10 -10 TNF-α-BV605 0.23 CD107a-PE 0.34 4.22 IL-2-APC -105 104 105 105 10<sup>5</sup> 10<sup>6</sup> 106 106 -104 0 0 0 *CD3+,CD4+ COVID-19* 106 106 106 5.60 1.10 0.36 1.48 0.15 0.12 105 105 105 104 104 10 0 0 -104 -104 -10 0.65 0.68 4.19 -105 -105 Granzyme B-BV421 10<sup>5</sup> 105 106 IL-17-PE-Cy7 -104 0 106 IFN-γ-FITC

## DURACIone\* & DURActive\* Product Families

	405 excitation		480 excitation					633 excitation		
	Pacific Blue	Krome Orange	FITC	PE	ECD	PC5.5	PC7	APC	APC- A700	APC- A750
DURACIone IF T Helper Cell	IL-17		IFN-γ				IL-4	CD4		CD3
DURAClone IF T Activation Tube	CD4		IFN-γ	TNFα			IL-2		CD8	CD3

DURAClone IM Treg Tube Helios	CD45	CD45RA	CD25		CD39	CD4	FoxP3		CD3
DURAClone IM T cell subsets CD57	CD45	CD45RA	CD197	CD28	CD279	CD27	CD4	CD8	CD3





### DURAClone\* & DURActive\* Product Families

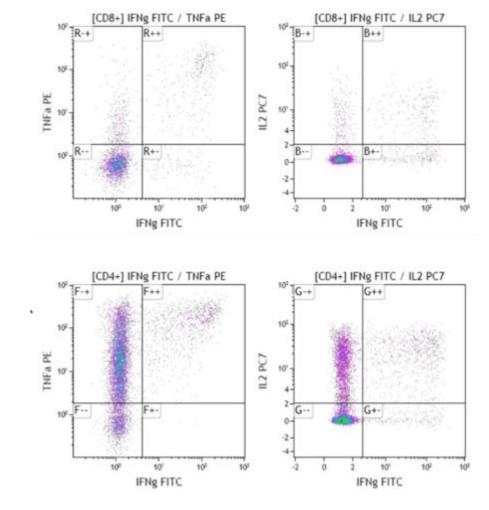
### DURActive 1 25 tests RUO C11101

Dry Stimulation Mixes for Immune Functional Assays PMA, Ionomycin, Brefeldin A

### DURActive 2 25 tests RUO C11102

Dry Stimulation Mixes for Immune Functional Assays PMA, Ionomycin

**DURActive 3** 25 tests RUO C21857 Dry Stimulation Mixes for Immune Functional Assays LPS, Brefeldin A



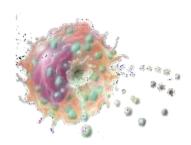




#### **REVIEW ARTICLE**



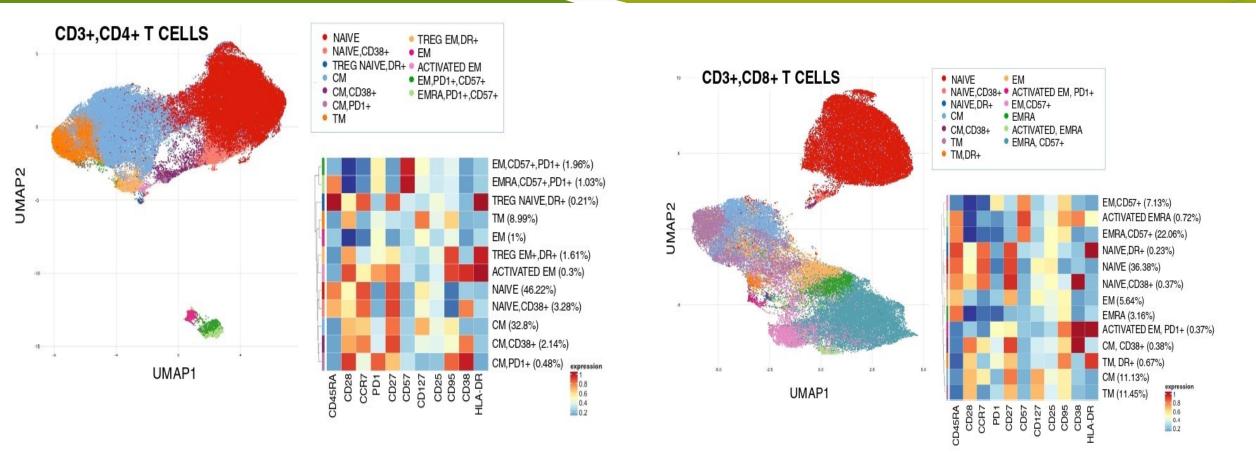




#### **Relevance of Antibody Validation for Flow Cytometry**

Tomas Kalina,1\* Kelly Lundsten,2 Pablo Engel3

Thus, at present, **multiparametric cytometry assays require antibody conjugates with known performance criteria under several conditions;** for several cell types, validation data shall be presented for monoclonal antibody reactivity and for antibody conjugate performance. **Consensus on benchmarking methods**, aggregation of comparable data sets across manufacturers and users and public availability of the performance data on clones as well as on antibody conjugates will lead to a qualitatively higher level of information generated by flow cytometry and to a further spread of its applications to basic, translational, and clinical research.

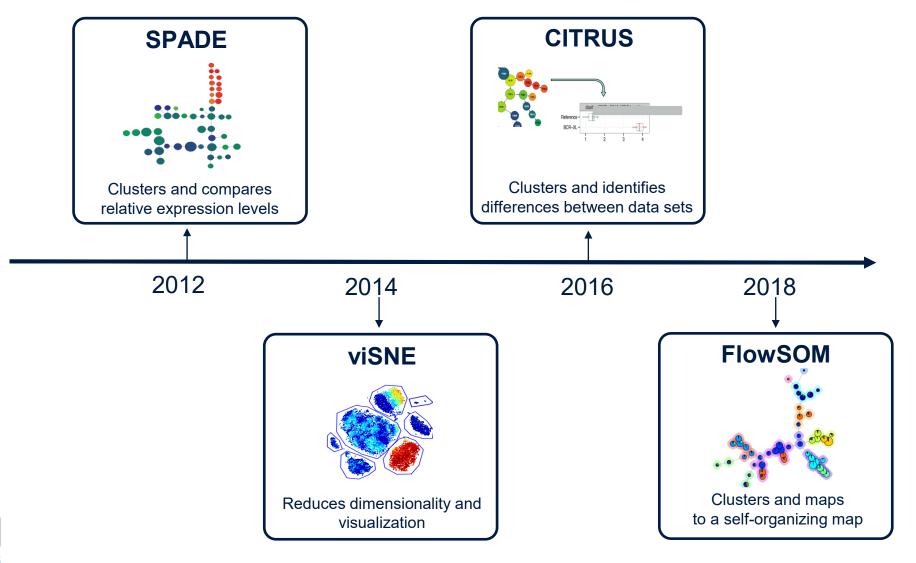


**Figure 1.** Differentiation, activation and exhaustion of CD4+ or CD8+ T-cell subsets in patients with Covid-19. Representation of an 18-parameter analysis of peripheral blood cells. Cells were stained with the Duraclone IM T cell panel (from Beckman Coulter, FL, USA) added with another five fluorescent mAbs and a marker of cell viability and analyzed on a CytoFLEX LX flow cytometer (Beckman Coulter). Beside side and forward scatters, markers were CD45 conjugated with Krome Orange, CD3 APC-A750, CD4 APC, CD8 AF700, CD27 PC7, CD57 Pacific Blu, CD279 (PD-1) PC5.5, CD28 ECD, CCR7 PE, CD45RA FITC, HLA-DR BUV661, CD127 BV650, CD25 BV785, CD95 BUV395, CD38 BUV496, and PromoFluor-840 (Promokine). Unsupervised analysis of electronically gated CD45+, CD3+, CD4+ or CD45+, CD3+, CD8+ T-cells was first performed by using the Catalyst package (Bioconductor) (16). Second, for both CD4+ and CD8+ T-cell analysis, 10,000 cells per sample were concatenated and transformed. FlowSOM was used to perform the metaclustering (*K* = 20); then, data were represented by the dimensionality reduction method named Uniform Manifold Approximation and Projection (UMAP). It is possible to observe the presence of 12 clusters among CD4+ and 13 among CD8+ T-cells and to see the high amount of naive cells in both T cell populations, and how the distribution of naive, memory, activated and exhausted lymphocytes is different among different types of CD4+ or CD8+ T lymphocytes.



Unsupervised and Supervised Machine Learning Algorithms in Cytobank

Diagnostics





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