

A First Nowcasting System for the Complex Orography of the Canary Islands

David Quintero Plaza, AEMET, Spain. EWGLAM-SRNWP Meeting, September 30th, 2021

Index.

- The Canary Islands and the difficult forecasts for the islands.
- Optical Flow as a nowcasting tool for the islands. First results and validation.
- Conclusions and future work.

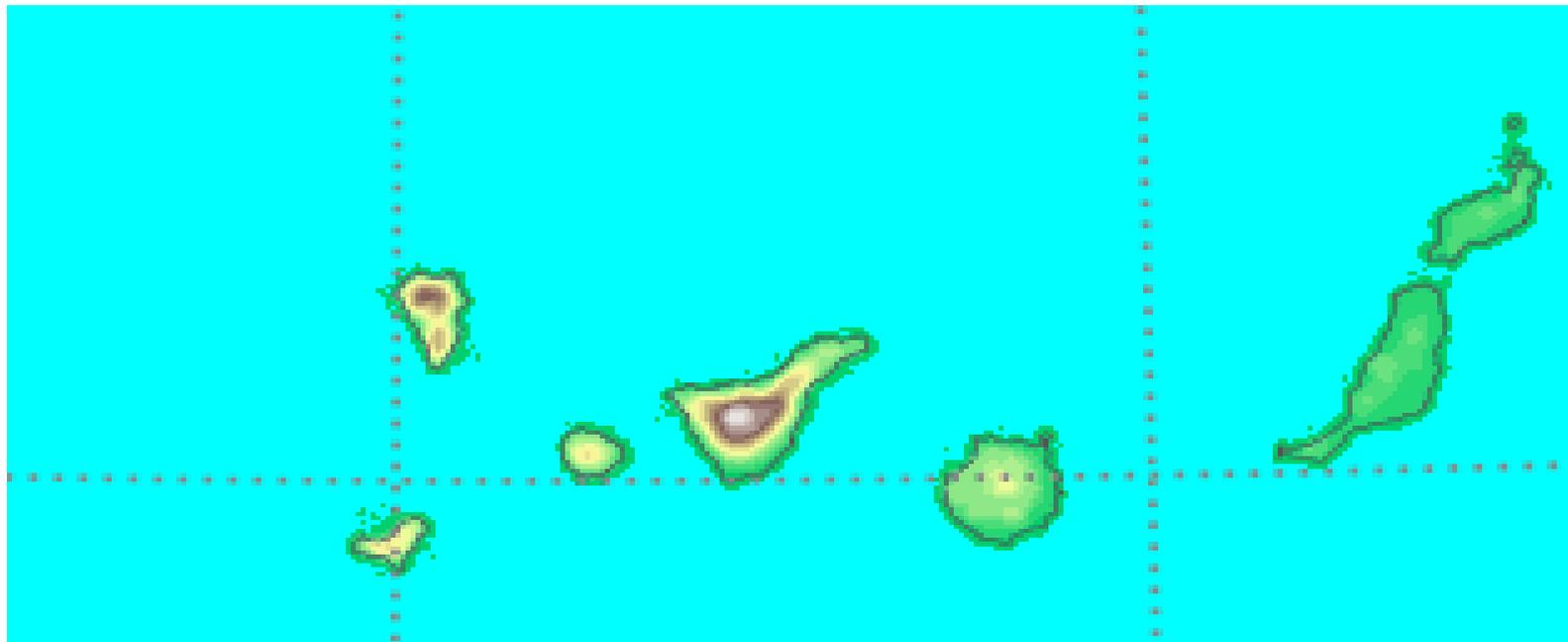
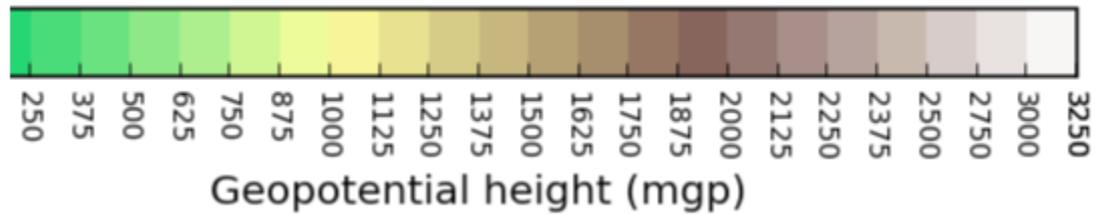
- The islands have a complex, mountainous and fastly changing terrain, with the so called local “micro-climates”.



- Even High Resolution Models (Harmonie & gSREPS at 2.5 Km) still show limitations for the islands (not to mention models with less resolution).

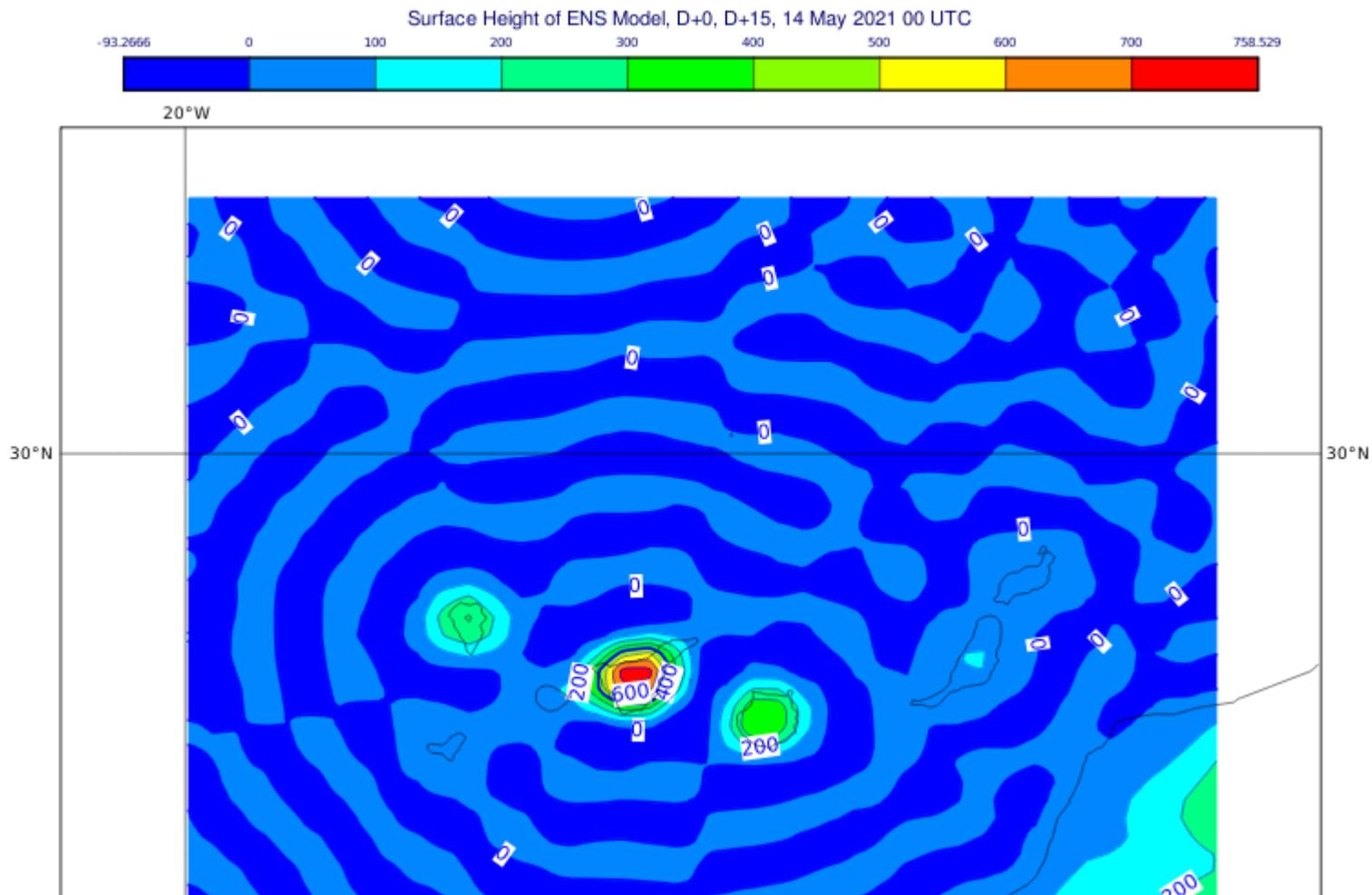
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- Below: gSREPS orography.



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- Below: ECMWF ENS orography.



- Regarding the nowcasting: 3 possibilities,
 - Fast Integration Models.
 - Optical Flow.
 - Neural Networks and other Machine Learning approaches.
- The variable chosen was **SRI (Surface Rainfall Intensity)**, an estimated value of the instant precipitation using data from all the vertical, as offered by the IRIS SIGMET-VAISALA radar, **each 10 minutes**.
- The nowcasting method chosen was Optical Flow, more specifically, *rainymotion*, by Georgy Ayzel: <https://rainymotion.readthedocs.io/en/latest/>
- Beware of the limitations: radar located at 1778 meters above sea level!

- Optical Flow:

Estimate motion between 2 small instants of time. Assume conservation of brightness:

$$I(x, y, t) = I(x + \Delta x, y + \Delta y, t + \Delta t)$$

Taylor expanding at first order and taking limits: $\frac{\partial I}{\partial x} V_x + \frac{\partial I}{\partial y} V_y + \frac{\partial I}{\partial t} = 0$

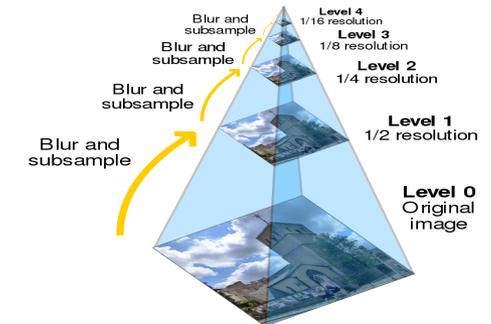
One equation, 2 unknowns (V 's, *aperture problem*): extra assumptions needed.

Typical assumption: Lucas-Kanade \Rightarrow flow constant in some region. The problem now has more equations than unknowns (over-determined). Solved with least squares.

$$\begin{aligned} I_x(q_1)V_x + I_y(q_1)V_y &= -I_t(q_1) \\ I_x(q_2)V_x + I_y(q_2)V_y &= -I_t(q_2) \\ \vdots & \\ I_x(q_n)V_x + I_y(q_n)V_y &= -I_t(q_n) \end{aligned} \quad \begin{bmatrix} V_x \\ V_y \end{bmatrix} = \begin{bmatrix} \sum_i I_x(q_i)^2 & \sum_i I_x(q_i)I_y(q_i) \\ \sum_i I_y(q_i)I_x(q_i) & \sum_i I_y(q_i)^2 \end{bmatrix}^{-1} \begin{bmatrix} -\sum_i I_x(q_i)I_t(q_i) \\ -\sum_i I_y(q_i)I_t(q_i) \end{bmatrix}$$

- Optical Flow:

rainymotion has 2 approaches: Sparse and Dense. Sparse: look for corners and evaluate and propagate velocity through Lucas-Kanade.



Dense (Kroeger et al., 2016): from coarse to fine (pyramid):

- 1) Create regular grid with overlapping patches
- 2) Find velocity of displacement (Lucas-Kanade)
- 3) Weighted averaging of displacements
($\lambda = 1$ means overlap between patch i and point x ,
 d measures difference of intensities)
- 4) Variational “energy” minimization of U
($\psi(E) \propto E$)

$$\mathbf{U}_s(\mathbf{x}) = \frac{1}{Z} \sum_i^{N_s} \frac{\lambda_{i,\mathbf{x}}}{\max(1, \|d_i(\mathbf{x})\|_2)} \cdot \mathbf{u}_i$$

$$Z = \sum_i \lambda_{i,\mathbf{x}} / \max(1, \|d_i(\mathbf{x})\|_2)$$

$$E(\mathbf{U}) = \int_{\Omega} \sigma \Psi(E_I) + \gamma \Psi(E_G) + \alpha \Psi(E_S) dx$$

$$E_I = \mathbf{u}^T \bar{\mathbf{J}}_0 \mathbf{u} \quad \bar{\mathbf{J}}_0 = \beta_0 (\nabla_3 I)(\nabla_3^T I)$$

$$E_G = \mathbf{u}^T \bar{\mathbf{J}}_{xy} \mathbf{u}$$

$$\bar{\mathbf{J}}_{xy} = \beta_x (\nabla_3 I_{dx})(\nabla_3^T I_{dx}) + \beta_y (\nabla_3 I_{dy})(\nabla_3^T I_{dy})$$

$$\nabla_3 = (\partial x, \partial y, \partial z)^T$$

$$E_S = \|\nabla u\|^2 + \|\nabla v\|^2$$

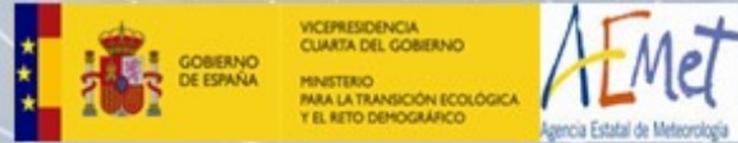
- Optical Flow:

Dense propagates velocity allowing for rotation (semi-Lagrangian field) or with a constant vector, using times t and $t-1$.

After propagating pixels to $t+n$, with the calculated velocities, new intensities are interpolated between t and $t+n$ with **inverse distance weighting**.

rainymotion offers other options and tweaking of the parameters.

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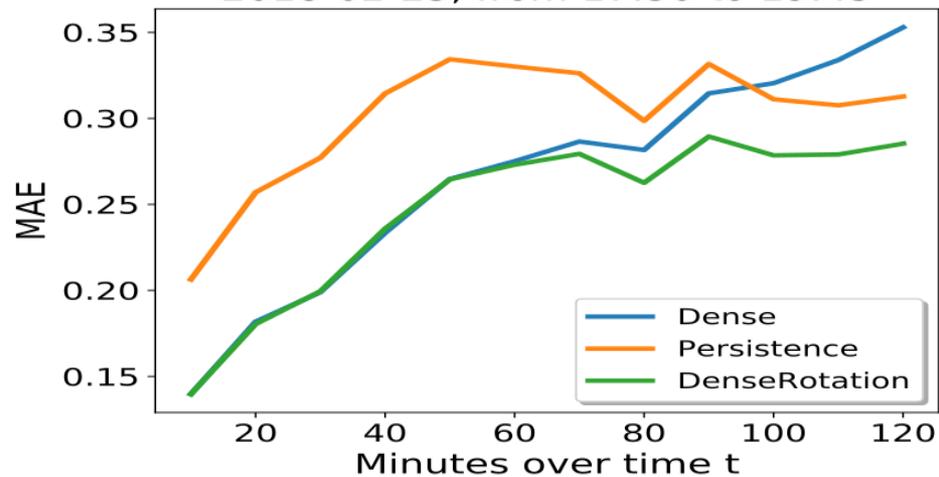


- **Our results:** 5 heavy rainfall situations under study. Scores: CSI (accuracy; higher => better) and MAE (error; lower => better).

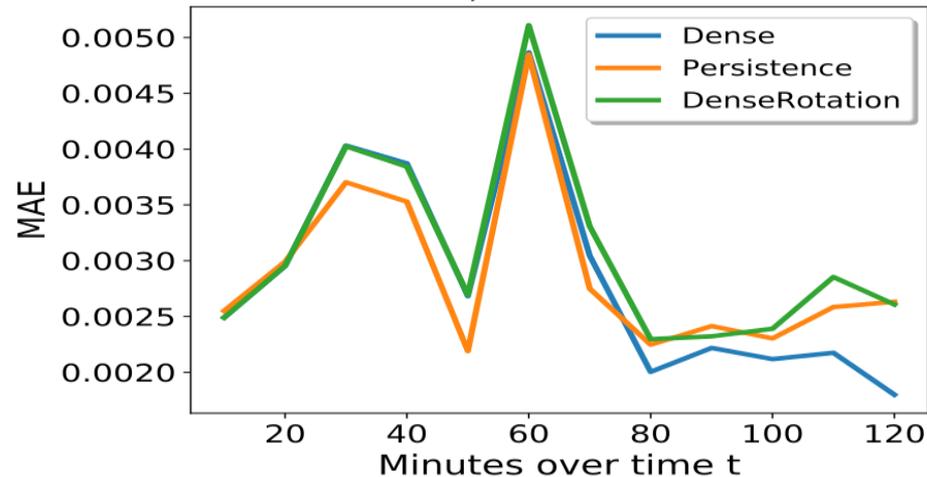
$CSI = \text{hits} / (\text{hits} + \text{misses} + \text{false alarms})$

$MAE = (1/N) \sum |obs - fcst|$

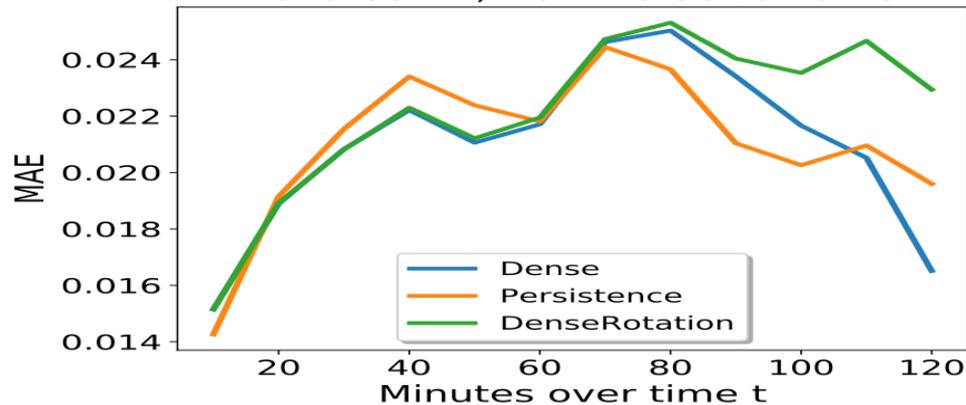
2018-02-23, from 17:30 to 19:45



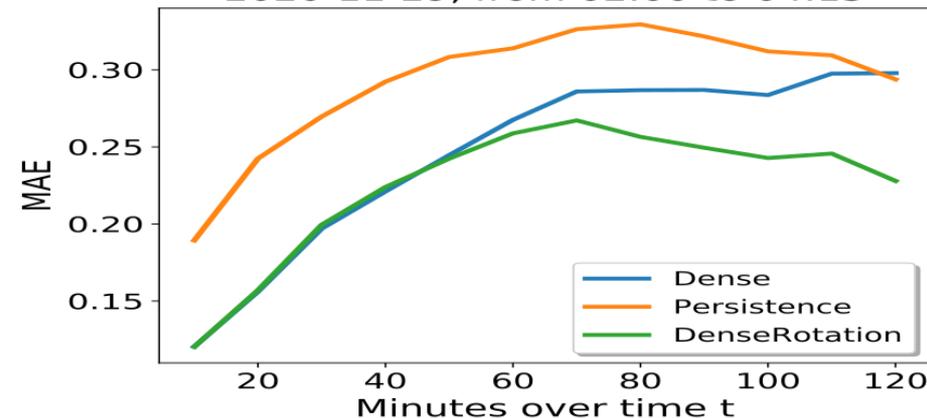
2018-10-10, from 21:30 to 23:45



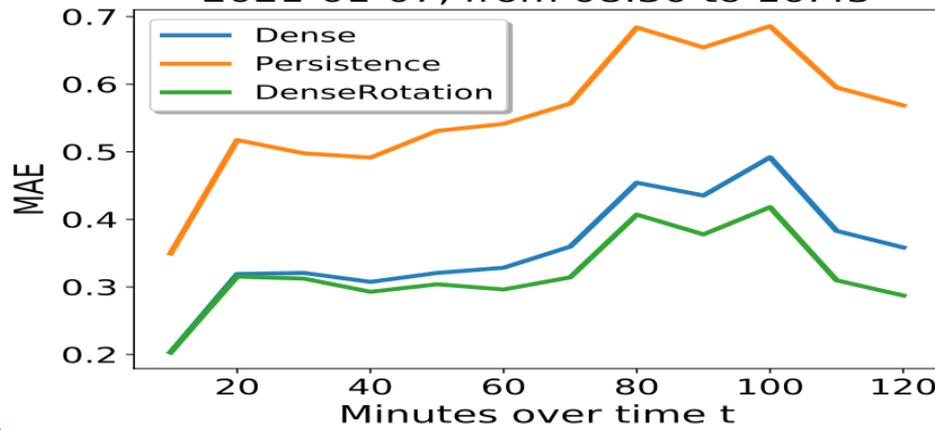
2019-03-27, from 13:30 to 15:45



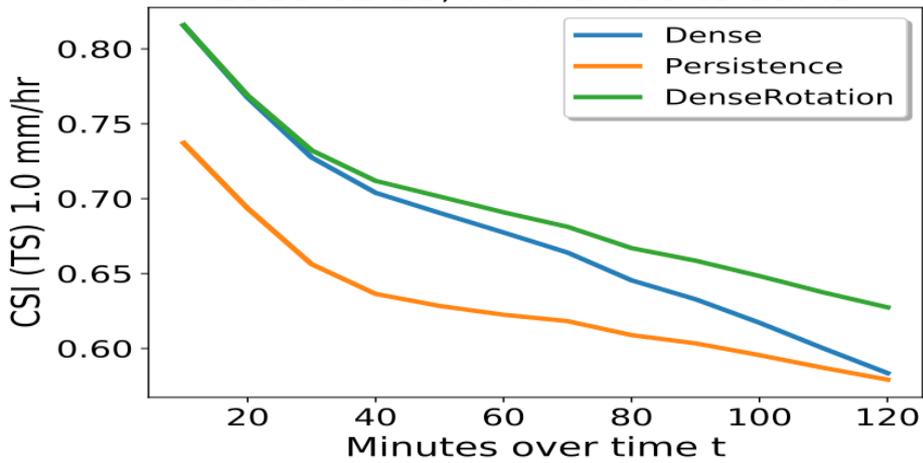
2020-11-29, from 02:00 to 04:15



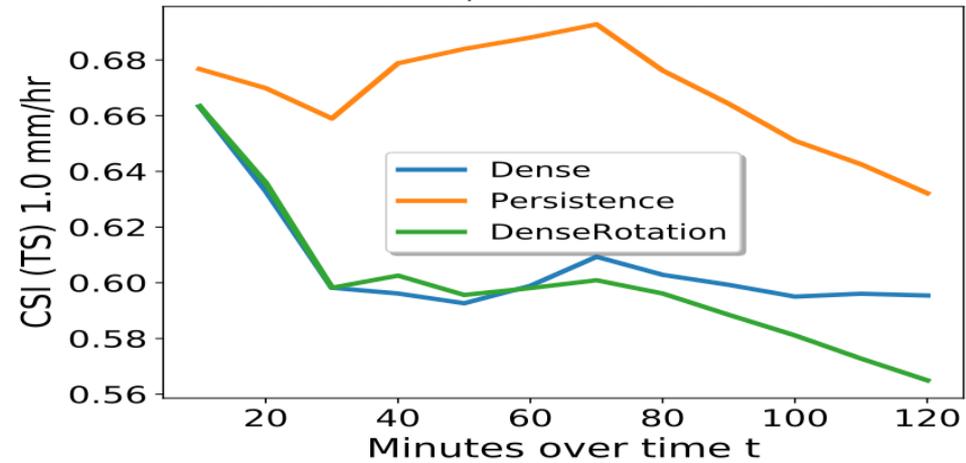
2021-01-07, from 08:30 to 10:45



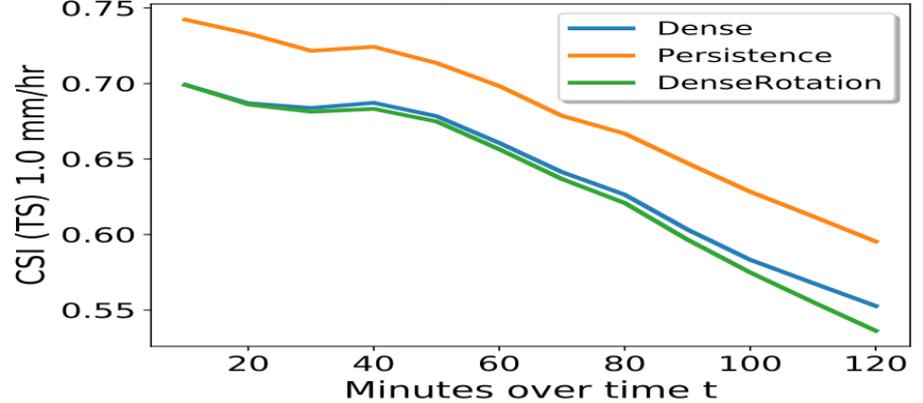
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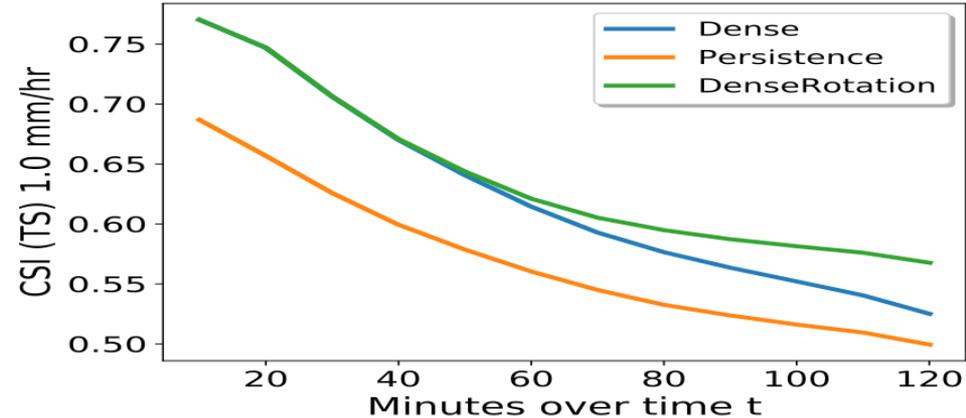
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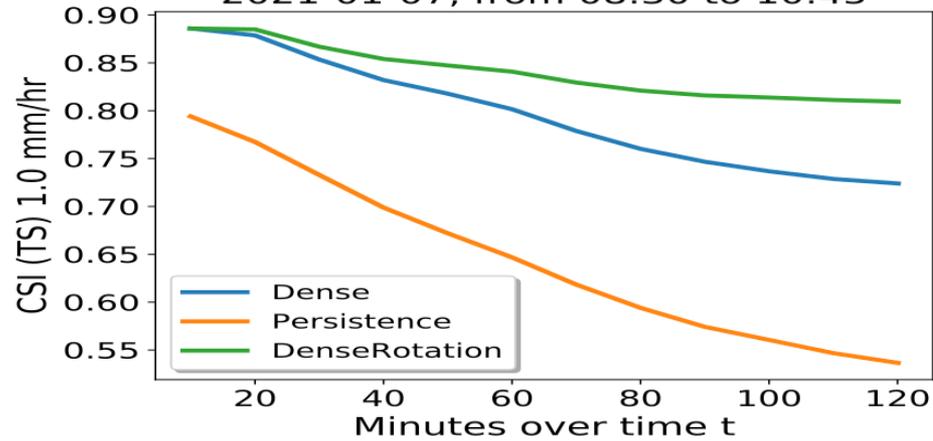
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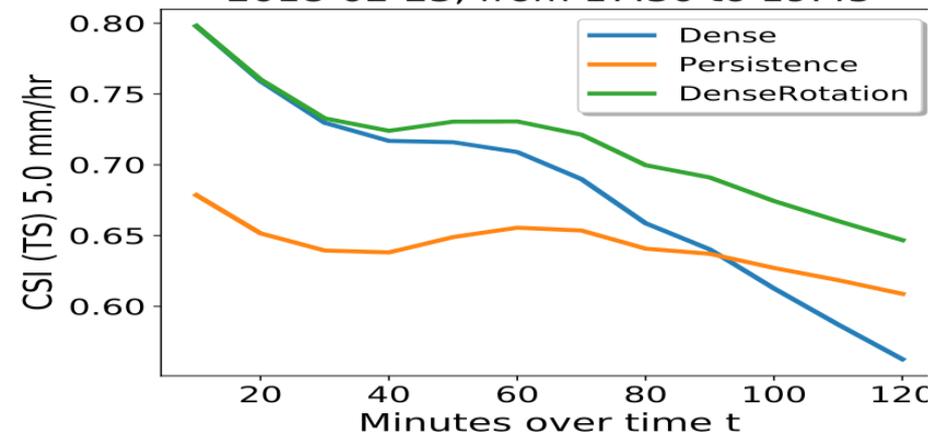
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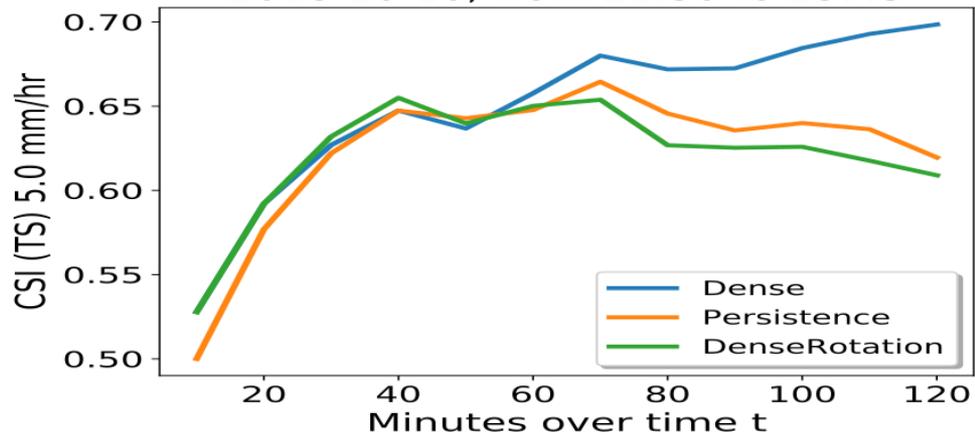
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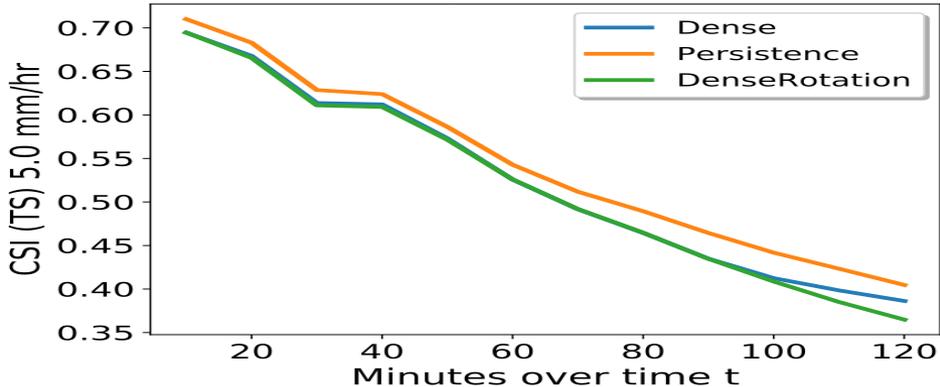
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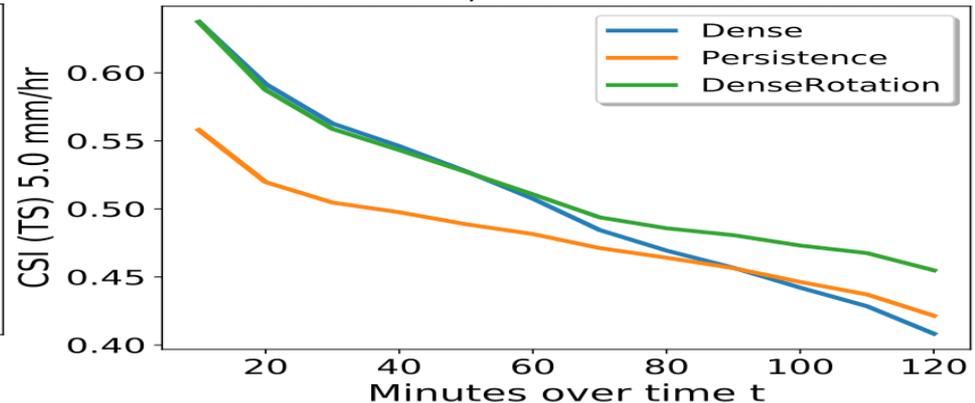
2018-10-10, from 21:30 to 23:45



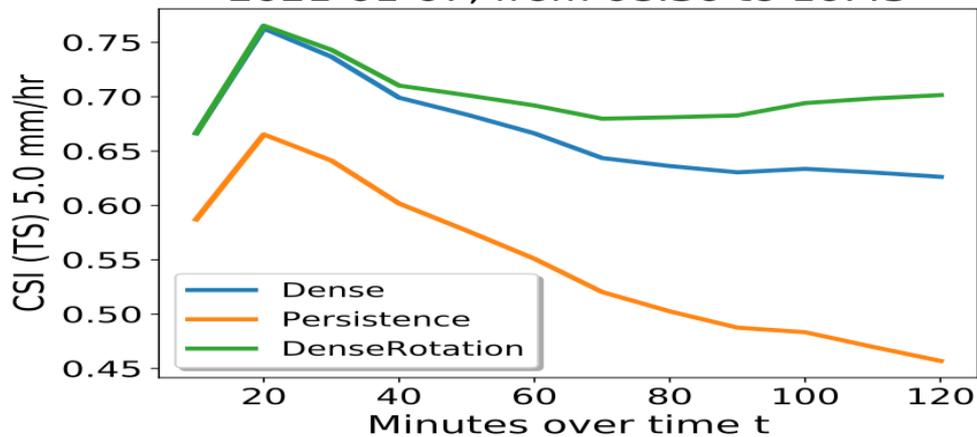
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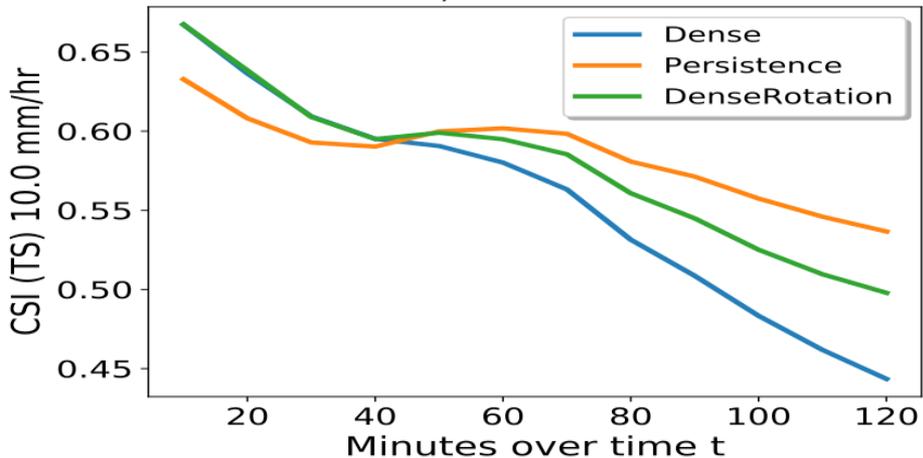
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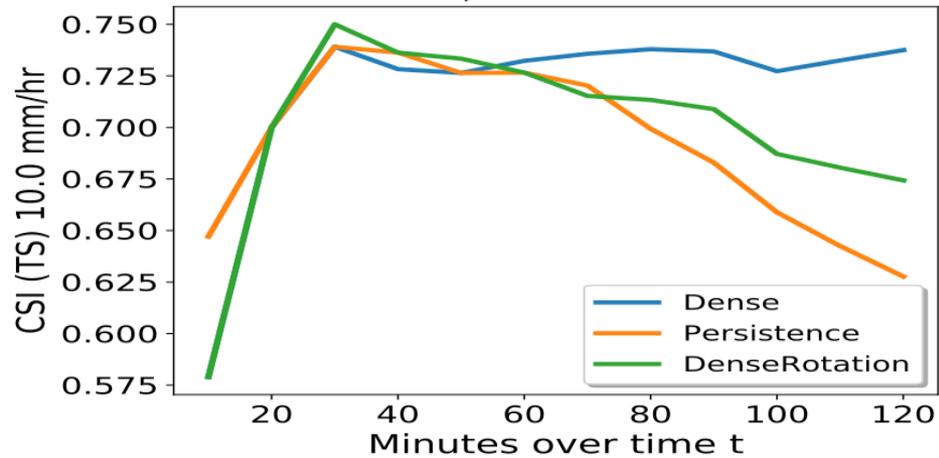
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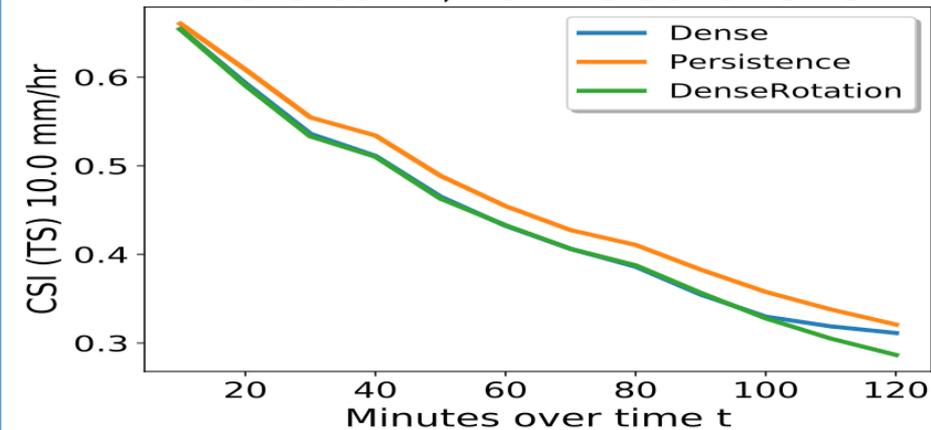
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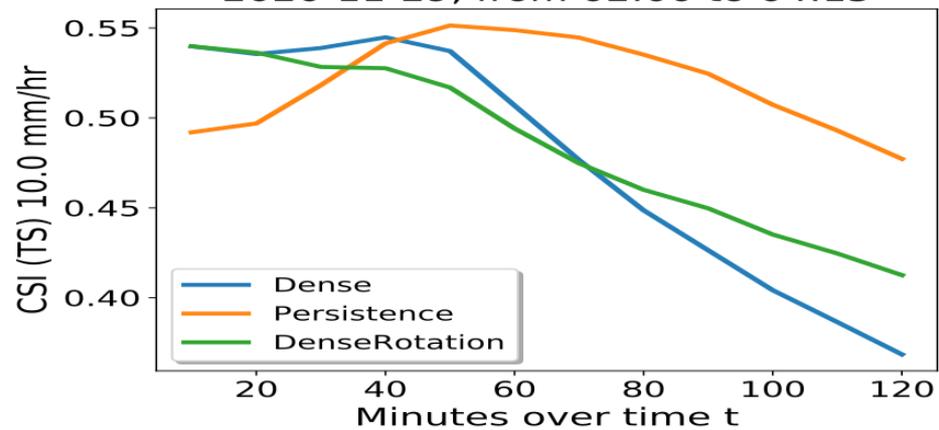
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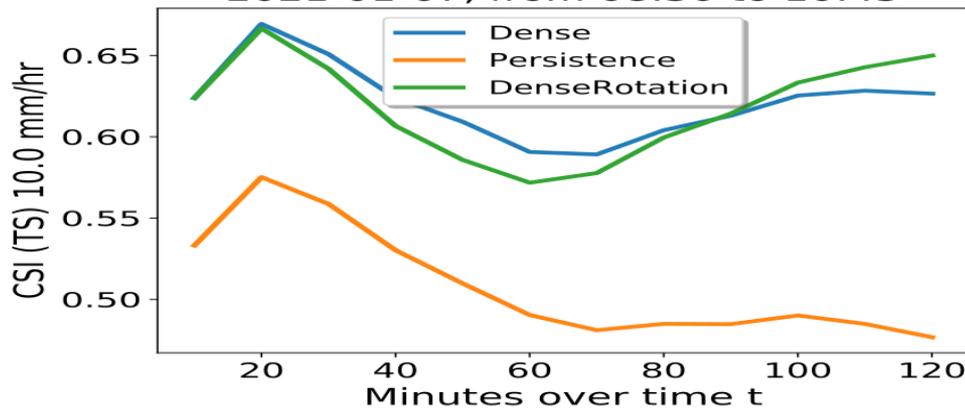
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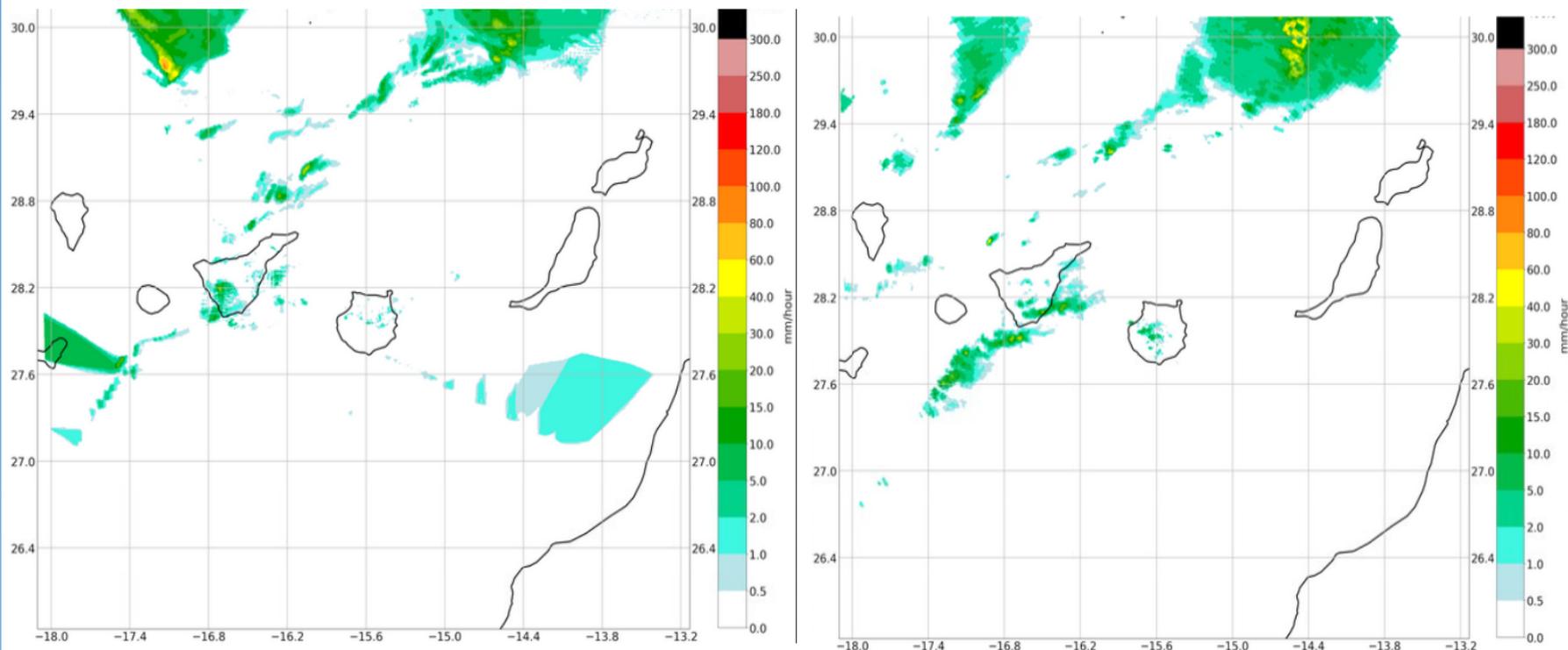


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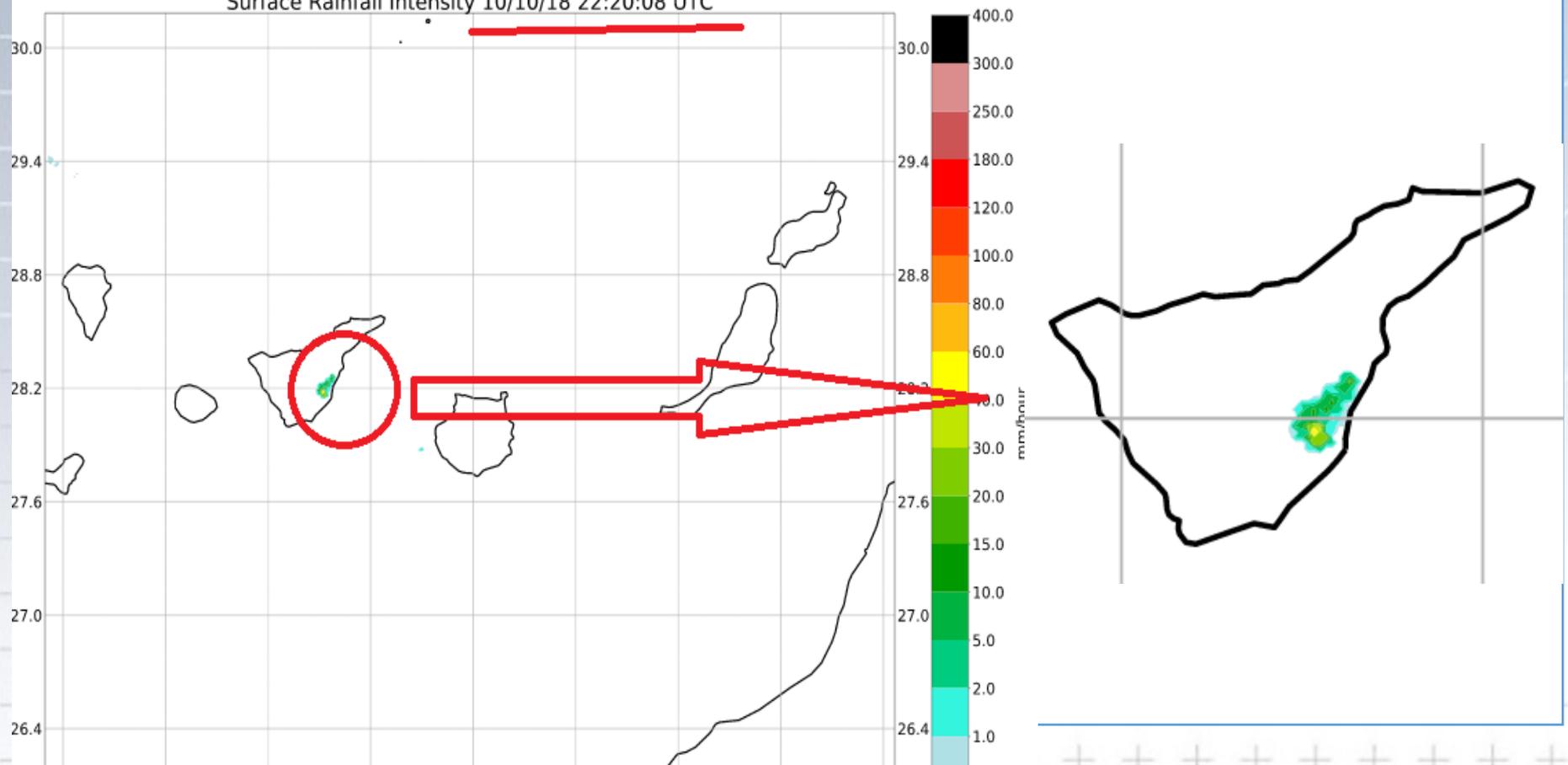
- To give an example: SRI from *rainymotion* at 09:10 UTC, 2021-01-07, with SRI from 07:50 and 08:00 (left); observed SRI at 09:10 (right).



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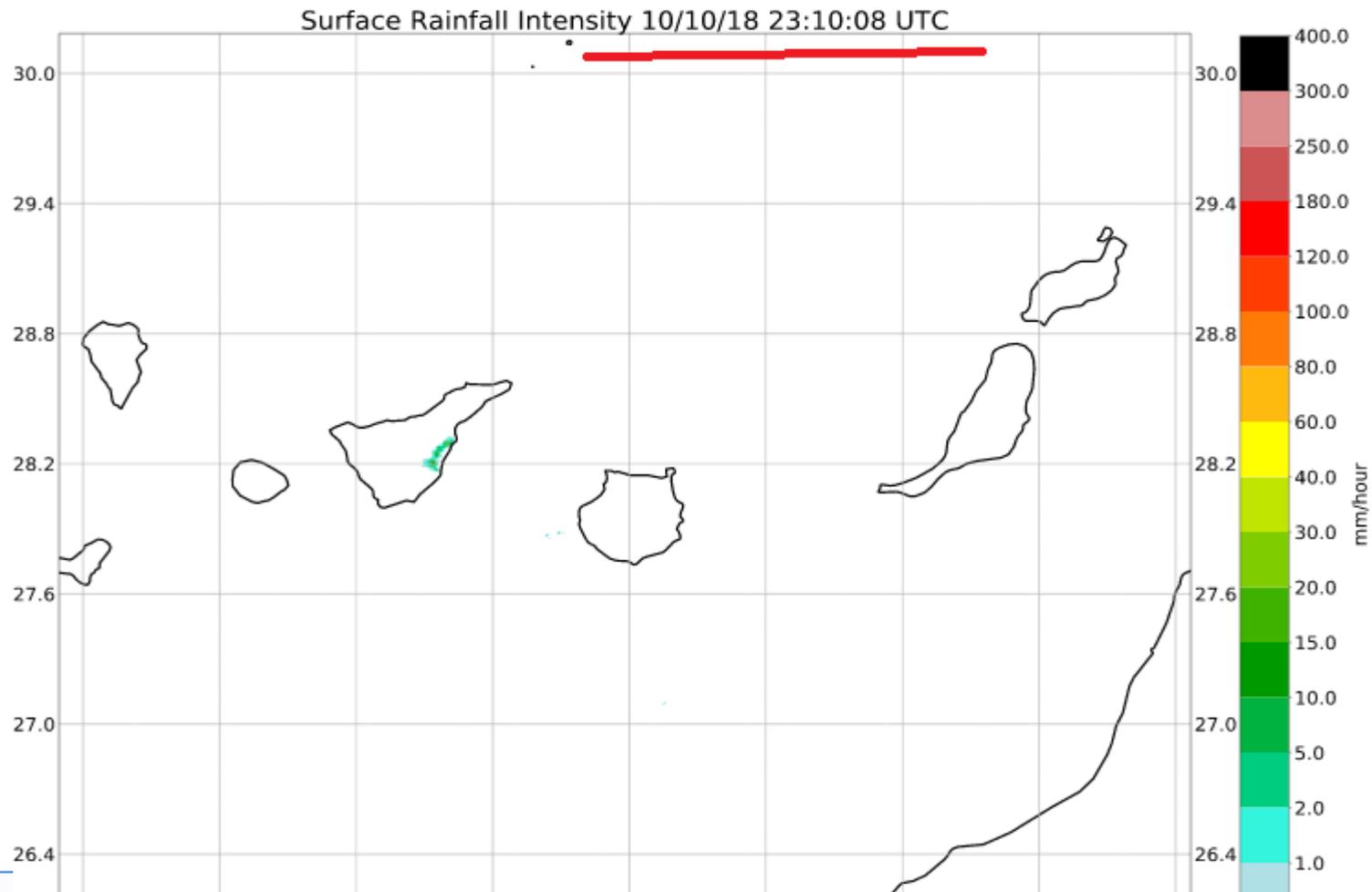
- **What did it happen when persistence was the best choice?** Observations at 2018-10-10. It seems we were under very localized and static conditions.

Surface Rainfall Intensity 10/10/18 22:20:08 UTC



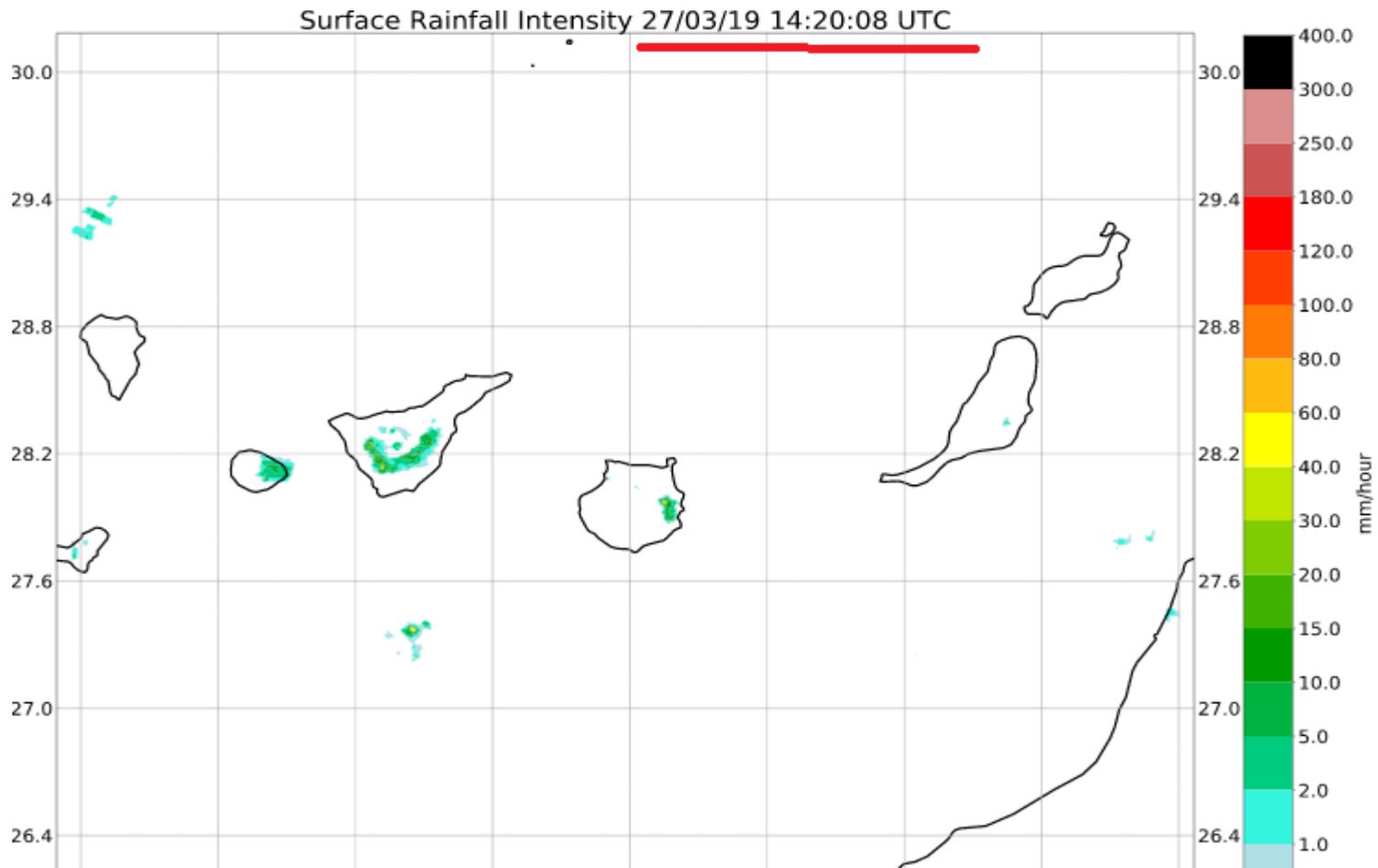
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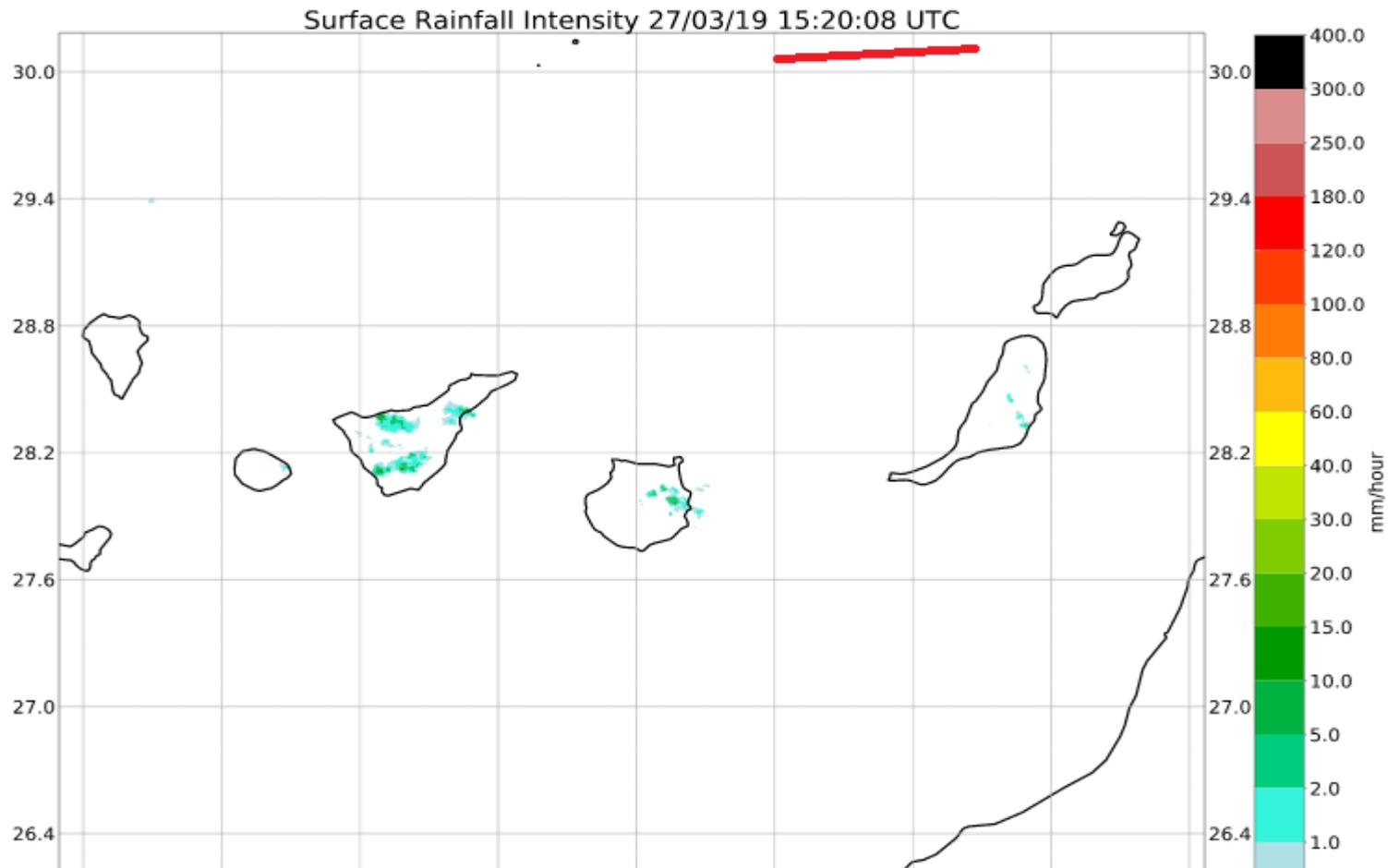
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- **What did it happen when persistence was the best choice?** It seems we were under *very* localized, *very* intense precipitation. Observations event 2019-03-27:



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- **What did it happen when persistence was the best choice?** It seems we were under *very* localized, *very* intense precipitation. Observations event 2019-03-27:



- CONCLUSIONS AND FUTURE WORK:

- Optical flow with *rainymotion* is a good nowcasting system in the complexities of the Canary Islands. Possibly a nice choice for a cheap and fast nowcasting.
- It seems that DenseRotation is slightly better than Dense (different to G. Ayzel's original paper, probably due to the complex orography of the islands).
- Persistence seems to be better for very local and/or static precipitation systems. But errors with Dense do not seem significant.
- Radar limitations in the Canary Islands: SRI each 10 min, 1778 meter above mean sea level.
- Future work: more advanced method with neural networks (RainNet...)

References:

P. 3: By Iven Gummelt - Own work, CC BY-SA 3.0, <https://commons.wikimedia.org/w/index.php?curid=4040284>

By Jens Steckert - Own work, CC BY-SA 3.0, <https://commons.wikimedia.org/w/index.php?curid=765158>

By Daniel Gaínza (Tenerife) - Own work, CC BY-SA 4.0, <https://commons.wikimedia.org/w/index.php?curid=4555690>

P. 6: Ayzel, G., Heistermann, M., and Winterrath, T.: Optical flow models as an open benchmark for radar-based precipitation nowcasting (rainymotion v0.1), Geosci. Model Dev., 12, 1387–1402, <https://doi.org/10.5194/gmd-12-1387-2019>, 2019.

P. 7: Optical flow: https://en.wikipedia.org/wiki/Optical_flow. Lucas-Kanade: https://en.wikipedia.org/wiki/Lucas%E2%80%93Kanade_method

P. 8: Kroeger T., Timofte R., Dai D., Van Gool L. (2016) Fast Optical Flow Using Dense Inverse Search. In: Leibe B., Matas J., Sebe N., Welling M. (eds) Computer Vision – ECCV 2016. ECCV 2016. Lecture Notes in Computer Science, vol 9908. Springer, Cham. https://doi.org/10.1007/978-3-319-46493-0_29

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