

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

David Quintero Plaza, AEMET, Spain. EWGLAM-SRNWP Meeting, September  
30<sup>th</sup>, 2021

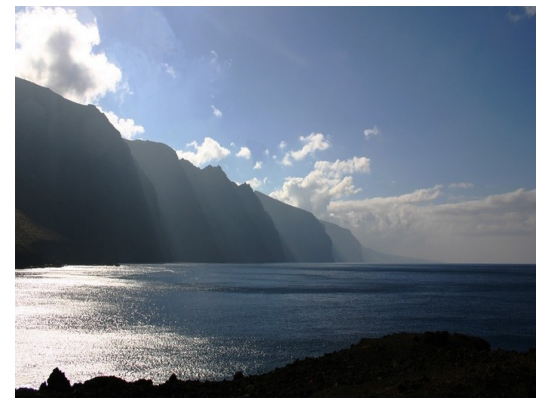


## Index.

- 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”.

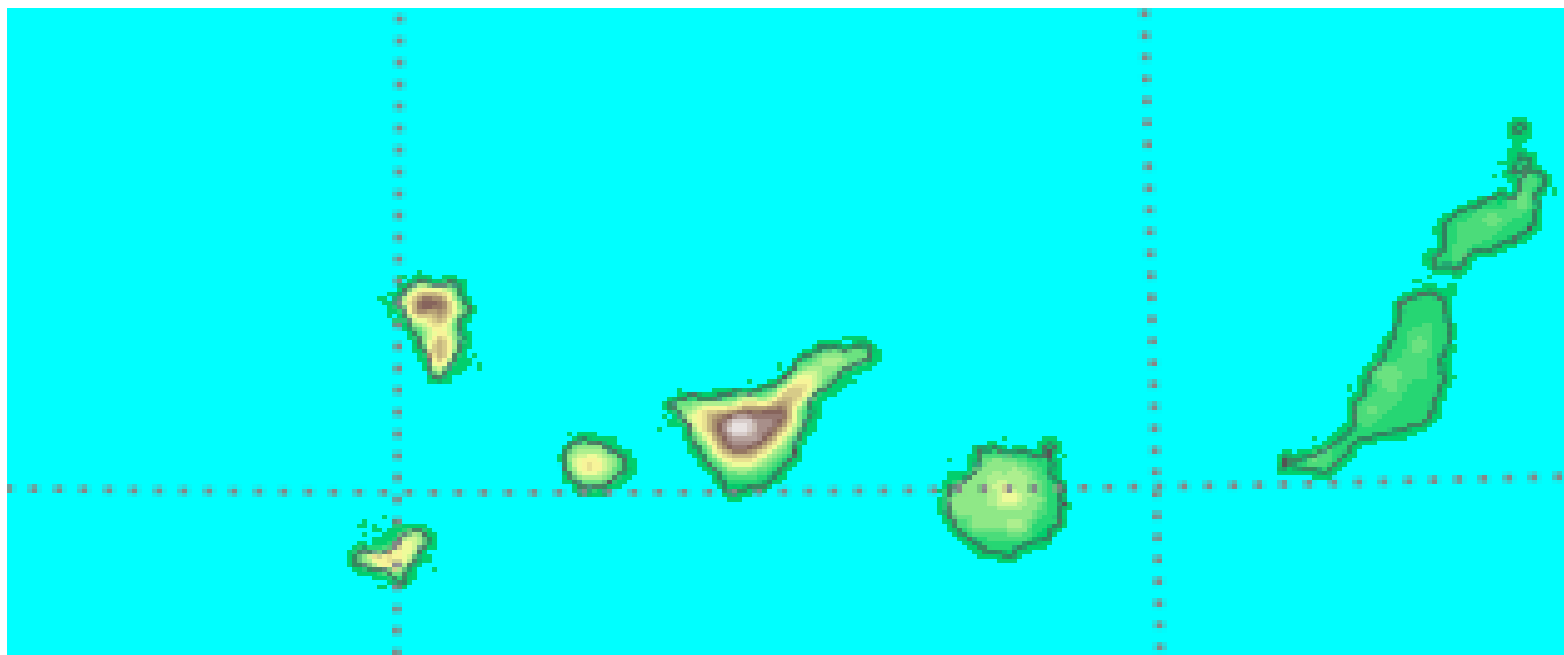
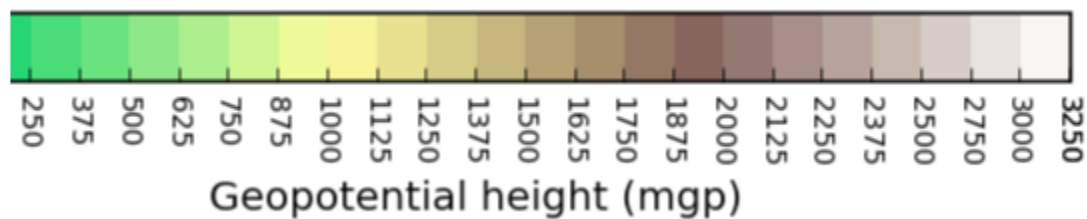


- Even High Resolution Models (Harmonie & gSREPS at 2.5 Km) still show limitations for the islands (not to mention models with less resolution).



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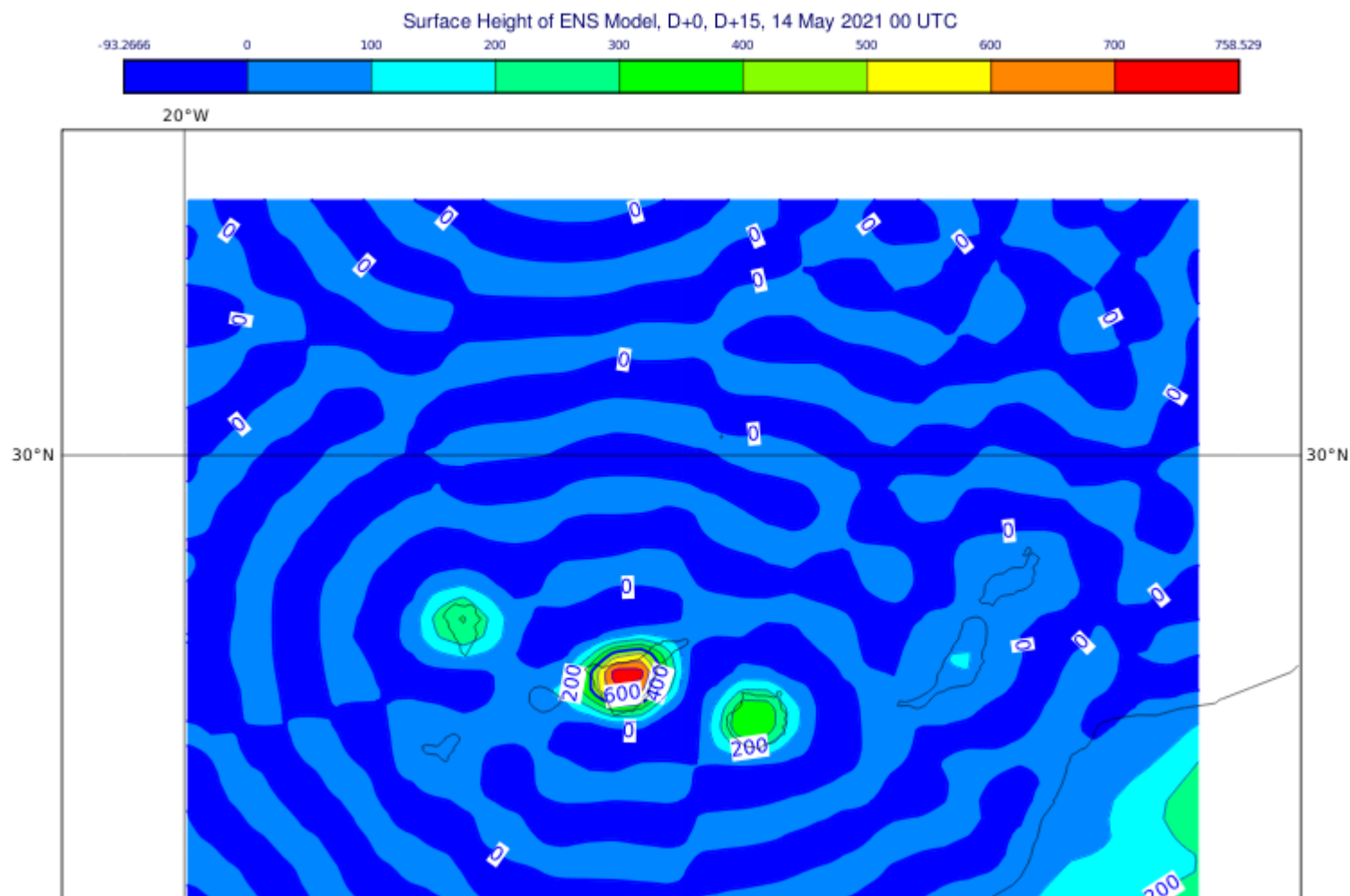
- Below: gSREPS orography.





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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.

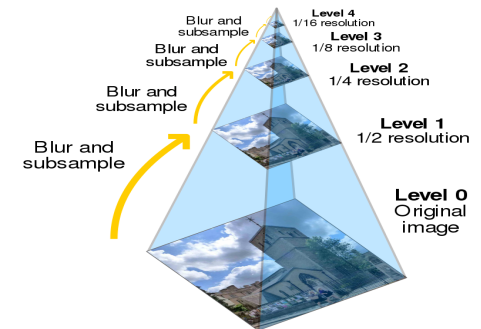
$$\begin{aligned} I_x(q_1)V_x + I_y(q_1)V_y &= -I_t(q_1) \\ I_x(q_2)V_x + I_y(q_2)V_y &= -I_t(q_2) \\ \vdots \\ I_x(q_n)V_x + I_y(q_n)V_y &= -I_t(q_n) \end{aligned} \quad \begin{bmatrix} V_x \\ V_y \end{bmatrix} = \begin{bmatrix} \sum_i I_x(q_i)^2 & \sum_i I_x(q_i)I_y(q_i) \\ \sum_i I_y(q_i)I_x(q_i)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



## - 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.



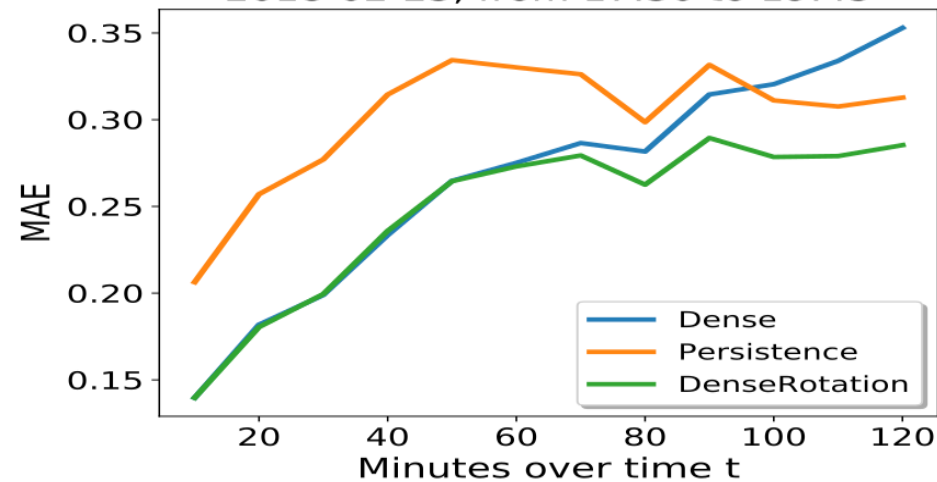
- **Our results:** 5 heavy rainfall situations under study. Scores: CSI (accuracy; higher => better) and MAE (error; lower => better).

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

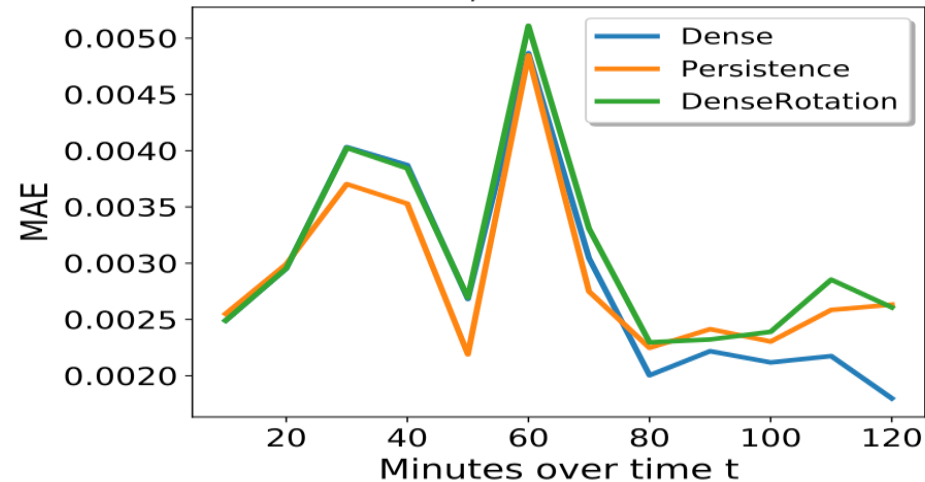
$MAE = (1/N) \sum |obs - fcst|$



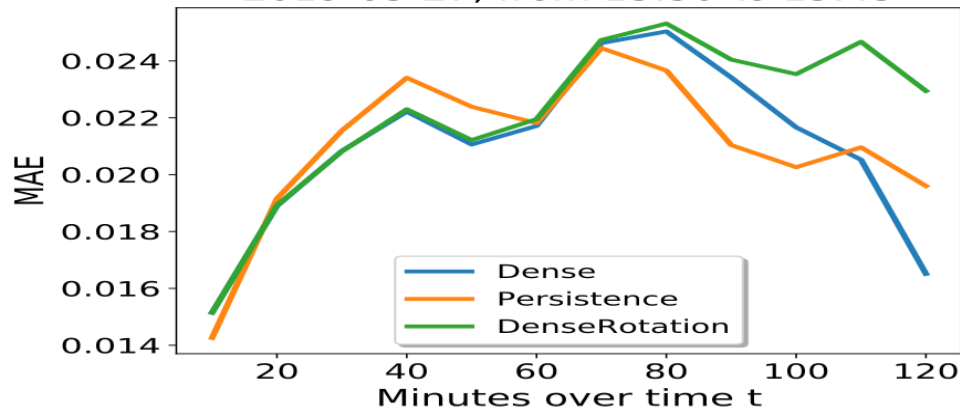
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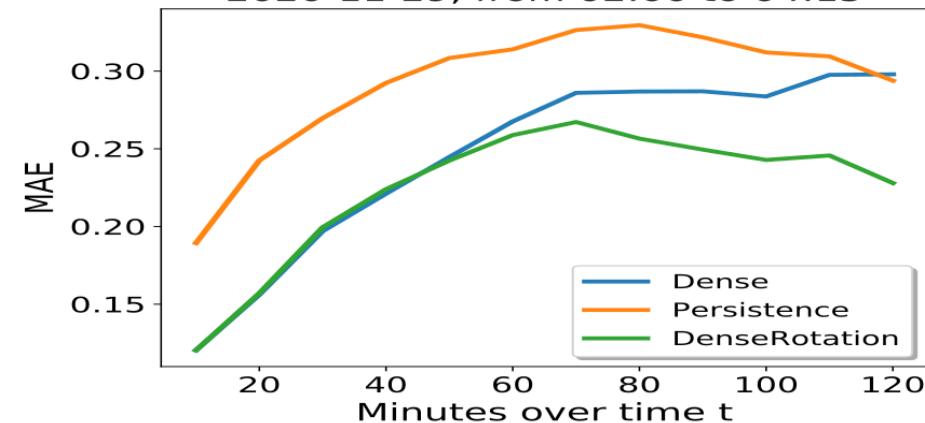
2018-10-10, from 21:30 to 23:45



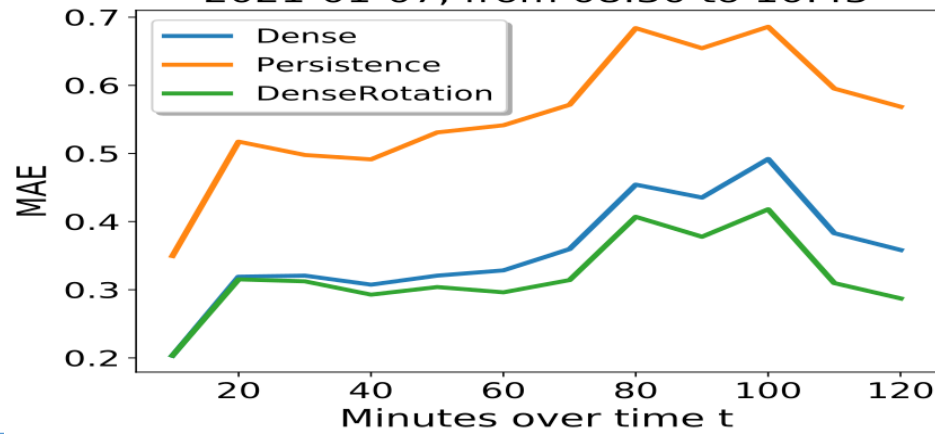
2019-03-27, from 13:30 to 15:45



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

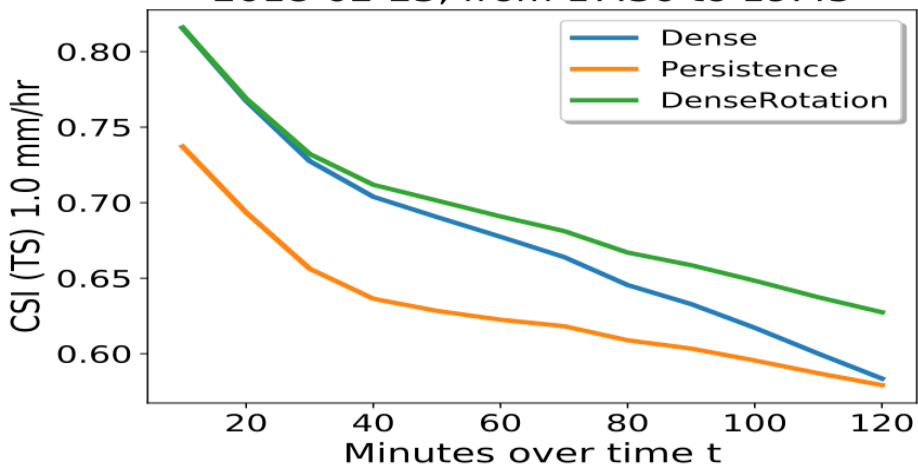


2021-01-07, from 08:30 to 10:45

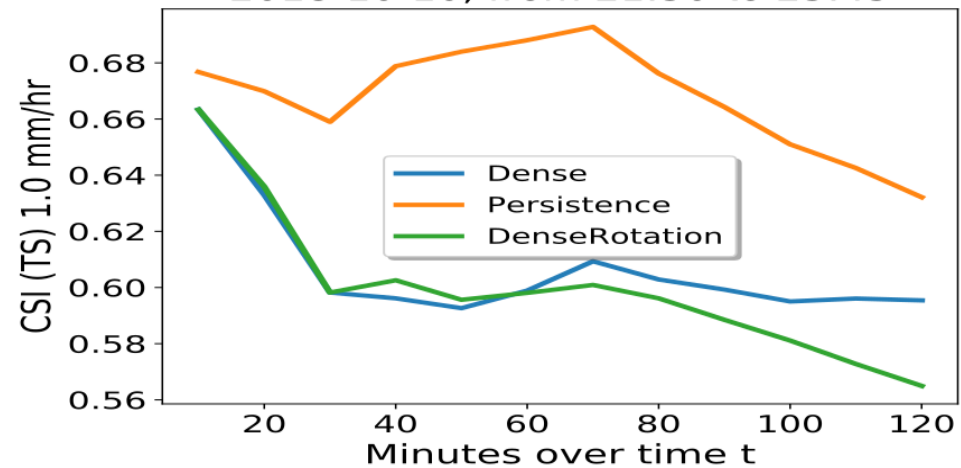




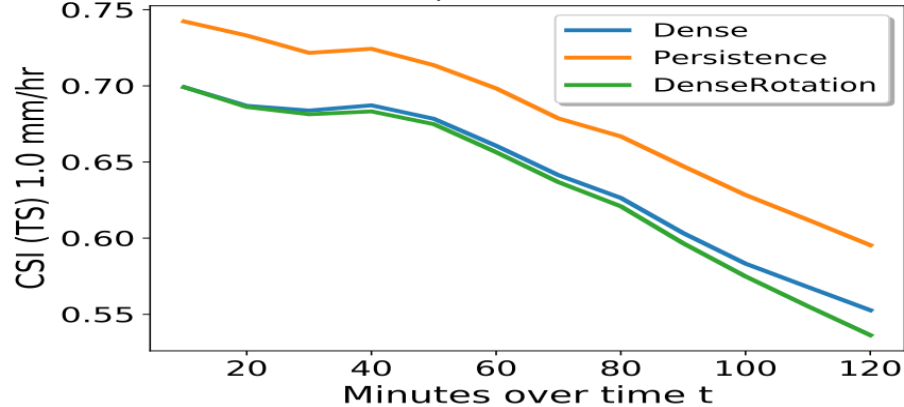
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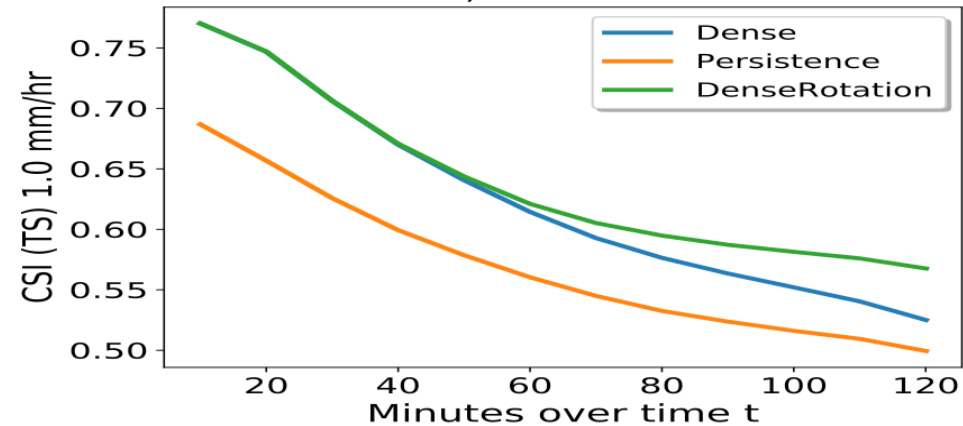
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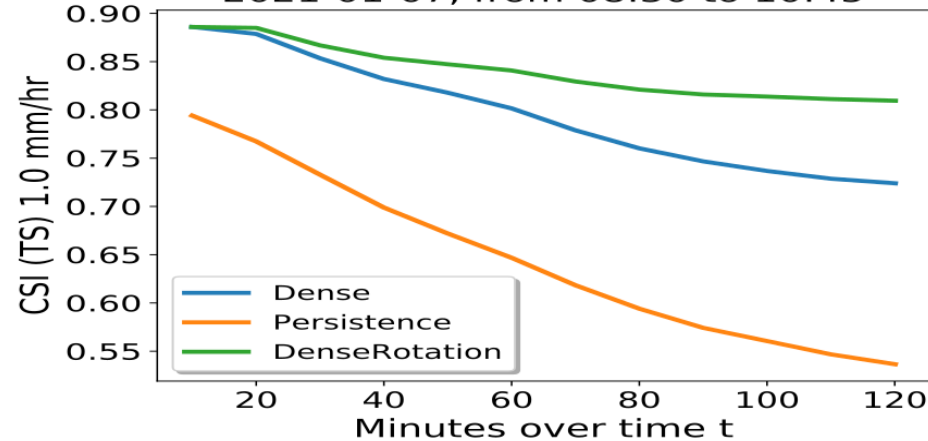
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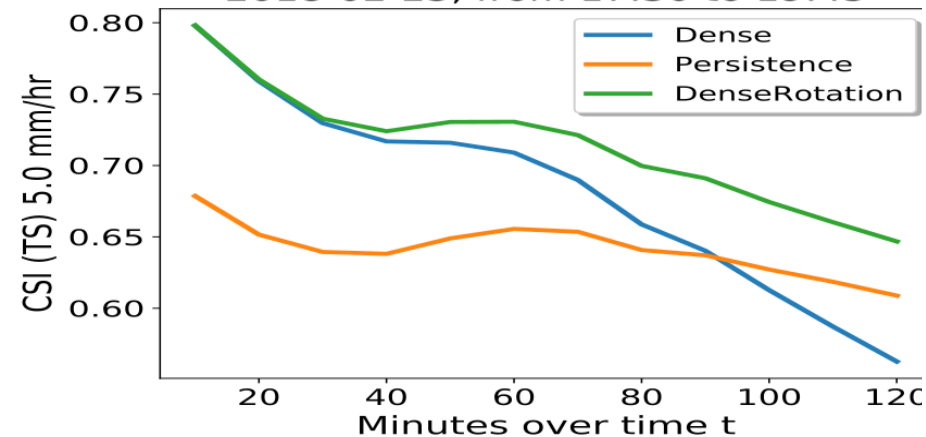


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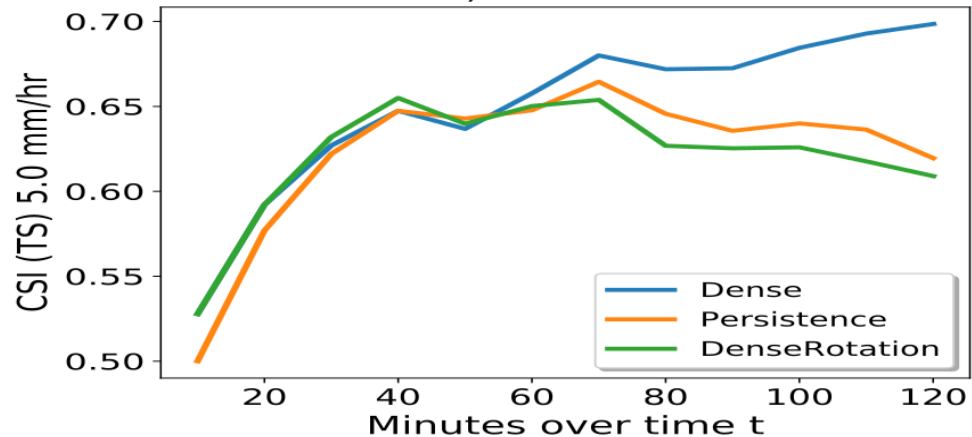




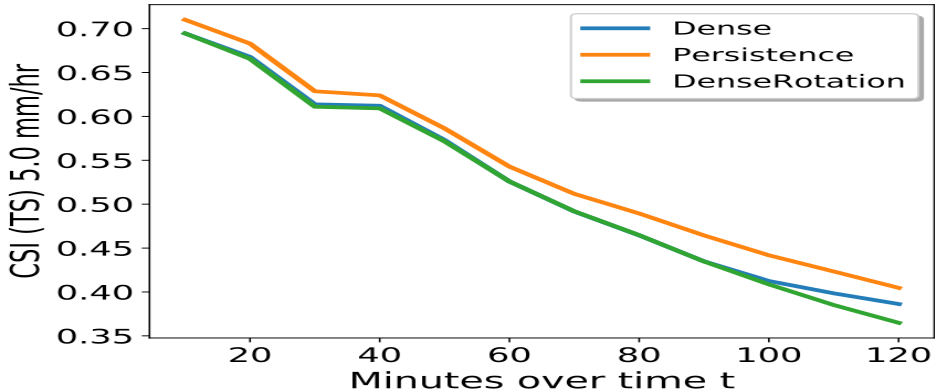
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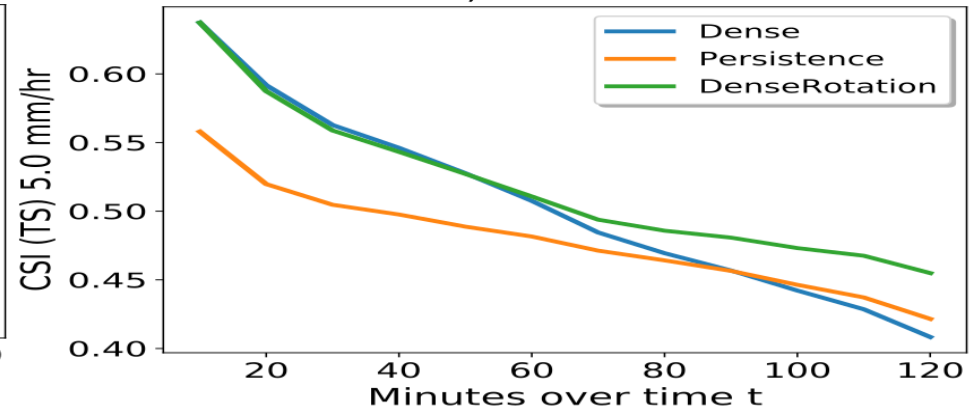
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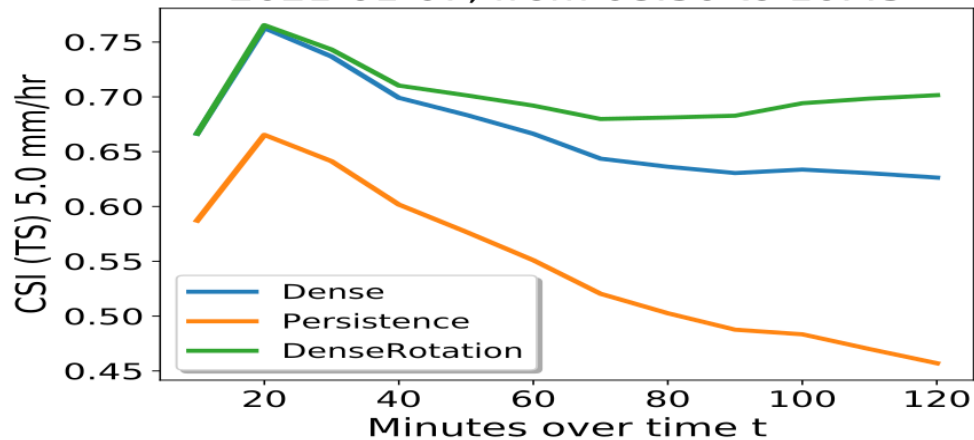
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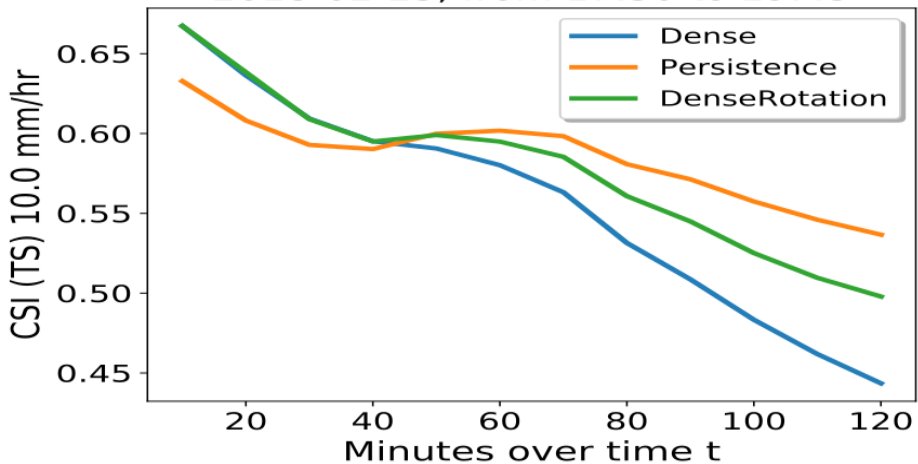


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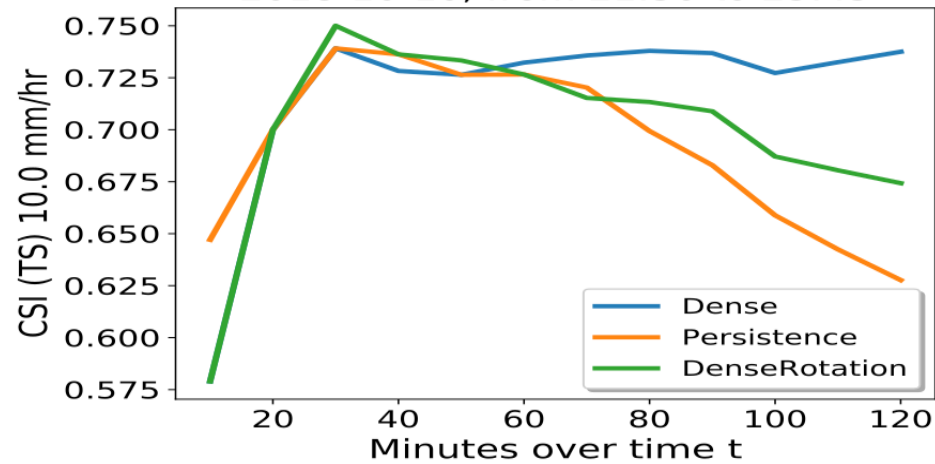




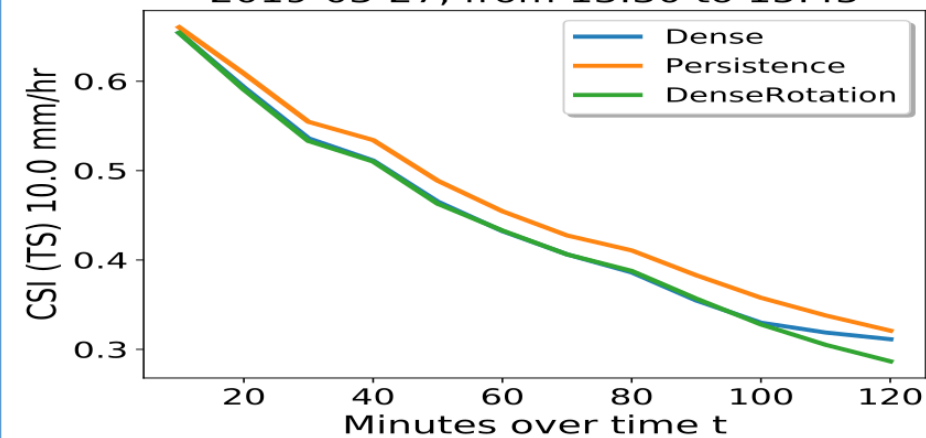
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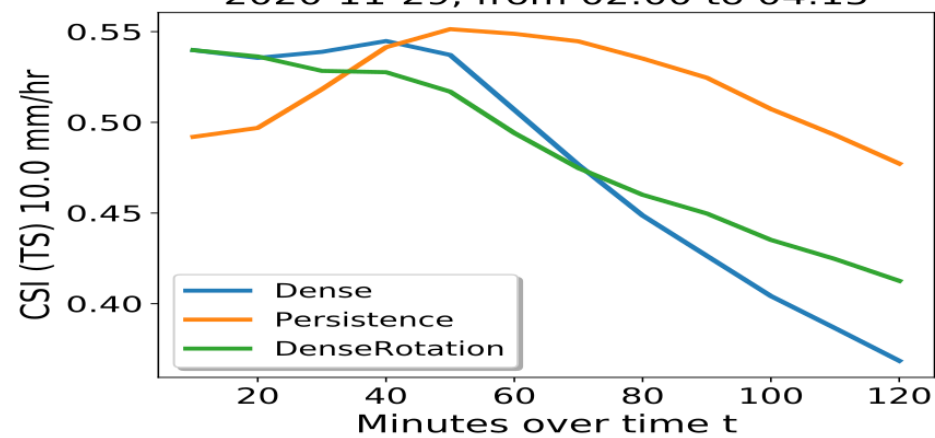
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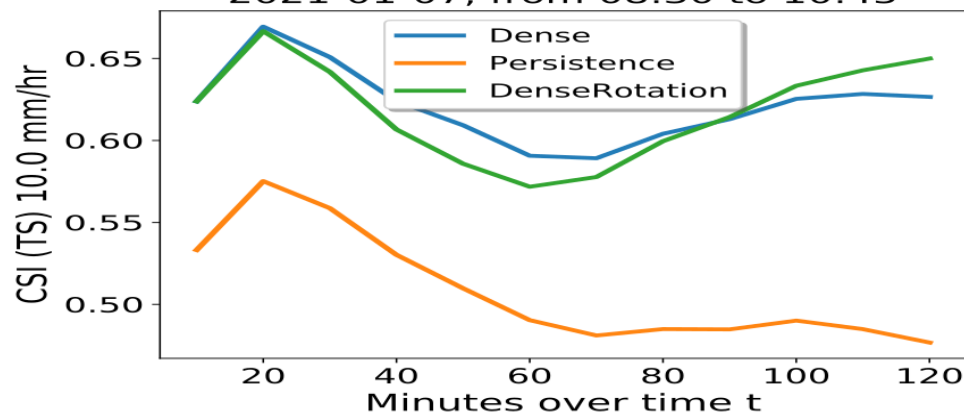
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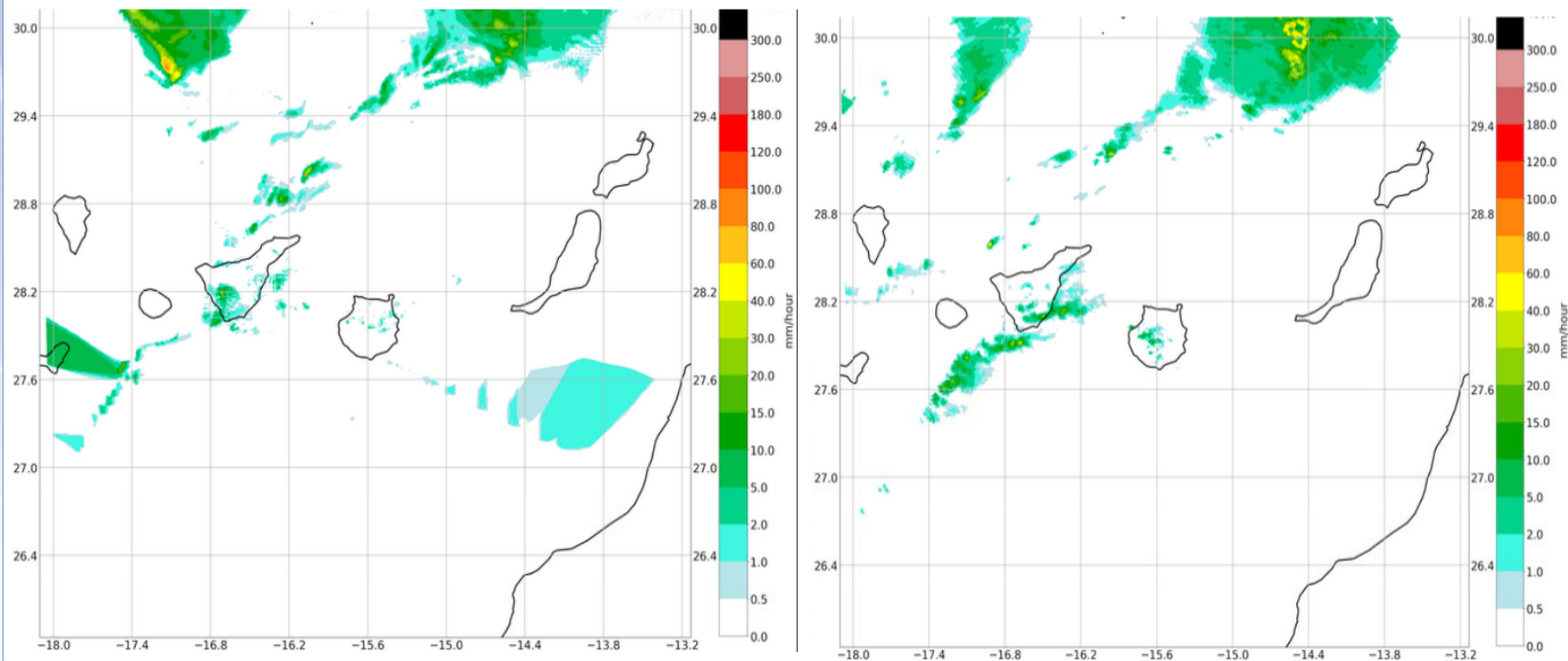


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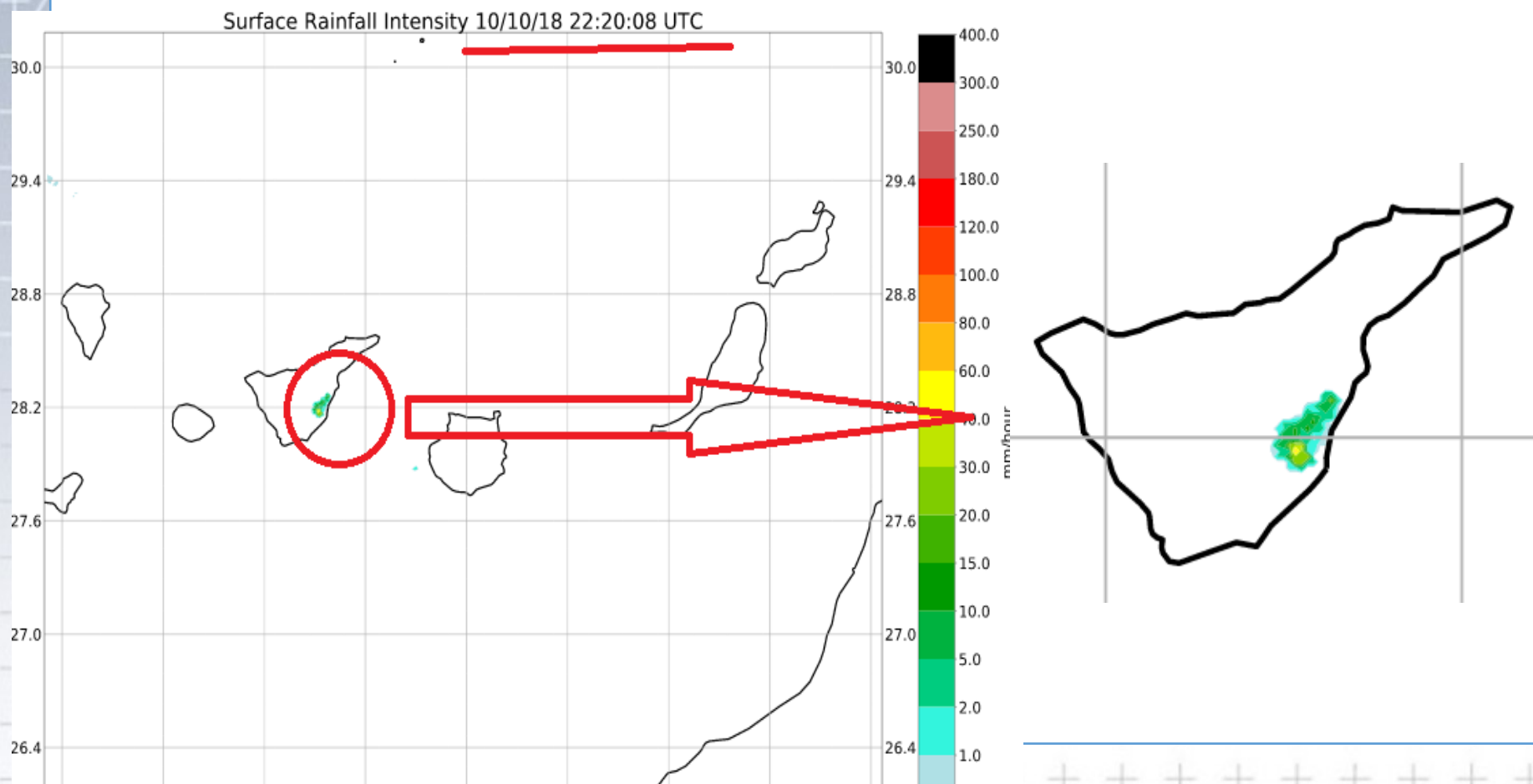
- **To give an example:** SRI from *rainymotion* at 09:10 UTC, 2021-01-07, with SRI from 07:50 and 08:00 (left); observed SRI at 09:10 (right).





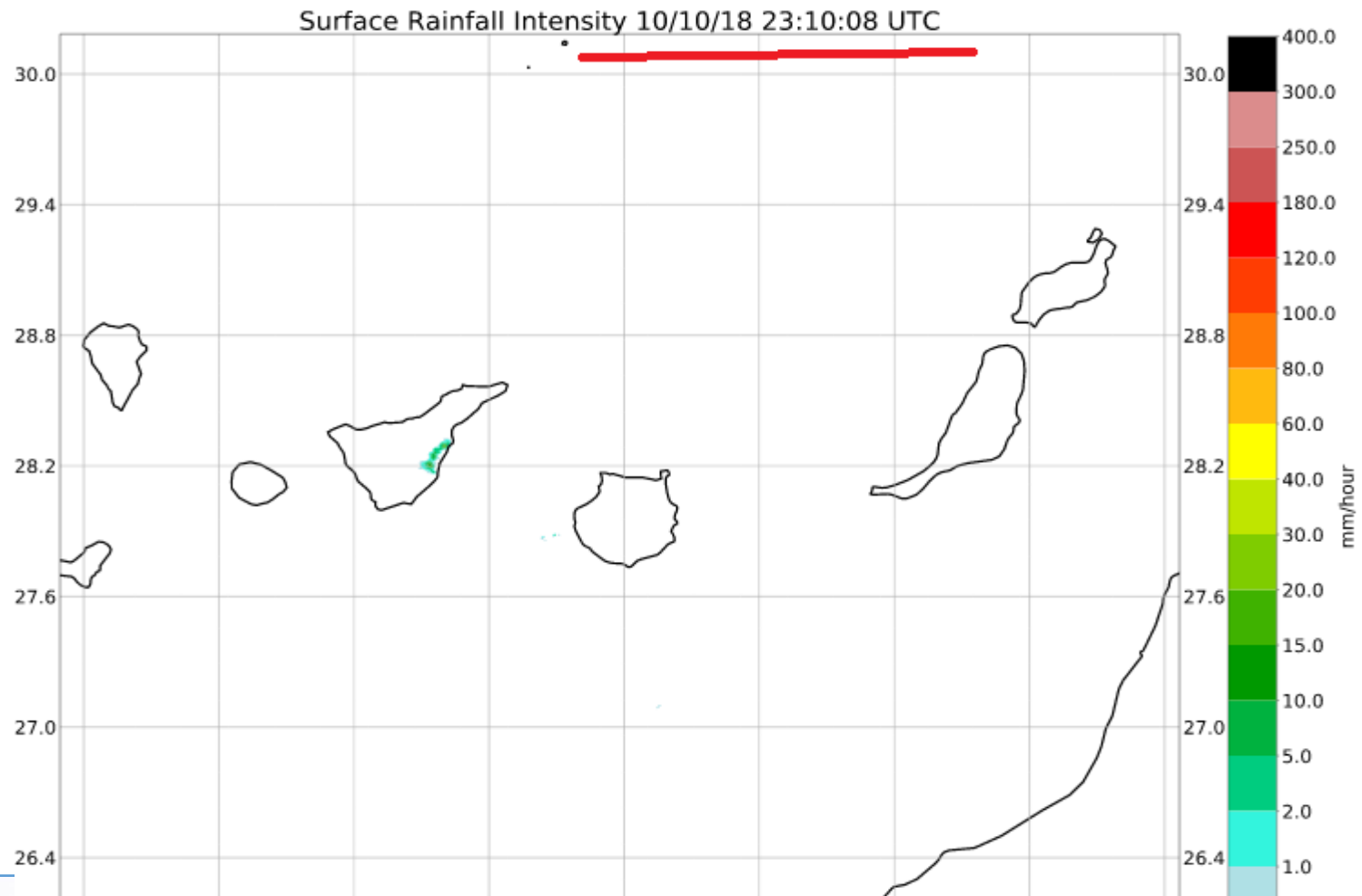
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- **What did it happen when persistence was the best choice?** Observations at 2018-10-10. It seems we were under very localized and static conditions.





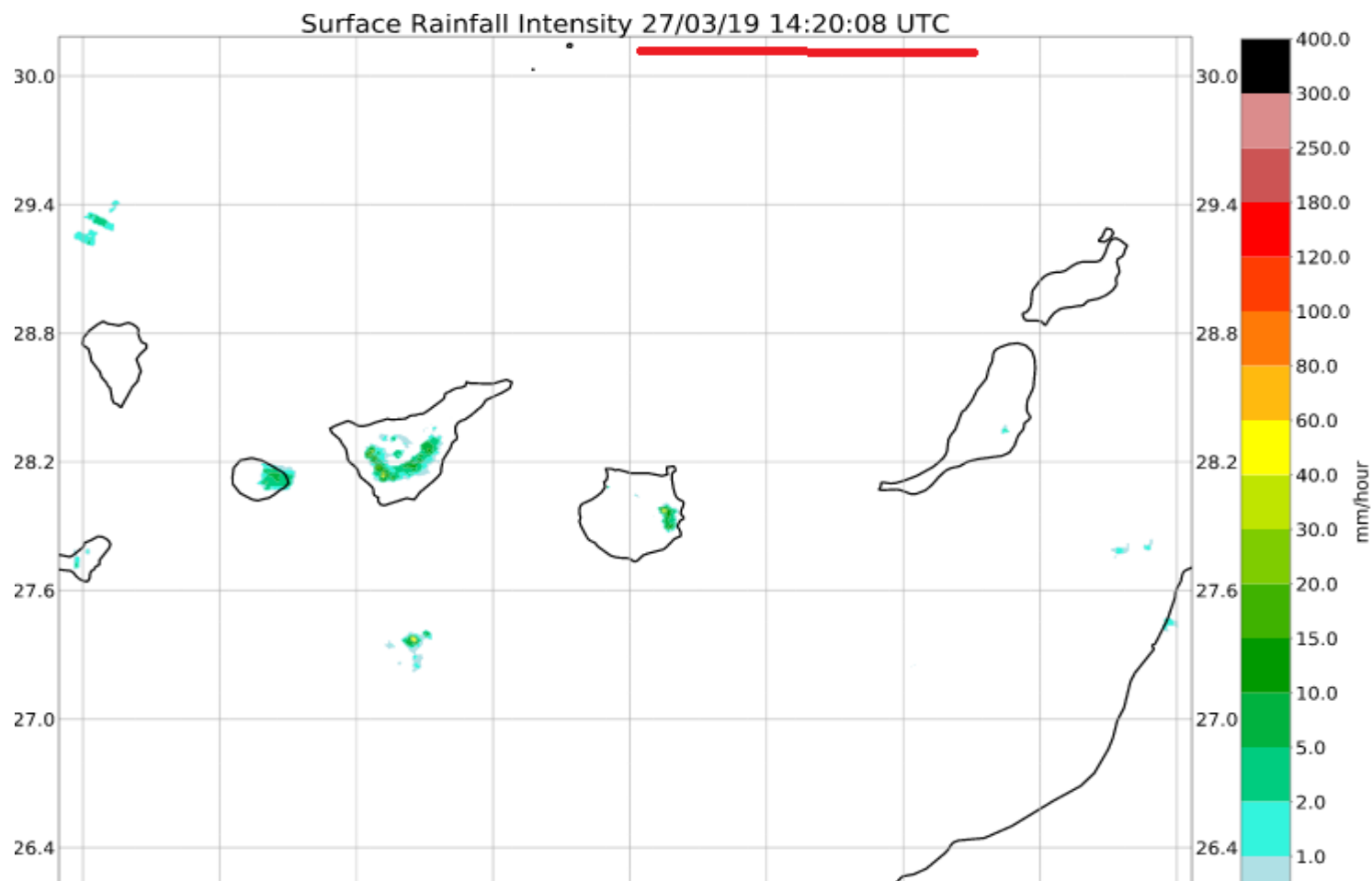
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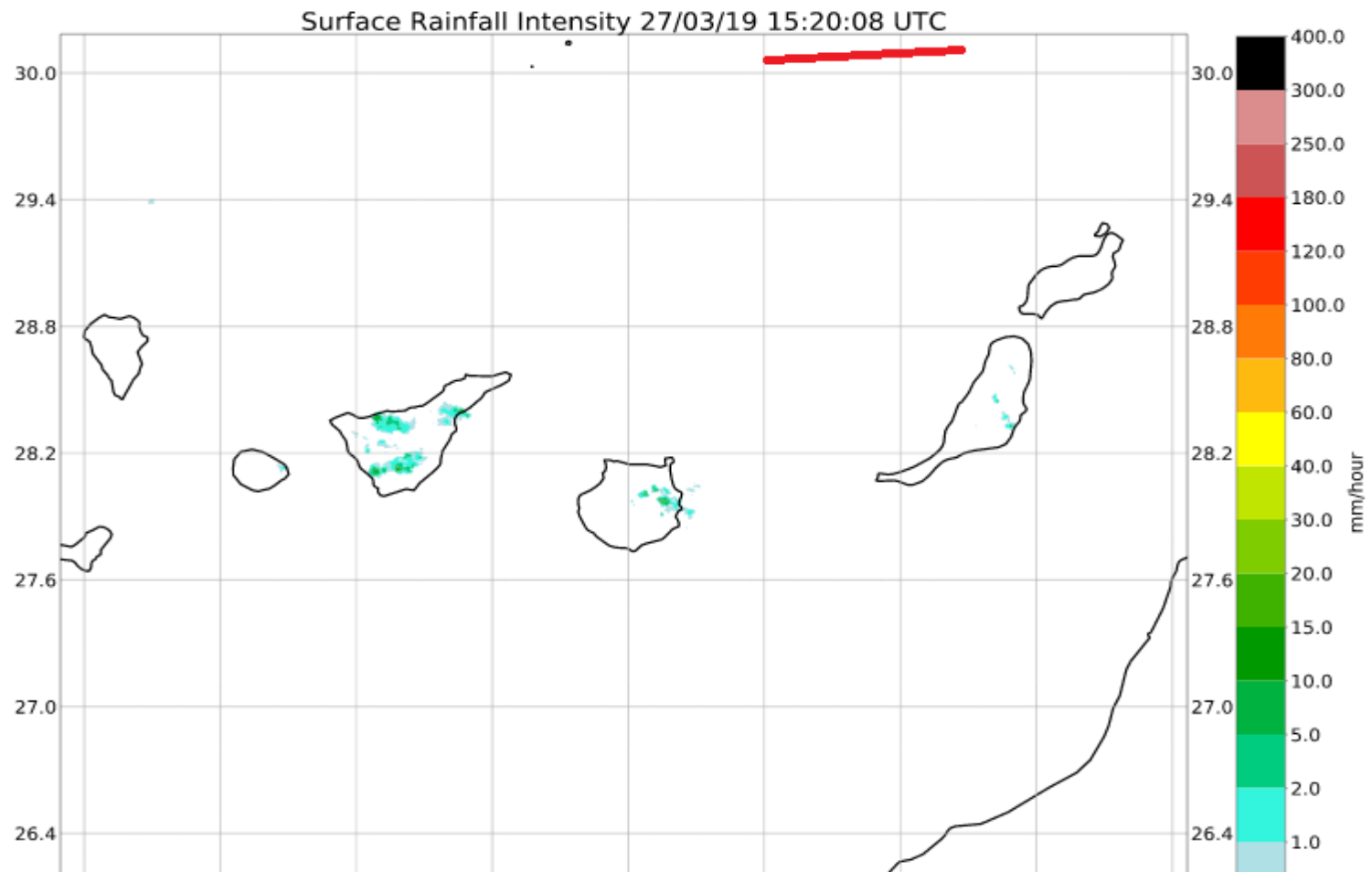
- **What did it happen when persistence was the best choice?** It seems we were under *very* localized, very intense precipitation. Observations event 2019-03-27:





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## - 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 significative.
- 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](https://en.wikipedia.org/wiki/Optical_flow). Lucas-Kanade: [https://en.wikipedia.org/wiki/Lucas%E2%80%93Kanade\\_method](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](https://doi.org/10.1007/978-3-319-46493-0_29)

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