

Artificial neural network models applied to olive tree phenology in Italy reveal daily insolation control of budbreak

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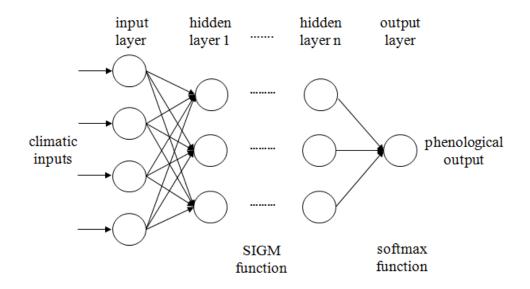


Artificial neural networks for phenology

A complementary tool to mechanistic models?

- ✓ reproduce non-linear relationships between climatic cues and phenological transitions
- ✓ include multiple climatic cues as they control different phenophase transitions
- √ account for different cultivars
- √ several applications available

Feed-forward neural network with stochastic gradient descent



Modified softmax:

$$p_i = \frac{e^{(x_i - \max(x))}}{\sum_i e^{(x_i - \max(x))}} \cdot 365$$

SIGM:

$$y_i = \frac{1}{1 + e^{-x_i}}$$



Olive (Olea europaea L.)

An under-studied, evergreen species

- ✓ A traditional element of the Mediterranean landscape, diet and rural socio-economics
- √ delivering multiple ecosystem services in arid areas
- ✓ under threat in Italy since the outbreak of the bacterium Xylella fastidiosa spp. pauca in Apulia region

Vegetative phenology comparatively less studied than reproductive phenology





The dataset

An Italian-wide datset, comprising

- √ 7 locations in the main olive-grove areas
- √ 4 years of observations in 1997-2000
- √ 16 cultivars
- ✓ multiple phenophse transitions during the year
- ✓ ~132 year-long data sets

The dataset does not account for the impact on olive phenology of the last 20-year climate change



Site ID	Location	Latitude	Longitude	Observation years
1	Montepaldi (Firenze)	43°40'02"	11°8'53"	1997-1998-1999
2	Villasor (Cagliari)	39° 25' 12"	8° 47' 24"	1997-1998-1999-2000
3	Valenzano (Bari)	41°02'11"	16°51'29"	1997-1998-1999-2000
4	Torre Allegra (Catania)	37°25'10"	15°00'36"	1997-1998-1999
5	Belice Mare (Trapani)	37°36'03"	12°51'20"	1997-1998-1999
6	Rende (Cosenza)	39°21'55"	16°13'50"	1997-1998-1999- 2017*-2018*
7	S. Apollinare (Perugia)	43° 00' 13"	12° 18′ 08″	1997-1998-1999-2000

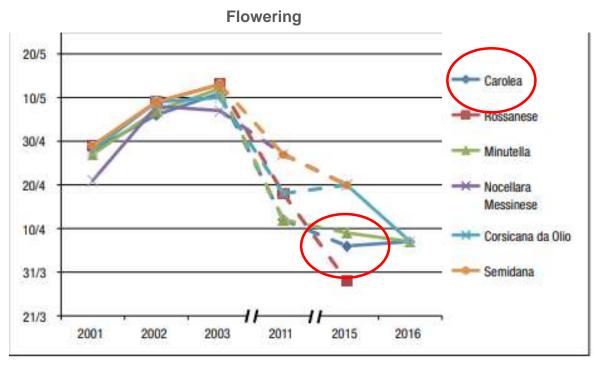


Model training and validation

- ✓ A k-fold cross validation was used to optimize the model (number of layers, activation function, convergence method) and hyperparameters
- ✓ Model performance is dependent on number of available observations

Overall RMSE of ~ 4 d at both training and validation when joining together all the cultivars (comparable with the state-of-the-art of mechanistic models).

However, phenological response to climate change is cultivar-dependent



Salimonti et al., 2017, n. 6/2017 OlivoeOlio• 27

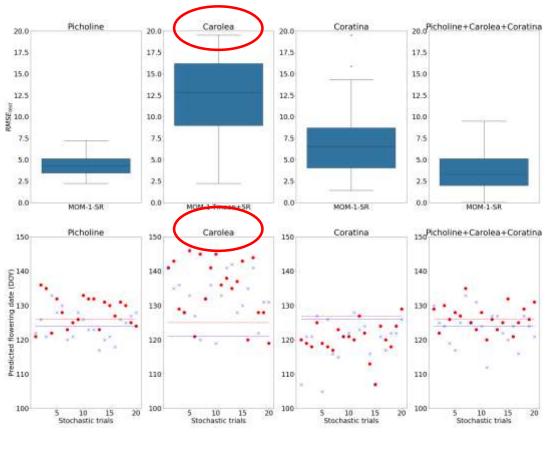


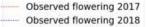
Olive cultivars matter for climate change impact investigations

- The model was tested on independent flowering observations collected in 2017 and 2018.
- The model trained in the period 1997-2000 showed some cultivar-specific overprediction.

In particular, Carolea c. flowering was overpredicted

Carolea flowering has been anticipated after the 1997-2000 training period, by roughly the same amount of days observed by Salimonti et al. (2017).







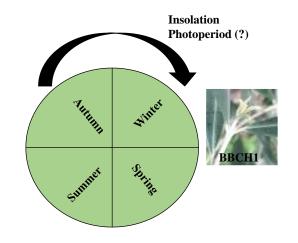


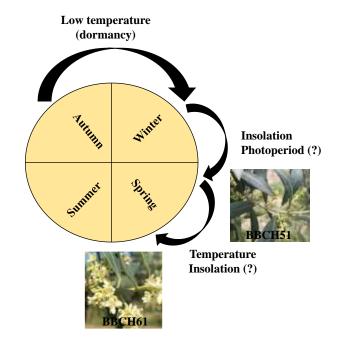
Daily insolation as budbreak predictor?

Daily mean temperature was replaced and/or complemented with:

- ✓ Daily minimum temperature
- ✓ Daily maximum temperature
- ✓ Daily insolation
- √ Photoperiod

Daily insolation consistently improved budbreak prediction for all the cultivars, while photoperiod effect was not consistent.







Budbreak transition

Inflorescence and flowering transitions



Way forward

- √ finalizing the investigation of climatic cues
- ✓ widening the dataset
- √ coupling olive phenology with pathogen phenology



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