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Article

A modelling strategy to improve cacao quality and productivity

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Abstract: Crop modelling can support agronomical decisions of crop production under a range of scenarios improving competitiveness. Cocoa production systems in Latin America has a high importance over social and economic development, facing the fight against hunger and poverty. Although Colombian cocoa has the potential to be in the high value markets for fine flavour, it is still not widely produced as the lack of adoptions of technologies by the traditional farmers. They empirically harvest after 5 or 6 months after flowering date. However, cocoa fruits development can be considered as the result of a number of physiological and morphological processes that can be described by mathematical relationships even under uncontrolled environments. Thus, we parametrized the SIMPLE crop model [1] to predict the best time for harvest cocoa fruits in Colombia. The results showed an RRMSE of 7.2% for the yield prediction, while the simulated harvest date varied between +/- 2 to 20 days depending on the temperature variations of the year between regions. This crop model application contributed to understand and predict the phenology of cacao fruits the varieties ICS95 y CCN51, which is key to produce high quality cacao beans. The aim of this study was developed a practical tool for farmers on cocoa fields to predict the best moment to harvest, using easily available data for crop modelling such as flowering date and weather variables (solar radiation, rain, and temperature).

Keywords: ICS95; CCN51; thermal time, flowering date

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1. Introduction

Cocoa (*Theobroma cacao* L.) is an important worldwide perennial tropical crop endemic to the South American rainforests [2–5]. Cacao plant is a member of the Malvaceae (formerly Sterculiaceae) botanical family such as cotton *Gossypium hirstium* [6]. Coton has been is modeled in SIMPLE model [1]. Cocoa is grown for its fruits, known as cacao pods [7,8]. Only the 5% of the world cocoa yield is destined for Fine-cocoa production due to the low productivity through the traditional crop management [4]. In Colombia, cocoa is traditionally consumed as a beverage. It is one of the crops promoted by the Colombian government in the social and agricultural development programs aimed at favouring peace in post-conflict regions [5,9]. This crop is grown by approximately 52.000 families [10] and 98% of production being carried out by small and medium-sized producers [11,12]. Colombia registered an increase of 3.750 tons in production in 2020 compared to the previous year [13].

Although Colombian cocoa has the potential to be in the high value markets for fine flavour [12], it is still not widely produced as the lack of adoptions of technologies by the traditional farmers. They empirically harvest at 5 or 6 months after flowering date, hence a mix of quality of cocoa beans can be fermented. This produce heterogeneous characteristics between each fermentation batch diminishing the quality of cocoa final product [14]. To identify the best moment to harvest is important to consider physiological responses affected by climate variables such as rain, solar radiation and wind. Thus, for cocoa in Colombia, physiological simulation models may be valuable to identify the best moment to harvest cocoa considering variable weather conditions, soil types and cultivar specifications.

Crop models represent a quantitative assumption of plant growth depending on sunlight interception efficiency values and climate data supported by a large amount of empirical and ground data. [15]. Physiological crop models have shown to be very useful tool for provided agronomical advices and improvements of the cropping systems of annual crops mainly. Recently crop modelling studies are focusing on perennial crops production [1,2,16,17]. However, the information reported is be less than for annual crops due to the lack of field data available , relatively high research costs and the difficulties of accumulated errors in long-term simulations [2]. For cacao there approaches to predict yield mainly using algorithms of machine learning [13] and just one mechanistic model simulates physiological cocoa performance "SUCROS-cocoa" [2]. This crop model calculates light interception, photosynthesis, maintenance respiration, evapotranspiration, biomass production and cocoa yield. It can be parametrised having data on cocoa physiology and morphology [2]. However, there is not specific cocoa physiology data available from small and medium-sized producers. Thus, we adapted the simple generic crop model (SIMPLE) that could be easily modified for any crop to simulate development, crop growth and yield using few parameters such as weather and cultivar specification [1].

In this paper, we present a physiological parametrization of SIMPLE crop model for cocoa to predict the best harvest time and yield production. We used the SIMPLE crop model [1] for three reasons: 1: That it is very comprehensively described in the original paper. 2: That the code was available in R for initial trials and 3: That it had already been successfully fitted to other perennial crops in south America. Overall, the model simulates crop development, growth and yield, and predict the maturation day when the fruit is ready to harvest. It includes 13 parameters (daily weather data, irrigation, and soil and key dates) to specify a crop type, with four of these for cultivar characteristics easily available from farmer. Thus, this could be used as tool for small farmers with the aim to improve the quality of cocoa to become more competitive in Fine-cocoa market.

2. Materials and Methods

2.1. Floral Phenology of Cocoa

Normally the phenological stages in cocoa are divided in two main phases: vegetative and reproductive. In the SIMPLE model we simulate the reproductive phase (fig. 1, a) described by the floral phenology from the date of inflorescence emergence (BBCH scale 5) (fig. 1,b) to predict the date of ripening of fruit and seed (BBCH 8) (fig. 1,c) [7] . In the Andean region the reproductive phase is cyclically fulfilled during two annual cycles passing by the following phases: inflorescence emergence, flowering, pollination, fruit development and harvest. Therefore, for modelling parametrization the crop cycle of cocoa as perennial plant does not start at the plantation date such as annual crops systems. Instead the start point of the cocoa crop cycle to model is the inflorescence emergence date (fig. 1,b). Consequently , the growth period of the fruit can varied from 110 to 150 daa (days after anthesis) [18] when cacao fruits reaches the physiological maturity, but it can be harvested at 170 days daa [7] for quality purposes.

Cocoa is a cauliflorous plant, which means that flowers grow on the trunk and branches. In Colombia cocoa trees usually produce flowers throughout the year. Cocoa trees produces with up to 10000 flowers per tree each year, which the 50 % do not develop into ripe fruits according to Fedecacao reports. The flower takes 30 days passing by 12 micro-stages from meristem development (stages 1to 6) to the fully developed flower (Stages 7 to 12) [19] when it is ready to be pollinated. The opening of flowers or anthesis occurs over a 12-hour period during the night and it is synchronised between the groups of mature flowers [7]. However, the live of a flower can last approximately 1 day after the opening falling form the trunk if it is unfertilised [7,20].

Subsequentially, after anthesis the fruit growths by approximately 150 days until the maturation, mucilage. Therefore, the complete maturation process of the fruit, from the pollination to fully mature fruit, takes 160- 210 days [21]. The accumulation of lipids, storage proteins and anthocyanin starts about 85 days after pollination when fruits have an active metabolism and seeds moisture content decreases up to 30% [7,22]. During this phase the quality of cocoa seeds is defined.

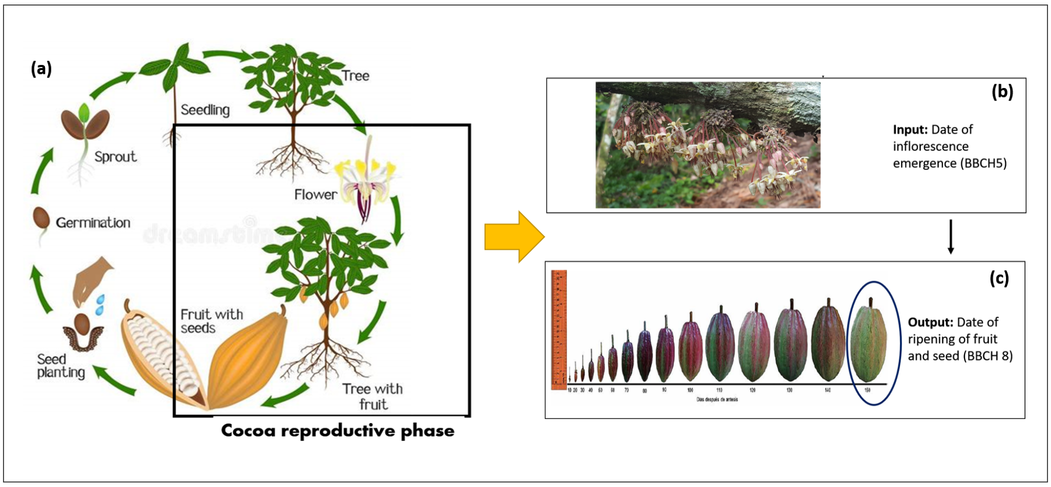

Figure 1. Phenology of cocoa in Colombia for crop modelling.

Source: Taken from from Dreamstime.com, phys.org [23] and [18]

2.2. Test site and yield production

Cacao is cultivated in 30 states out of 32 from the total Colombian territory with about 147,000 ha [24]. Cocoa fields studied were located in Saravena (Arauca), Rionegro (Santander), Cali (Valle del Cauca), Apartado (Antioquia) and Manizalez (Caldas) (Fig.2, b). We considered 112 parcels data from Fedecacao reports. Each parcel data contained age of crop, density of planting, yield, number of fruits harvested, flowering date, harvest date. Acording to personal comunication from framers the biggest flowering occurs in September and January to harvest in March and July. However, the data reported showed that this flowering and pod productions are not constant for all the regions. Cali and Caldas presented the lowest production. However, the pod harvest in Caldas increased in May and from October to December, and decreased from January to March. Meanwhile, Arauca and Apartado reported the highest yield in the months January, July, November and December. Santander has picks of production in March, May and September (Fig.2, a).

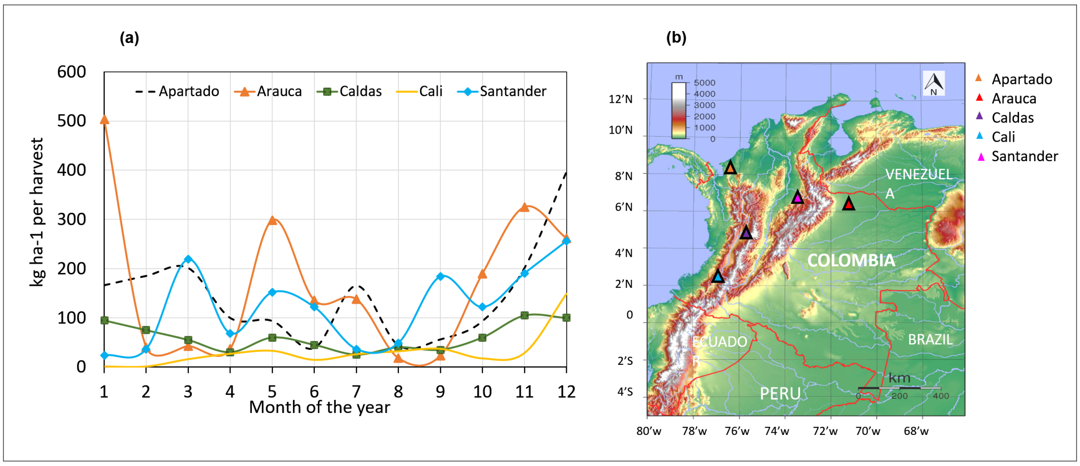

Figure 2. Cocoa production characterization of cocoa for five locations

111 ## 2.3. Weather conditions

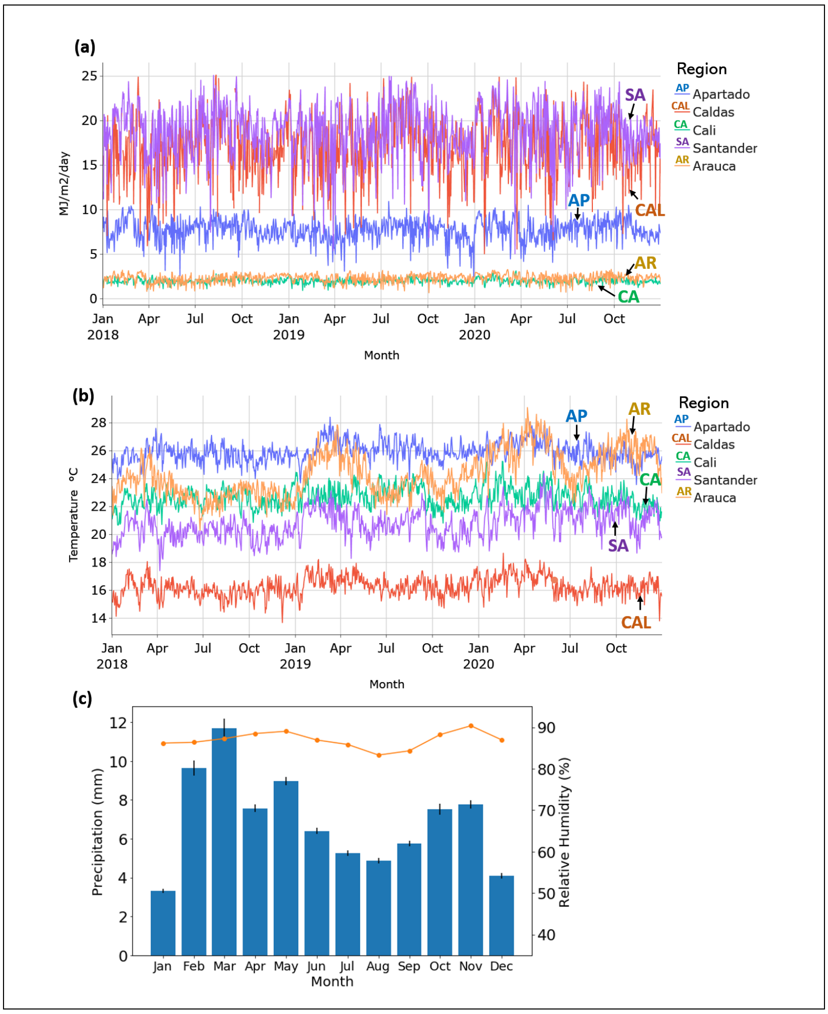

Figure 3. Colombian weather conditions.

(a), Available photosynthetic solar radiation (PAR). (b), Daily average temperature (c), Monthly average of precipitation (bars) and relative humidity (dotted line) from 2018 to 2021

Solar radiation (PAR), temperature, precipitation ant relative humidity was studied from 2018 to 2020 (Fig.3). Santander and Caldas had the biggest variability and the maximum of solar radiation values over 20 MJ $m^2 day^{-1}$. In contrast , Cali and Arauca presented the lowest values of PAR below 5 MJ $m^2 day^{-1}$. (Fig.3, a). Even though, Cali and Santander had contrasting PAR conditions, they regions presented have similar temperature during 2020 (Fig.3, b). The temperature ranges from 16 to 28 °C and it is relatively constant for each region. However, Arauca presented the biggest variability with hotter months during the first half of the year 2019 and 2020. Apartado was found as the hottest region studied with 26 °C and Caldas as the coldest site with 18 °C. Precipitation in Colombia is presented in two seasons per year from February to April and from October to November, while the relative humidity remain constant over 80% (Fig.3,c). In general, the coldest regions tested (Caldas and Santander) had the maximun values of solar radiation available for photosynthesis.

2.4. Thermal time for pod harvest date identification

It is required to model crop growth considering the cocoa base temperature (T_b), at which the plant development stops [25]. This characterization was conducted for the five regions counted 180 days (6 months) after flowering, as farmers used to harvest by calendar days. Characterizing the thermal time the models predicts the maturation day to harvest the cocoa can vary depending on temperature variations. Thermal time (degree days) was calculated from the flowering date to the pod harvest using the ecuation 1. Where tt is the cumulative sum of the daily temperature (T_i) and T_b for cocoa is 10°C according to [26].

$$Thermal\ time\ (tt) = \begin{cases} \sum_{i=1}^{n} T_i - T_b \\ \\ 0, \ \ Flowering\ date \end{cases} \tag{1}$$

2.5. Model Calibration

Calibration of crop models are conducted typically for particular cultivars and require site specific inputs of weather [27]. The procedure for the SIMPLE model [1] calibration was a sequential process of modifying physiological variables specific for cocoa in the inputs files and adding the appropriate files of weather for each region tested. In the SIMPLE model has seven input files where new cocoa crop data should be provided. Files in the list below can be edited to define the features of the new cultivars or experiments. Cocoa crop has not been simulated with SIMPLE model, hence consider previous cocoa studies, values such as leaf area index (LAI) [28] and Harvest index (HI) [29] where modified. Radiation Use Efficiency (RUE) was calibrate according to the perennial crops (banana and cotton) that had been calibrated previously in SIMPLE model [1] with a RUE of 0.8 and 0.85 respectively. As cocoa trees are under shadow the RUE was lower with values between 0.7 and 0.5 g MJ^{-1} m^2 (table 1). Once the physiological parameters were calibrated to the simulated yields for cocoa were reasonably close to the observed yield. 23 flowering dates from 12-July-2019 to 23-June-2020 were introduced in the treatment file, running the program and saving the results.

1. Input/Simulation Management.csv
2. Input/Species parameter.csv
3. Input/Cultivar.csv
4. Input/Treatment.csv
5. Input/Soil.csv
6. Observation/Obsdummy crop Exp name.csv
7. Weather/dummy weather.WTH

2.6. Inputs and parameters

Input variables required to run SIMPLE model for cocoa include the flowering date and daily weather of solar radiation (SRAD), maximum and minimum temperature (TMAX, TMIN) and rain. Weather data as csv file was downloaded from the POWER Data Access Viewer [30] from January 01 2018 to December 31 2020 for four locations in Colombia (Cali, Rionegro -Santander, Apartado - Antioquia, Saravena Arauca). The csv file had to be transformed to .WHT file using R 1.4 version [31].

There were three parameters varied by region (table 1) thermal time required for pod harvest after the flowering date (Tsum) , the Radiation use efficiency (RUE) and yield observed on field. Physiological parameters in table 2 are common for all the regions studied. These parameter were calibrate for cultivars ICS95 and CCN51 considering a range of time of 200 days (DAP) from flowering date to harvest day, even thought farmers collect the pod at 180 DAP. Heat and water stress parameters were not considered.

Table 1: Cocoa crop parameter values used per region.

Region	**Tsum**	**RUE**	**Yield***
Apartado	2906	0.6	3378
Arauca	2764	0.7	3981
Santander	2016	0.6	2687
Cali	1912	0.5	1900
Caldas	1192	0.6	740

RUE Radiation use efficiency (above ground only and without respiration)g MJ^{-1} m^2
* Yield observed kg ha^{-1} per year.

Table 2: Parameter values used to run SIMPLEcocoa model.

File	**Variable name**	**Value**
Treatment	SoilName	Loamy sand4
	InitialFsolar	0.01
	Weather	KOKO (.WTH file name)
	CO_2	400 ppm
	SowingDate	Flowering date
Observation	Crop cycle DAP	200 days
	LAI	1.8
	FSolar	0.70
	Biomass	40kg dry mass per plant
Cultivar	Harvest index	0.3
	150A	680 °C day
	150B	680 °C day
Species	Tbase	10°C
	Topti	26°C
	MaxT	35°C
	ExtremeT	40°C
	CO_2RUE	0.09°C
	S-water	0 ARID index

S-water is associated drought stress evaluations ranging from 0 (no water shortage) to 1 (extreme water shortage) [1]

2.7. Evaluation of model performance

The SIMPLE model performance was evaluated by comparing simulated values cocoa yield with those reported by Fedecacao from cocoa plantations, using the statistical indice of relative root mean square error (RRMSE) (Equ. 2) [1,16].

$$RMSE = \sqrt{\frac{1}{n}\sum_{i=1}^{n}(Y_i - X_i)^2} \tag{2}$$

3. Results

3.1. Weather effects over flowering time

Figure 4 shows the pearson correlation to study the weather data of the flowering time over the months of flowering (monthF), month of harvest (monthH) and their final yield (fruit_kg) . The results showed that thermal time with T_b (ttb), temperature minimum and maximum (TMIN and TMAX) and Dew Frost Point at 2 meters (T2MDEW) are positive related. The wind (WS2M) is clear correlated with ttb. However, less clear correlations were found of monthF with T2MDEW, relative humidity (RH), WS2M and rain.

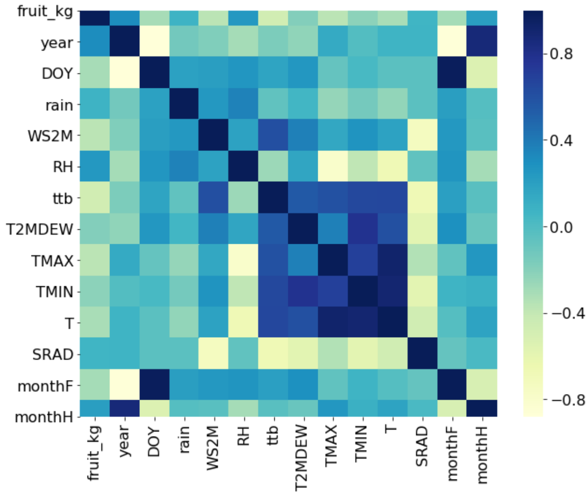

Figure 4. Pearson correlation average weather variables and flowering dates for all regions.

3.2. Thermal time

Thermal time characterization was made considering 180 days after flowering (daf) for each region. The boxplot in the figure 5 shows the data distribution where boxes indicate the range of the central 50% of the total data per region, the central line in the box is marking the median value and lines draw out from each box mean the range of the remaining data. Therefore, this boxplot shows differences between locations as was expected following the tendencies of the temperature (fig. 3,c). Apartado and Arauca had the highest temperatures hence, the highest thermal time values with 2909 and 2764 days°C respectively. Caldas had the lowest values with 1173 days°C. Meanwhile, Cali and Santander presented similar thermal time around 2000 °C (table 1). The accumulated temperature during the pod development (fig.5) depends on the region where cocoa is cultivated and the variety planted. Thermal time values are also proportional to the yield reported on field (table 1).

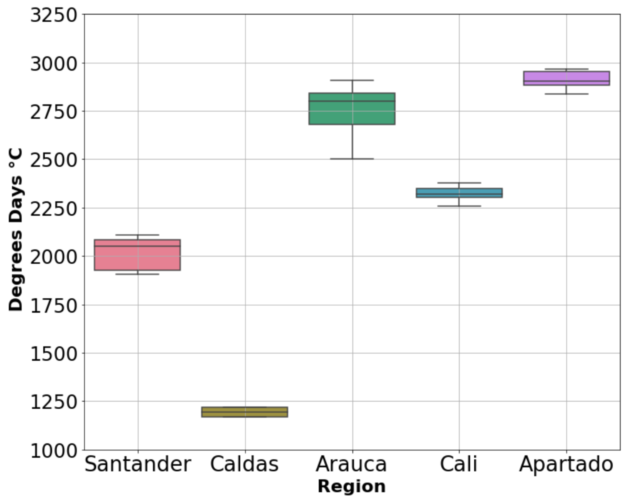

Figure 5. Cocoa yield and thermal time characterization at 180 days after flowering.

3.3. Model validation

Therefore, biomass production of the aerial part of the plant was simulated per region (fig. 6, a) as biomass-rate which is the daily biomass growth rate. As for cocoa simulation we did not consider water and heat stress variables, in this research biomass simulation is affected by the solar radiation, RUE, daily temperature, atmospheric CO_2 concentration (ppm) and the fraction of solar radiation intercepted by a tree of cocoa during the fruit development (fSolar) (fig. 6, b). Due to the lack of this kind of data from fields, the comparison was not possible with observed data of biomass production through the crop cycle. However, the biomass simulation also responded proportionally with the yield production. Biomass of cocoa fruits simulated include all the parts that compound the pod when they are fresh. Therefore, these this information may be of interest to farmers to predict pod fresh mass.

The figure 6, b the fSolar was simulated thought the fruit development resulting that crops in all regions reached the maximun fSolar at 0.94, except Caldas where crops reached 0.76. Apartado and Arauca reached faster this fSolar-max. These high values of solar radiation intercepted for photosynthesis, lasted differently depending on the region and their solar radiation (Fig.3, a) : Apartado 66 days from 69 to 135 days after flowering (daf), Arauca 61 days from 72 to 133 daf, Cali 38 days from 83 to 121 daf, Santander 24 days from 92 to 116 daf and Caldas 2 days at 125 daf. fSolar declined in the interception of solar radiation until the pod harvest day.

The cocoa yield simulation (kg ha^{-1} per year) was validated using observed data (table 1) showing a final RRMSE of 7.2% (fig. 6, c). Individual errors per region are presented in table 3, where the best fit of the calibration model for yield prediction was for crops in Apartado and the highest error was calculated for Caldas crops. The model responded to the variations of temperature and solar radiation. Therefore, the highest yield values simulated were obtained for Arauca over 4000 kg ha^{-1}, followed by Apartado Santander with yields over 2000 kg ha^{-1}. The lowest yield was simulated for Caldas region with less of 1000 kg ha^{-1} . Final yield in the model was is calculated as the product of biomass of aerial part and harvest index (HI) [1,32], where the HI is similar to the CropSyst [33] and AquaCrop [34].

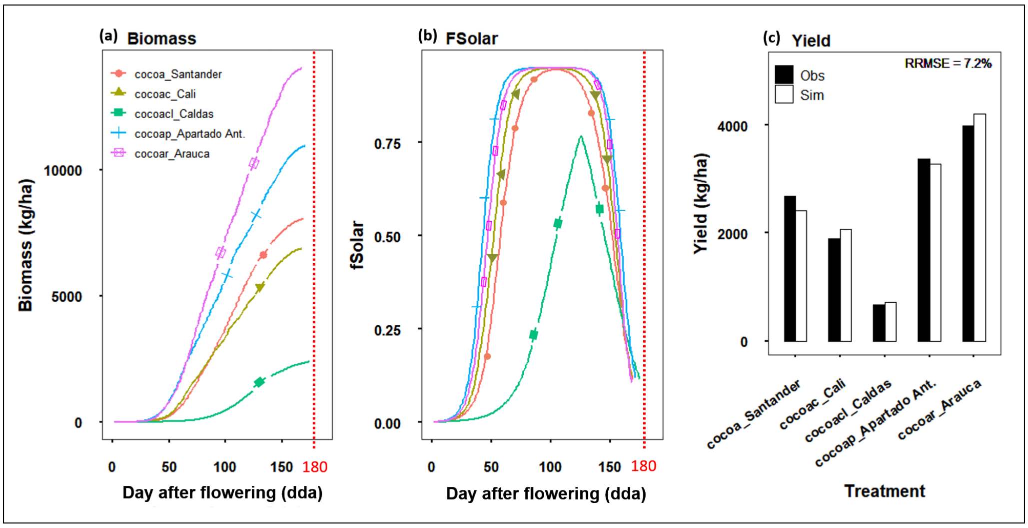

Figure 6. Model predictions. (a) Biomass aerial part. (b) Interception of solar radiation. (c) Yield. Crop cycle close to 180 daf (vertical red line) base on figure 5.

Table 3: Summary of relative root mean square error RMSE for the predicted cocoa yield using SIMPLE model.

Region	**Apartado**	**Arauca**	**Santander**	**Cali**	**Caldas**	**Overal**
RMMSE %	3	6.05	10.06	8.5	14.90	7.2

3.4. Prediction of the pod harvest day

These result demonstrated that the traditional way to harvest always at 180 daf is not having account physiological and environmental conditions that can be affecting the pod development and maturation. Figure 7 present that observe day of harvest or maturity day is 180 daf independently of the region. All the regions except Santander, presented the earliest harvest when the flowering date was between December and February. The fruit can be ready to harvest depending of the enviromental condition per region. Thus, when the fruit development was simulated the maturity day in Cali and Caldas were from 3 to 12 and 4 to 23 days before 180 daf (Fig. 7, c and f respectively). Apartado presented the most similar predicted dates of harvest to 180 daf with 170 to 182 daf (Fig. 7, d). The pod may be harvest in Arauca (Fig.7, b) between 165 and 193 daf, Santander (Fig.7, a) between 165 and 183 daf, ten days less than in Arauca for the same months of flowering of July and August of 2019 . Only Arauca presented longer crop cycles when the flowering was between during that period of time. This means that Arauca had bigger variation of temperature between months (Fig. 3, b).

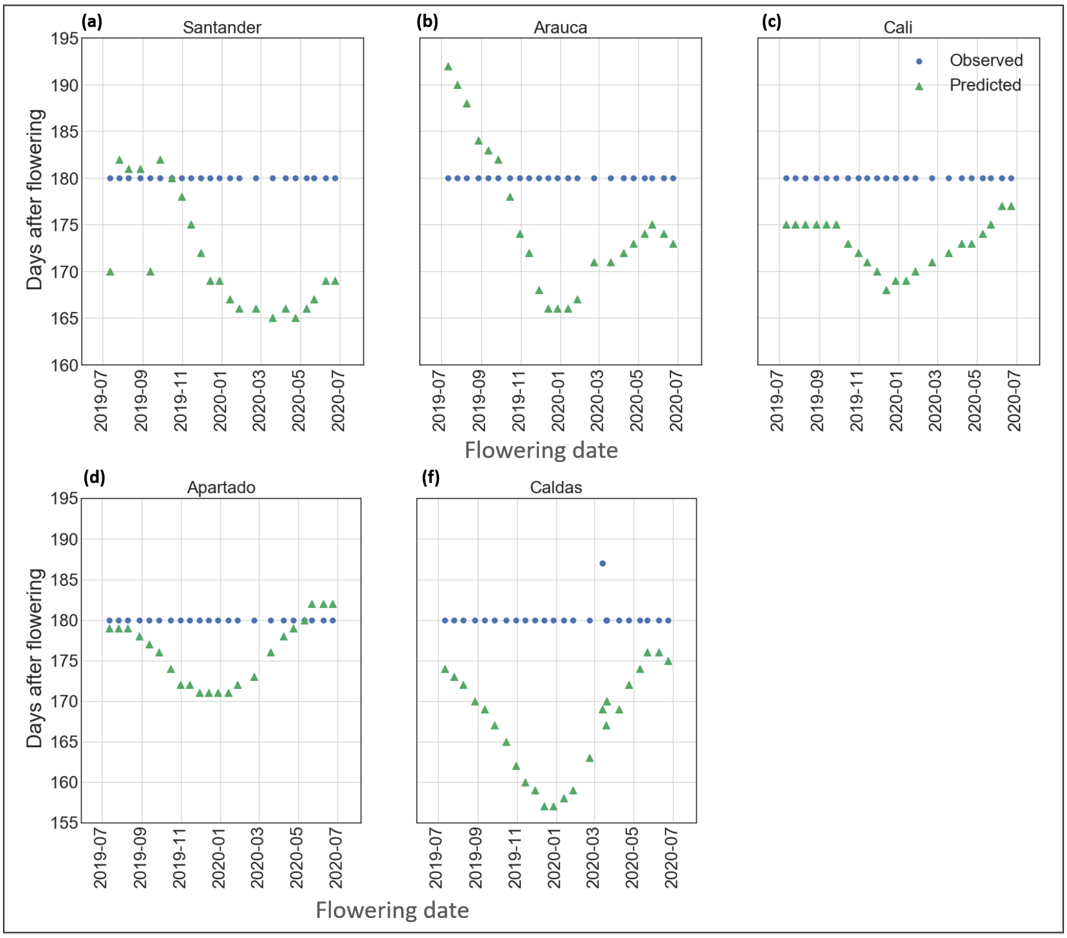

Figure 7. Harvest day prediction from flowering date for (a) Santander, (b) Arauca, (c) Cali, (d) Apartado,(f) Caldas.

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4. Discussion

4.1. Weather Effects over Flower Stability and Pollination

This study presents a new approach of SIMPLE model calibration for cocoa as tropical crop in South America for five environments. The analysis of the weather conditions (Fig. 4) did not confirmed clearly that flower are affected by the wind and rain by mechanical damage. However, farmers stated that the number of flowers pollinated decrease by months where wind and rain are high. Moreover, wind can affect the availability of tiny flies pollinators from Diptera order and from the families of of the biting midges *Ceratopogonidae*, genus *Forcipomyia* [35–37] to reach the cocoa flowers. However, the stability of cocoa flowers is influenced by seasonal wheather conditions (abiotic) and pollination (biotic) [38]. Therefore, pollinator population shoud be coincidence with the phenology of the flowering cocoa trees [39,40]. Flower opening is very well synchronised between the cohorts of mature flowers opening each night [7]. The flowers open at almost exactly the same time and rate, irrespective of their position on the trunk. Thus, unfertilised flowers abscise from the trunk approximately 1 day after flower opening [7]. Hence more than 90% of unpollinated flowers fall or abscised within 32 hours after anthesis [41]. Abscission processed of flowers are mainly controlled by three hormones: auxin, ethylene, and abscisic acid (ABA) [41]. Ethylene generally promotes abscission because it may inhibit the transport of auxin from the leaf blade, which allow the action of ABA to promove the fall of flowers [42]. In general, environmental conditions can also stimulate to a decrease the auxin/ethylene relationship [41].

In our preliminary analysis, Santanter data, showed that the number of successful flowers pollinated to produce final yield can be affected negatively mostly by the rain, TMAX and wind. Nevertheless, a better field data tracing flowers development is essential to understand if it is a mechanical or physiological effect. In contrast, [43,44] indicated that the numbers of cocoa pollinators were reduced during the dry season, but increased in the wet season. This, could be due to midges need a moist environment to develop [38], which is difficult during the dry season as cocoa leaves create a dried ground mat [38,43]. Moreover, the lack of water during dry seasons may reduce the nutrients uptake provoking the massive flower drops [45]. For future studies, wind could be included in the SIMPLE model as an input to simulated this mechanical effects over number of flowers. In general, field data regarding counting flowers pollinated by month should be better reported for the region here studied.

4.2. Thermal Time for Harvest day predictions

We define the thermal time required to harvest cocoa pods for five Colombian regions as maturation of the fruit is related with temperature during the growth cycle [18]. Previous studies have calculated thermal time for different cocoa cultivar in Brazil and Ghana [46]. They also confirm that the fruit maturation time decrease in with an increase in temperature as was presented on others researches [46–48]. the effects of temperature and solar radiation on fruit growth and development was previously studied by [46], showing that crop under higher temperatures thought the crop cycle induce greater fruit losses because of physiological maturation (cherelle wilt). When the fruit is mature, seeds are able to germinate inside the pod before the harvest day [18].

Our results showed that the hotter regions such as Apartado and Arauca presented higher thermal time values (Fig. 5 and 3). Even though, Arauca had very low values of SAR but very high temperatures, this may be caused by clouds cover. The opposite can be seen for Caldas and Santander. These, extreme relations T/SAR can compensate the crop efficiency, for example in Santander (Fig. 6). The thermal time calculation was defined base on 180 daf because farmers cut the pods by calendar days. Previous studies stated that evaluating the level of knowledge of growers regarding cocoa crop management, showed that the harvest was in the group of activities that presented the lowest level of information by the farmers [10]. That is why, these results present important temperature boundaries to predict fruit maturation day. Therefore, may be other environmental factors that should be studied for further research.

4.3. Cocoa Crop Model Simulations

Although cocoa is a relevant crop and there is an extensive agronomic literature, there is only one physiological crop model specific for cocoa so far. The (SUCROS-Cocoa) developed by [2]. However, the code was not easy available for adaptations. In contrast, the SIMPLE model has an open code in R, which we could adapt such a model would be very useful to compare yields and predict harvest date in different climates. As the harvest day was predicted form the flowering date (Fig. 7, consequently, biomass production and fSolar presented a crop cycle shorter than 180 daf (Fig. 6, a and b). These simulations are coincident with results presented by [18], where physiological maturity of coca pod varies from 140 to 162 daf. Our results showed how the harvest day can vary depending on the accumulate temperature during each especific crop cycle simulated (Fig. 7).

Biomass simulations use SIMPLE model presented similar predicted values (10000 Kg ha^{-1}) for coca drops in Costa Rica using SUCROScocoa [2]. Biomass simulations are a common evaluation in crop modells such as Sirius [27], SUBSTOR-potato [49] and DSSAT, CropSyst, STICS and WOFOST [50]. The approach of this research was focus on the harvest date prediction, hence the leaves crop cycle was evaluated indirectly this study. The fraction of intercepted photosynthesis active radiation (fSolar) decreased (Fig. 6, b) when the senescence of the canopy [2]. In cocoa canopy senecence refereed to a group of leaves responsible at the moment of the fruit formation to produce carbohydrates. These leaves eventually drop becoming on litter over the soil. Leaves life cycle has been simulated using crop models [2,51], which can be the reference to improve our SIMPLE coca crop model in future studies.

Yield prediction presented a RRMSE values 7.2%, which were significant lower than those presented for other crops using the SIMPLE model which reported and RRMSE of 24.4% [1]. Resulting in reliable approach for cocoa yield prediction. In general, these results may help to improve the quality of cocoa seed considering the moment to harvest can be variable depending on weather changes.

4.4. App development and future challenges

The original code of SIMPLE model of [1] was modified to make easy the implementation of this cocoa model as and app to be used in smartphones and desktops by farmers in Colombia. Therefore, the new version for cocoa crops simulation will be used to predict yield, date of harvest and biomass production, inserting only the date of flowering and region. The app development is on charge of Grupo BIOS to be deliver to farmers in Caldas initially at the end of 2021. This research presented and initial crop calibration that can be improved with further studies, including effects over the pod production by diseases, nutritional diffidences and abiotic stresses. The future challenge will be that traditional farmers start to harvest more aware of the environmental effects over their crops. It will be necessary that they engage growers with adapt founding from scientific studies. Moreover, It will be required the help of entrepreneurs, researchers, academics and non-specialized communities to transfers the knowledge to cocoa growers.

5. Conclusions

Cocoa fruit development for harvest in the right time depend on whether conditions and principles of crop physiology and flower phenology. This was common for the five regions.Thermal time characterization range from 1200 to 3000 days °C, with a T_b of 10 °C for the fruit development. The SIMPLE model as a generic crop model allowed to predict the harvest date with better precision than only considering days by calendar. Thus, the crop cycle of cocoa for harvest should be shorter than 180 days after flowering. These results confirm the potential of crop model for tropical crop in Latin America.

6. Author contributions

ARV and ARC model calibration and data analsis. AMG field data collection. ARV were primarily responsible for writing the manuscript.

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Conflicts of Interest: The authors declare no conflict of interest.

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