

Assessing The Economic Impact of Insect Pollination on the Agricultural Sector: A Department-Level Case Study in France

Yasmine Blili^{1,2}, Elie Abou Nader¹, Iciar Pavez¹, Paolo Prosperi¹, Rachid Harbouze^{1,2}, Leonidas Sotirios Kyrgiakos⁴, Christina Kleisiari⁴, Marios Vasileiou⁴, Vasileios Angelopoulos⁴, George Vlontzos⁴, Georgios Kleftodimos^{1,3*}

¹ CIHEAM-IAMM - Mediterranean Agronomic Institute of Montpellier, 34090 Montpellier, France

² IAV Hassan II – Hassan II institute of agronomy and veterinary sciences, Rabat BP 6202, Morocco

³ UMR MoISA, Univ Montpellier, CIHEAM-IAMM, CIRAD, INRAE, Institut Agro, IRD, 34090 Montpellier, France

⁴ Department of Agriculture Crop Production and Rural Environment, University of Thessaly, Fytoko, Volos 38446, Greece

Corresponding Author:

*Yasmine Blili

CIHEAM-IAMM - Mediterranean Agronomic Institute of Montpellier, 34090 Montpellier, France

E-mail: kleftodimos@iamm.fr

Copyright: ©2025 Blili Y. This is an open-access article distributed under the terms of the Creative Commons Attribution License, which permits unrestricted use, distribution, and reproduction in any medium, provided the original author and source are credited.

Received: 28-07-2025, Manuscript No. JQR/IJROA/16; Editor Assigned: 29-07-2025, Manuscript No. JQR/IJROA/16; Reviewed: 15-08-2025, Manuscript No. JQR/IJROA/16; Published: 31-08-2025

Abstract:

Pollination is a critical ecosystem service for agriculture, with 76% of European food crops and 80% of wild plants depending on it. However, bee populations are declining due to diseases, pesticides, and climate change, with major economic and environmental impacts. In France, pollination services are valued between 2,3 and 5,3 billion euros annually, but detailed data at Department scale (NUTS 3) is lacking. This study assesses the Economic Value of Crop Production, Insect Pollination, and Agricultural Vulnerability across French departments, using 2022 data and the dependence ratio method. Among 34 crops analyzed, 26 were pollinator dependent. We estimate France's Economic Value of Crop Production at 34,8 billion € and Economic Value of Insect Pollination at 4,2 billion €, with an Agricultural Vulnerability rate of 12%. The highest Economic Value of Insect Pollination per hectare was recorded in Loire-Atlantique (19302,5 €/ha) and the lowest in Seine-Saint-Denis (575,5 €/ha). By analyzing crop-specific dependencies and regional production patterns, the study reveals that southern and western France, particularly departments specialized in fruit and vegetables, are most economically dependent and vulnerable to pollinator decline. A Generalized Additive Model was created to understand the variability of the Economic Value of Insect Pollination per hectare in the different departments of France. This model explained 97,6% of Economic Value of Insect Pollination per hectare variability, showing that vegetable and fruit production strongly drives pollination value. The results highlight spatial disparities in pollination dependency and underscore the need for territorially targeted conservation strategies. Compared to previous studies, our findings suggest a significant underestimation of pollination value, highlighting the need for fine-scale entomological research and territorially targeted conservation strategies to support sustainable agricultural development.

Keywords: Pollinators, Economic Value, Vulnerability Ratio, Dependence Ratio Method, Agriculture, Policy

1. Introduction

Pollination is a vital ecological process that ensures plant reproduction, supports agricultural productivity, and maintains biodiversity (Katumo et al. 2022). Nearly 87,5% of flowering plants and about 75% of the most important global food crops depend, at least partially, on animal pollinators, highlighting their indispensable role in ecosystems and in securing global food supply (Ollerton, Winfree, and Tarrant 2011; Klein et al. 2009).

Pollinators include a wide variety of animals such as insects (bees, butterflies, beetles, flies), birds, bats, and even some small mammals (Katumo et al. 2022; Ollerton 2017). Among these diverse pollinators, insects, particularly bees, are especially valuable for agriculture due to their efficiency in pollinating crops and their significant contribution to agricultural economies (Leonhardt et al. 2013). In fact, pollination directly enhances the quality and nutritional value of many vital food sources, notably fruits, vegetables, and oilseeds. These nutrient-rich foods are fundamental components of balanced human diets, significantly contributing to nutrition and food security worldwide (IPBES 2016). In other words, pollination is recognized as a valuable ecosystem service that contributes broadly to human well-being, providing benefits such as medicinal plants, ornamental aesthetics, genetic diversity, and enhanced ecosystem resilience (Millennium Ecosystem Assessment 2005).

CITE THIS ARTICLE: Blili, Y. ., Nader, E. A. ., Pavez, I. ., Prosperi, P. ., Harbouze, R. ., Kyrgiakos, . L. S. ., Kleisiari, C. . ., Vasileiou, M. ., Angelopoulos, V. ., Vlontzos, G., & Kleftodimos, G. . (2025). Assessing The Economic Impact of Insect Pollination on the Agricultural Sector: A Department-Level Case Study in France. *International Journal of Research in Organic Agriculture*, 1(1), 17-38. <https://ijroa.com/ijroa/article/view/16>

The global economic value of animal pollination services is considerable, estimated at between 127 and 152 billion USD annually (Bauer and Sue Wing 2016; Gallai et al. 2009). Remarkably, despite their relatively modest share of agricultural land use, fruits and vegetables alone constitute over 30% of this economic value, highlighting their dependency on pollinators (Gallai et al. 2009).

However, despite their critical ecological and economic importance, pollinator populations around the world have been alarmingly declining. Numerous studies have identified multiple interconnected factors responsible for these declines, including habitat destruction, intensive farming practices, widespread pesticide usage, diseases caused by introduced pathogens, invasive species competition, and climate change effects (Potts et al. 2010; Cameron et al. 2011; IPBES 2016; Katumo et al. 2022). Agricultural intensification, driven by the increasing global demand for food, exacerbates these negative trends. Intensive farming often reduces habitat diversity, encouraging monoculture systems heavily reliant on pollinator-dependent crops, thereby heightening vulnerability to pollinator population declines. The decline in pollinator populations, particularly observed in Europe and North America, poses severe risks to both food security and economic development, especially in regions highly reliant on agriculture (Steffan-Dewenter et al. 2002; Potts et al. 2010).

Several studies report the decline in pollinator populations, particularly in Europe and North America, where the most consistent data have been gathered. In Europe, for example, approximately 25 % of wild bee species have been lost since the 1980s, while in parts of North America, honeybee colony losses frequently exceed 30 % annually (Potts et al. 2010). Such reductions in pollinator abundance have critical implications for agricultural productivity because many fruit and vegetable producing crops depend on animal-mediated pollination services (Klein et al. 2009; Gallai et al. 2009).

Modeling exercises indicate that, once wild pollinator densities fall beneath certain thresholds, crop yields do not decrease linearly but instead exhibit rapid collapse (Gallai et al. 2009). Gallai et al. (2009) estimated that a 50 % loss of pollinator species, without compensatory measures, could result in a 5 % to 9 % reduction in the total global value of crop production. This nonlinear response arises because crops often require multiple visits by different pollinator taxa to achieve complete fertilization. In regions of southern Europe, where small-scale farmers rely heavily on fruits and vegetables, yield losses of up to 50 % have been projected under severe pollinator decline scenarios, threatening both local food security and rural livelihoods (Potts et al. 2010).

The economic consequences of reduced pollination services extend beyond yield loss. Even when managed pollinators (e.g., commercial honeybees or bumblebee colonies) are deployed to partially substitute for wild insects, Europe could suffer from a multibillion-dollar economic loss, particularly in fruit-dominated systems (Bauer and Sue Wing 2016).

To mitigate these risks, integrated landscape-level and farm-level interventions are required. Agri-environmental schemes, such as establishing flower-rich field margins, reducing pesticide usage, and restoring semi-natural habitats, have been shown to increase wild pollinator abundance and diversity. Simultaneously, the protection and expansion of natural areas provide critical nesting and foraging resources that support pollinator resilience amid intensifying agricultural land use (Potts et al. 2010). In addition, the introduction of alternative managed pollinator species (e.g., *Osmia bicornis*) can supplement honeybee services; however, such strategies require careful management to avoid disease transmission and adverse effects on native pollinator communities (Cameron et al. 2011).

Failure to implement these measures risks a future in which fruit and vegetable production declines sharply, commodity prices escalate, and rural communities suffer decreased food security and economic stability (IPBES 2016). Therefore, investments in pollinator-friendly agricultural practices, habitat conservation, and diversification of pollination management are essential to maintain crop productivity and ecosystem health under ongoing environmental change.

Addressing pollinator decline requires coordinated international actions, combining existing national and local monitoring efforts into a comprehensive global initiative. Given the multifaceted and interacting threats pollinators face, continuous improvement in our understanding of pollinator health and dynamics at local, national, and global scales remains indispensable. Extensive global and national assessments of pollination's economic value have been conducted (e.g., (Carreck and Williams 1998; A. Morse and W. Calderone 2000)), smaller-scale analyses at regional (NUTS 2) and departmental (NUTS 3) levels are relatively uncommon (Borges et al. 2020). These localized assessments are particularly important due to the limited foraging distances of many wild pollinator species, such as solitary bees (Gathmann and Tschamtkke 2002). Precise, smaller-scale evaluations are thus more accurate and relevant for conservation efforts and policy planning. Consequently, the objective of this study is to perform an economic valuation of pollination services at the departmental level (NUTS 3) in France in order to better assess the contribution of pollinators to local and national economies as well as to better assess the vulnerability of these regions in an even of a "pollination services crisis".

The choice of France as a case study stems from its remarkable agricultural heterogeneity and the availability of detailed crop- and region-specific data. France encompasses a wide range of production systems, ranging from extensive cereal belts in the north to intensive fruit and vegetable zones in the south. This diversity allows for the simultaneous examination of both pollinator-dependent

(e.g., fruits, vegetables, nuts) and less-dependent (e.g., cereals) crop categories within the same national framework. Moreover, France maintains comprehensive statistical records at the departmental level through sources such as the Agreste database (Ministère de l'Agriculture), which facilitate the disaggregation of crop-specific values and surface areas. Finally, ongoing national initiatives, such as the "Plan national Pollinisateurs 2021-2026" launched by the French Ministry of Agriculture, underscore the policy relevance of quantifying pollination services in a country where agricultural lobby groups and environmental agencies actively seek to balance productivity with biodiversity conservation.

In order to do so, we employ the Dependence Ratio method introduced by Gallai et al. (2009), which assesses the economic contribution of pollinators to agriculture based on crop-specific pollination dependency levels. France offers a suitable case study due to its diverse agricultural practices and a variety of cultivated crops, including cereals, fruits, vegetables, tubers, and vineyards.

This research seeks to accomplish four primary objectives. We first derive, for each department, the total Economic Value of Crop Production (EVCP). We then calculate the share of this value that can be directly attributed to insect pollination by aggregating crop-level dependency estimates across the chosen cultivated species. Next, we identify those departments whose agricultural output is most reliant on pollination services, thus revealing spatial hotspots of vulnerability. Finally, we explore the underlying drivers of these spatial patterns by correlating departmental pollination values with factors such as crop types, and the proportion of agricultural land.

To achieve these objectives, we construct a detailed set of economic indicators and examine their relationships through correlation analysis. Furthermore, we apply Generalized Additive Models (GAM) to investigate non-linear relationships between crop compositions and the economic importance of pollination.

By leveraging the results of our study, it becomes possible to guide more effectively the scientific research and financial investment toward French departments where urgent measures are needed to protect pollinators. This approach can help minimize the risks linked to the loss of pollination services (lower agricultural productivity, food insecurity, economic instability in rural areas, etc.). The structure of the paper is as follows: the next section details our methodological approach, followed by a comprehensive presentation of our results. We then interpret these findings within the context of existing literature, concluding with implications for policy and practice, along with recommendations for future research.

2. Materials and Methods

2.1. Study area

France was selected as the location of interest for two reasons. First and foremost, detailed data concerning agricultural production at the departmental level, with indicators like total quantity, yield, and surface cultivated, is available and easily accessible. Secondly, and due to its geography, France is rich in terms of biodiversity. It is bordered by the Mediterranean Sea to the south, the Atlantic Ocean to the west, and the English Channel to the north. Its diverse topography includes mountains, forests, rivers, lakes, and plains. Thus, its agriculture sector benefits immensely from its biodiversity, especially from the wide range of pollinators it provides. For that reason, France is a suitable candidate for this research. According to data.gouv.fr, France is divided into 13 regions and 96 departments (excluding overseas territories). In the results part, calculated values will be grouped into departments for comparison and modelling purposes.

2.2. The dependence ratio method

Calculating the economic benefits of pollination services involves multiple approaches with fluctuating intricacy and regularly entails a volume of notable assumptions. That said, initial research adopted the full price of pollinated crops to estimate the benefits of pollination services, leading to an overestimation of the latter. The dependence ratio method resolves this overvaluation issue by using crop pollination ratios, hypothetical measurements of the percentage of the decline in crop production in the case of a pollination services void. These measurements are obtained from published literature (IPBES 2016) reflecting the highest advantages of insect pollination to crop species, irrespective of the applied growing systems, variety, and cultivar. Even though this method features the proportion of crop production lost to the lack of pollinators (wild and managed), it doesn't factor in other crop-producing variables, which can lead to the overvaluation of pollination services (Breeze et al. 2016).

Because of its adaptability to small geographical scale models, this method helps us harvest the advantages of pollination services across the 96 departments and the different crop types grown by French farmers. With a representative set of data, and while still recognizing the shortcomings of this methodology with the overgeneralization amongst the different varieties, the method assumes that pollination services are at their peak, ignoring other ecosystem services and inputs.

a. Economic Value of Crop Production

Interventionary studies involving animals or humans, and other studies require ethical approval must list the authority that provided approval and the corresponding ethical approval code.
 This is an example of a quote.

$$EVCP = \sum_{i=1}^I \sum_{x=1}^X (P_i \times Q_{ix})$$

b. Economic Value of Insect Pollination

The second indicator is the Economic Value of Insect Pollination (EVIP). Insects and other pollinators provide value for the agriculture sector. This value is determined by multiplying the market value of crop production by the dependence ratio of each crop, as shown in the equation below. P_i represents the price of the crops, Q_{ix} the quantity produced in the department, and D_i the dependence ratio, which is a specific value to each crop that was taken from Appendices 1 and 2 of the work by Klein et al. (2009), which classifies crops based on their level of reliance on animal pollinators. The declared result is in euros. Likewise, the results are grouped into the French departments, with i representing the designated crop, while x represents the department where it is produced.

$$EVIP = \sum_{i=1}^I \sum_{x=1}^X (P_i \times Q_{ix} \times D_i)$$

c. Vulnerability Ratio

After calculating the EVCP and the EVIP for each department, we move to understand the degree of dependence of the department on pollinators and pollination services. Following Gallai et al. (2009), the vulnerability ratio (VR) is the ratio of the economic value of insect pollination and the economic value of crop production. A low VR suggests that the agriculture sector is resilient to pollinator's decline, and a high VR implies the opposite, which means that there is a significant dependence of the agriculture revenue on insect's pollination. The equation can be found below.

$$VR = \frac{EVIP}{EVCP} = \frac{\sum_{i=1}^I \sum_{x=1}^X (P_i \times Q_{ix} \times D_i)}{\sum_{i=1}^I \sum_{x=1}^X (P_i \times Q_{ix})}$$

d. Economic Value of Insect Pollination per hectare of land

This indicator is important due to its ability to scale the importance of pollinators according to the size of the agricultural land of each department. It serves as an important comparative measure between the different geographical areas. The economic value of insect pollination per hectare of agricultural land (or EVIP/ha) is calculated by dividing the EVIP of a department by the sum of the agricultural land available in that department. In other words, it is the total economic value of insect pollination per department, accounting for the agricultural area of the respective department.

$$EVIP/ha = \frac{\sum_{i=1}^I \sum_{x=1}^X (P_i \times Q_{ix} \times D_i)}{\sum \text{Agricultural Land}}$$

Data collection

To analyze crop dependence on pollinators, three main datasets were compiled.

The first dataset contains crop production figures by department for the year 2022. These data were collected from the French Ministry of Agriculture's official database ('Agreste, La Statistique Agricole', n.d.). In this source, crops are grouped into five major categories: Fruit, Vegetables, Tubers, Vineyards, and COP (cereals, oilseeds, and proteinaceous).

The second dataset includes crop-specific pollination dependence ratios. These values were taken from Appendices 1 and 2 of the work by Klein et al. (2009), which classifies crops based on their level of reliance on animal pollinators.

The third dataset concerns crop prices, which required a more nuanced approach. Most price data were retrieved from the ('FAOSTAT', n.d.) platform, complemented by information from the ('European Commission's Eurostat', n.d.) database.

CITE THIS ARTICLE: Blili, Y. ., Nader, E. A. ., Pavez, I. ., Prosperi, P. ., Harbouze, R. ., Kyrgiakos, . L. S. ., Kleisiari, C. . ., Vasileiou, M. ., Angelopoulos, V. ., Vlontzos, G., & Kleftodimos, G. . (2025). Assessing The Economic Impact of Insect Pollination on the Agricultural Sector: A Department-Level Case Study in France. *International Journal of Research in Organic Agriculture*, 1(1), 17-38. <https://ijroa.com/ijroa/article/view/16>

For several crops, we attempted to obtain specific 2022 price data from ('IndexBox Platform', n.d.), however these figures were behind a paywall and not publicly accessible. As an alternative, we used the freely available data for production volume and production value from the year 2020 to estimate average unit prices. These 2020 values were then adjusted to 2022 levels using the Index of Agricultural Product Prices at Production (IPPAP) (base 100 in 2020) that we got from the Agreste databases, following this formula:

$$2022 \text{ Price} = \frac{2020 \text{ Price} \times IPPAP \ 2022}{100}$$

For others, such as tomatoes and sunflowers, we applied a different method, given the unavailability of data for these crops. We used the producer price index (2014–2016 = 100) for 2022 and the producer price in 2015, both available on FAOSTAT. The 2022 estimate was derived using this formula:

$$2022 \text{ Price} = \frac{2015 \text{ Price} \times IPPAP \ 2022}{100}$$

These procedures allowed us to approximate crop prices with a reasonable degree of confidence.

Finally, each crop was assigned a pollination dependence level (no increase, little, modest, great, or essential) corresponding to numeric values of 0, 0,05, 0,25, 0,65, and 0,95. All datasets were consolidated into an Excel file and then imported into RStudio for processing and analysis.

3. Results

3.1. National level

Although we emphasize the significance of small-scale evaluations of pollination services, assessing them at the national level can still provide crucial insights into the overall importance of pollinators to the French agricultural sector.

For the studied crops, in 2022, 34 crops were investigated, of which 26 depend on pollinators in varying degrees, 8 have No Increase, 3 have little increase, 7 have a modest increase, 11 have a great increase, and 5 are essential. 11 crops belong to the fruit category, 9 to the COP, 12 are vegetables, 1 is vineyard, and 1 is tubers.

The total national EVCP is 34,8 billion €, of which 4,19 billion € (EVIP) is directly attributed to pollinators, making the national vulnerability 12%, and the EVIP/ha equals 591,4 € per hectare. Below is the table for crops identified, including crop type, pollinator dependence, EVCP, EVIP, VR, and EVIP/ha.

Table 1: Crops produced in France, their dependence on pollinators, and pollination service value

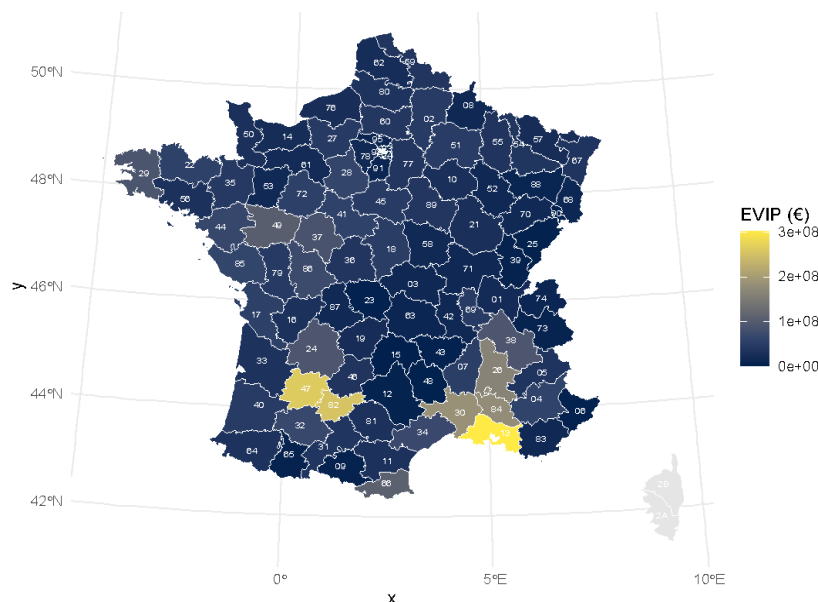
Crop	Crop type	DR	Dependance	EVCP	EVIP	VR	EVIP/HA
Apricot	fruit	0,65	Great	215911117,83	140342226,59	0,65	12355,16
Eggplant	vegetable	0,25	Modest	52180207,89	13045051,97	0,25	11483,32
Oat	cop	0,00	No increase	84345663,64	0,00	0,00	0,00
Cherry	fruit	0,65	Great	169372311,31	110092002,35	0,65	14593,32
Chestnut	fruit	0,25	Modest	46434047,35	11608511,84	0,25	1290,84
Pumpkin	vegetable	0,95	Essential	235868073,14	224074669,49	0,95	30065,03
Rapeseed	cop	0,25	Modest	3025374540,38	756343635,10	0,25	614,84
Cucumber	vegetable	0,65	Great	163675370,94	106388991,11	0,65	101516,21
Zucchini	vegetable	0,25	Modest	174974488,85	43743622,21	0,25	10763,69
Strawberry	vegetable	0,25	Modest	316490471,41	79122617,85	0,25	20303,47
Raspberry	fruit	0,65	Great	54714725,15	35564571,35	0,65	55743,84
Green bean	vegetable	0,05	Little	283654970,87	14182748,54	0,05	451,69

Kiwi	fruit	0,95	Essential	146543511,00	139216335,45	0,95	35433,02
Oilseed	cop	0,05	Little	42636567,52	2131828,38	0,05	73,73
Corn	cop	0,00	No increase	3366165532,94	0,00	0,00	0,00
Melon	vegetable	0,95	Essential	369269128,92	350805672,47	0,95	27992,79
Turnip	vegetable	0,65	Great	15757733,22	10242526,59	0,65	4097,01
Hazelnut	fruit	0,95	Essential	19566651,78	18588319,19	0,95	2460,40
Nut	fruit	0,95	Essential	128452919,25	122030273,29	0,95	4535,94
Barley	cop	0,00	No increase	3283259404,97	0,00	0,00	0,00
Peach	fruit	0,65	Great	348428885,16	226478775,35	0,65	19809,93
Small peas	vegetable	0,00	No increase	110203798,82	0,00	0,00	0,00
Pear	fruit	0,65	Great	137713670,14	89513885,59	0,65	15159,00
Leak	vegetable	0,65	Great	133342851,11	86672853,22	0,65	16152,23
Pepper	vegetable	0,05	Little	40120989,46	2006049,47	0,05	1921,50
Apple	fruit	0,65	Great	950398616,60	617759100,79	0,65	15661,28
Potato	tubercule	0,00	No increase	2964227949,53	0,00	0,00	0,00
Plum	fruit	0,65	Great	324088637,64	210657614,47	0,65	14037,29
Grape	vineyard	0,00	No increase	15353138199,59	0,00	0,00	0,00
Rye	cop	0,00	No increase	33494414,92	0,00	0,00	0,00
Soy	cop	0,25	Modest	232947013,45	58236753,36	0,25	316,67
Sorghum	cop	0,00	No increase	67150953,05	0,00	0,00	0,00
Tomato	vegetable	0,65	Great	590379776,10	383746854,47	0,65	76079,87
Sunflower	cop	0,25	Modest	1362521668,40	340630417,10	0,25	391,28
Total	-	-	-	34842804862,34	4193225907,59	0,12	591,42

3.2. Departmental level

To better understand how pollination services are distributed across French departments, we imported our data into RStudio. We calculated the indicators for each department (EVIP, VR, and EVIP per hectare). We then created maps using Rstudio to visualize the spatial distribution of these indicators and highlight regional patterns.

Map 1: Mapping the Economic Value of Insect Pollination in French departments



a. Economic Value of Insect Pollination per hectare

To enhance the interpretation of EVIP per hectare (EVIP/ha) across French departments, a quantile-based classification was applied. Given the limited dispersion of EVIP/ha values, using tertiles improved the visualization of spatial contrasts by categorizing departments into high, medium, and low pollination values per hectare.

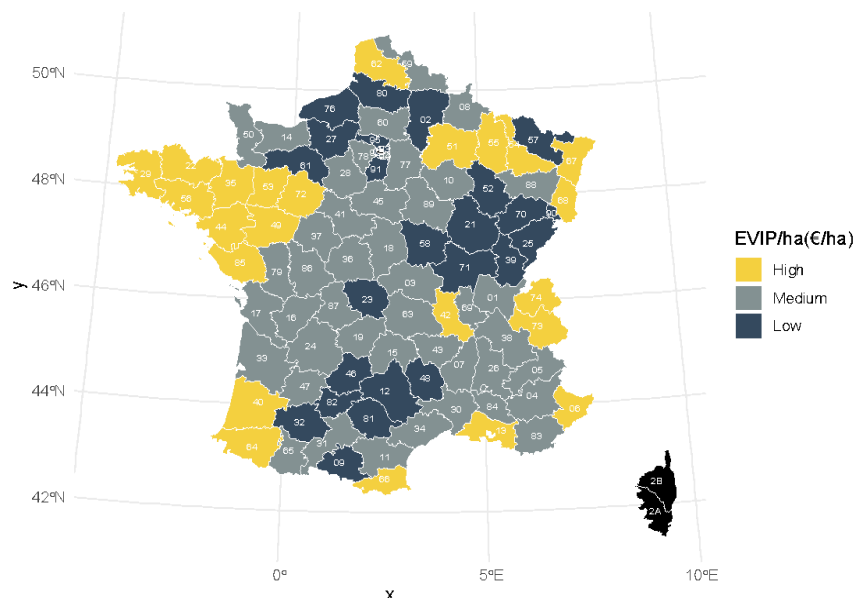
Unlike the absolute EVIP map, which reflects total economic contributions, the EVIP/ha map (map 2) emphasizes pollination efficiency relative to cultivated land. This reveals new patterns: western coastal regions, notably Bretagne and Pays de la Loire (e.g., Loire-Atlantiques and Finistère). In these regions, horticulture, floriculture, and market gardening are widespread, often in small plots that require intensive pollination, leading to high economic returns per unit of land. Similarly, Bouches-du-Rhône in the Mediterranean zone exhibits the highest EVIP/ha values as it benefits from favorable climatic conditions and diversified cropping systems, heavily reliant on pollination services.

Conversely, departments within the Île-de-France region (e.g., Seine-Saint-Denis, Hauts-de-Seine, Essonne) display low EVIP/ha values, which illustrate the dilution effect of land pressure and urban expansion. Here, the agricultural footprint is minimal and often fragmented. Similarly, departments in central and eastern France, dominated by cereal and livestock farming, show generally moderate to low values.

Notably, some departments with high absolute EVIP, such as Lot-et-Garonne (47) or Gard (30), do not necessarily appear in the top tier for EVIP/ha. This discrepancy reveals that their high value stems more from land area than pollination efficiency, emphasizing the need for both indicators in complementary use.

Ultimately, the quartile-based EVIP/ha map illustrates that high pollination value per hectare is not evenly spread across France, but rather clustered in regions where climate, land use, and crop systems create favorable conditions for pollinator-dependent production. It shifts the discussion away from "where is there the most agriculture" to "where does agriculture rely most efficiently on pollinators." The EVIP/ha map offers a refined lens on where pollination services are most productive per hectare, but it still does not reflect the economic risk associated with pollination dependence. For that, we must turn to the vulnerability rate (VR), which quantifies the proportion of total agricultural value at risk from pollinators decline.

Map 2: Mapping the Economic Value of Insect Pollination per hectare of agricultural land in French departments



b. Vulnerability Ratio

The third map representing the absolute vulnerability ratio provides a continuous view of the proportion of agricultural value dependent on insect pollination within each French department. Unlike EVIP or EVIP/ha, the VR reflects the structural exposure of agricultural systems to potential pollinator decline, independent of the total or per-hectare value.

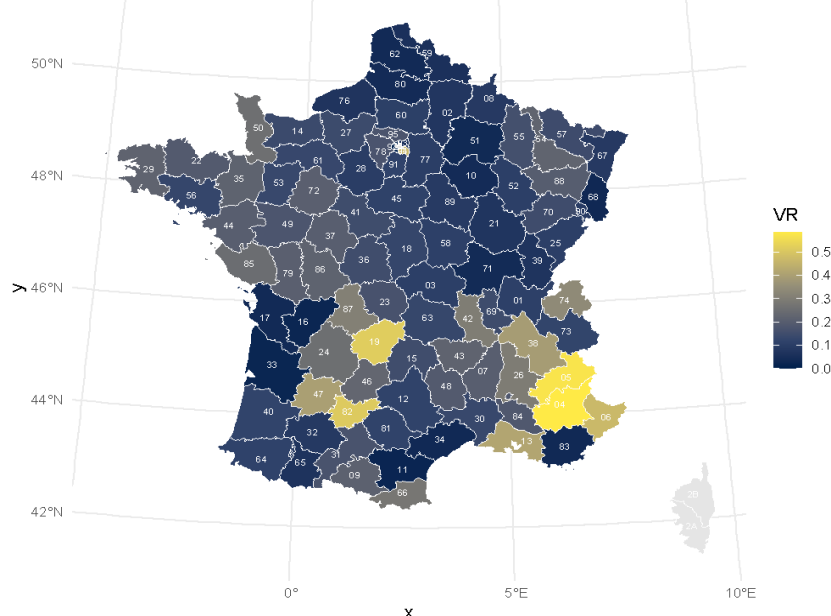
Departments such as Alpes-de-Haute-Provence (04), Hautes-Alpes (05), Corrèze (19), and Tarn-et-Garonne (82) exhibit the highest vulnerability rates, exceeding 0,4 and reaching up to 0,58. These regions, often characterized by mountainous or Mediterranean climates, have fragmented agricultural landscapes dominated by high-dependence fruit and vegetable crops, such as apricots, apples, or melons.

In contrast, departments in northern and central France, where urbanization, cereal production, and livestock farming prevail, show lower VR values. These systems are typically more mechanized and less reliant on ecological services like pollination.

Interestingly, while Tarn-et-Garonne (82) appeared among the top departments in both EVIP and VR, others like Corrèze (19), with modest absolute EVIP, rank high in VR due to their structural crop dependency. These high vulnerability rates are not necessarily correlated with total agricultural output or land area.

Many of the most vulnerable departments have limited agricultural surfaces but are highly specialized in pollination-sensitive production.

The VR map thus highlights structural vulnerabilities that could result in significant economic impacts if pollination services decline, particularly in regions with small-scale, specialized agriculture. Moreover, combining VR analysis with climate and land use data reveals that Mediterranean and mountainous areas, already prone to climatic stresses like drought and frost, face compounded risks, reinforcing their critical dependence on maintaining pollinator populations.

Map 3: Mapping the Pollination vulnerability in French departments


c. Implications and key insights

The combined analysis of EVIP, EVIP per hectare, and vulnerability rate (VR) offers a detailed view of how insect pollination supports agriculture across French departments. It shows that pollination services do not carry the same weight everywhere. In southern and western regions, especially under Mediterranean and oceanic climates, the absolute EVIP is highest, reflecting the large-scale production of fruit and vegetables that heavily depend on pollinators. However, when looking at EVIP per hectare, coastal regions like Bretagne and Pays de la Loire emerge, where smaller but more diverse and intensive cropping systems make pollination services highly valuable per unit of land.

The VR indicator brings another dimension: it highlights areas such as Alpes-de-Haute-Provence and Corrèze, where even modest agricultural sectors are structurally very dependent on pollinators, making them more vulnerable to their decline.

Together, these three perspectives show that pollination's economic role varies by both scale and intensity, shaped by climate, crop diversity, land use, and regional specialization. These findings make clear that protecting pollination services requires targeted, region-specific strategies, balancing both economic importance and ecological risk to support sustainable agriculture.

3.3 Identifying departments where pollination has a significant importance to the agriculture sector

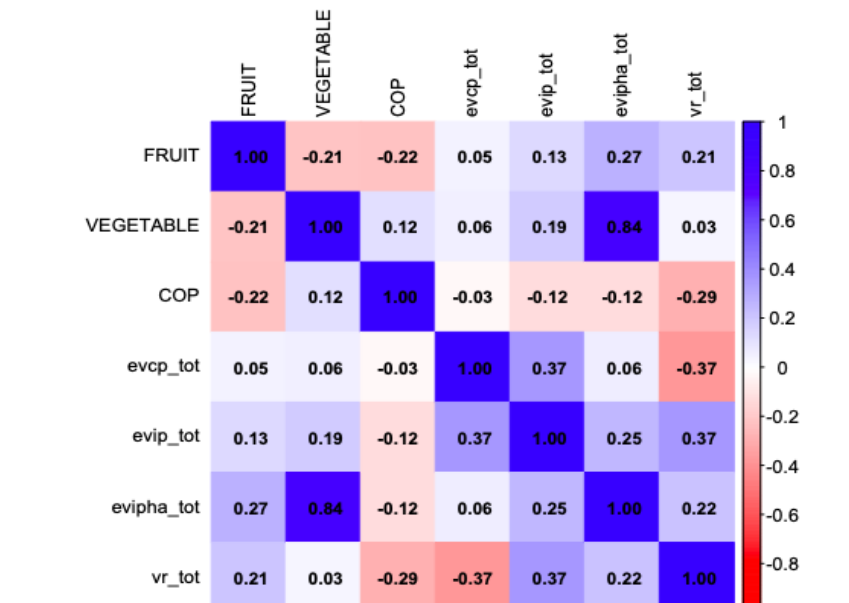
a. Correlation matrix

i. Correlation matrix of crop types and pollination indicators:

The correlation matrix analyzes the relationships between three major crop categories, FRUIT, VEGETABLE, and COP (Cereals, Oilseeds, Protein crops) (Vineyard and tuber crops were excluded from the correlation analysis, as their associated EVIP, EVIP per hectare, and vulnerability values were systematically zero across all departments, making it statistically impossible to compute meaningful correlations for these categories) and four key indicators reflecting economic dependency on pollination services:

- EVCP_tot: total agricultural income,
- EVIP_tot: total economic value of insect pollination,
- EVIP/ha: economic value of pollination per hectare,
- VR_moy: average vulnerability to pollinator decline.

Figure 1: Correlation Matrix of crop types and pollination indicators



The analysis of the correlation matrix (Figure 1) reveals important statistical patterns across French departments, shedding light on the differentiated role of crop types in shaping the economic significance and ecological vulnerability of pollination services.

Starting with vegetable cultivation, we observe a strong and significant positive correlation with EVIP/ha ($r = 0.84$, $p = 0 < 0.001$), confirming the critical dependence of vegetables such as melon, cucumber and pumpkin on insect pollination. In departments where vegetable farming is prominent, the economic value per hectare is notably high. Additionally, these regions display a weak but statistically significant positive correlation with total EVIP ($r = 0.19$, $p = 0.03 < 0.05$). These areas combine economic efficiency with a certain degree of ecological fragility, especially in the absence of crop diversification.

Fruit crops, although known for their pollination dependency, show a weak but significant correlation with EVIP/ha ($r = 0.27$, $p = 0.009 < 0.01$). Despite high dependency ratios in species like apple, pear, and apricot, their impact remains regionally confined and diluted.

In contrast, COP crops display a moderate and significant negative correlation with VR ($r = -0.36$, $p = 0.004 < 0.01$), confirming their role in providing structural resilience. This inverse relationship is coherent with the biological characteristics of major staples like barley, oats, and sorghum ($RD = 0$), as well as partially dependent crops like sunflower and rapeseed ($RD = 0.25$).

Finally, the link between EVIP/ha and vulnerability (VR) is moderate and statistically significant ($r = 0.38$, $p = 0.006 < 0.01$). This result points to a clear pattern: departments deriving the highest per-hectare value from pollination services are also among the most ecologically sensitive. This reinforces the need for targeted conservation policies in economically valuable yet ecologically exposed areas.

ii. Scatterplot of EVIP/ha vs Vulnerability (VR)

To better understand the link between the economic value of pollination services and the vulnerability of agricultural systems, we constructed a scatterplot crossing EVIP/ha with the vulnerability rate for each department. This approach allows us to visualize how the value of pollination per unit of land is related to the risk of pollinator decline.

The scatterplot (figure 2) plots French departments based on two variables: EVIP per hectare (€ / ha) on the x-axis and vulnerability to pollination decline (VR) on the y-axis. Each department is represented by a numerical label (e.g., 006 for Alpes-Maritimes). Overall, we can see from figure 2 that the form of the scatterplot is an elliptical cloud with a slight upward trend, indicating a general but non-linear association between EVIP/ha and vulnerability. A dense cluster is visible between EVIP/ha values of 10000 to 15000 euros and VR levels between 0.3 and 0.4.

At the upper end of the spectrum, departments such as Loire-Atlantiques (044) and Pyrénées-Orientales (066) report both high EVIP/ha values (19303 € and 16879 €, respectively) and high vulnerability rates ($VR = 0.41$ and 0.44). These results reflect their specialization in high-value, pollination-dependent crops, particularly vegetables (e.g., over 43000 €/ha for VEGETABLE production

International Journal of Research in Organic Agriculture Volume 1, Issue 1, Year 2025

in Loire-Atlantiques). Similarly, Côtes-d'Armor (022) and Finistère (029) display EVIP/ha values exceeding 17000 €, consistent with their intensive agricultural systems fostered by oceanic and temperate climates. Same thing for Alpes-Maritimes (006), despite a relatively small agricultural base (EVCP \approx 9 million €), stands out with high vulnerability (VR = 0,48) and a high EVIP/ha (15989 €), reflecting the strong reliance of its Mediterranean horticultural production on pollinators.

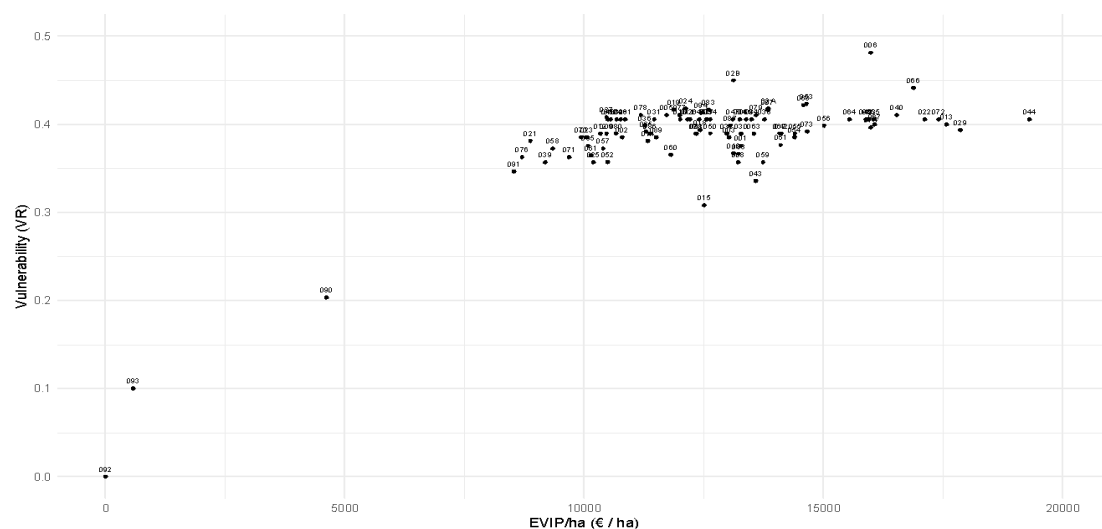
At the opposite end, highly urbanized departments such as Hauts-de-Seine (092) and Seine-Saint-Denis (093) exhibit very low EVIP/ha (0 € and 576 €, respectively) and low vulnerability (VR = 0 and 0,10), reflecting the near-absence of productive agricultural land. Territoire de Belfort (090) also reports low EVIP/ha (4618 €) and vulnerability (VR = 0,20), consistent with its small surface area and industrial-economic orientation. Other departments like Essonne (091) and Seine-Maritime (076) present modest EVIP/ha values (8533 € and 8700 €) and moderate vulnerability, linked to the dominance of COP crops and relatively less land of fruits and vegetables.

Between these two extremes, departments such as Cantal (015) and Haute-Loire (043) show moderate EVIP/ha levels (12508 € and 13584 €) combined with lower-than-average vulnerability (VR = 0,31 and 0,33), characteristic of livestock-oriented agricultural systems in the Massif Central.

Southern and western departments, with Mediterranean or oceanic climates, tend to favor intensive vegetable and fruit farming, increasing both EVIP/ha and vulnerability. Northern and eastern departments, where cereals and industrial crops dominate, show a little less reliance on pollination. Mountainous and livestock-oriented areas like Auvergne and the Massif Central lie in a balanced middle zone

However, the scatterplot as a whole reveals a more nuanced reality: most departments cluster within a relatively narrow range of vulnerability and EVIP/ha values. This indicates that, regardless of crop specialization, the decline in pollination services has the potential to affect nearly all regions to varying degrees. The observed patterns suggest that additional factors, beyond agricultural structure alone, must be considered to fully understand the exposure of each region to pollinator decline.

Figure 1: Scatterplot of EVIP/ha vs Vulnerability (VR)

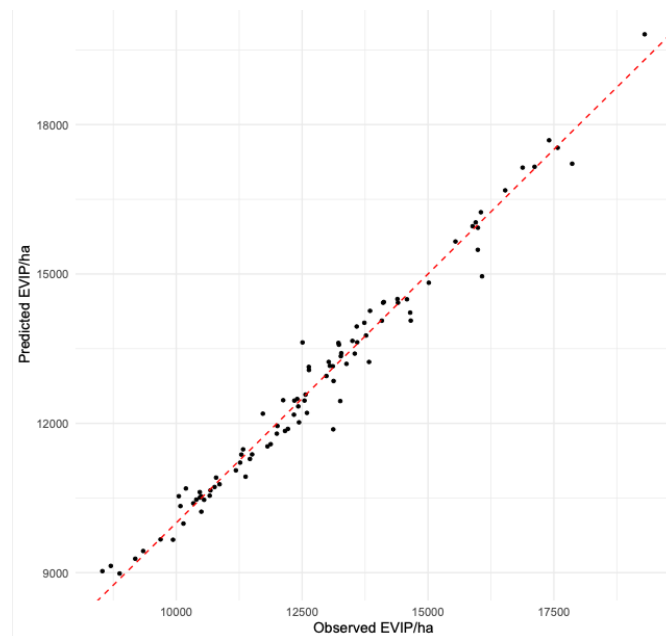


b. GAM model

iii. Observed vs predicted values of the Economic Value of Insect Pollination per hectare of land

Figure 3 illustrates the relationship between the observed and predicted values of EVIP/ha from the GAM model. The points cluster tightly along the reference line, indicating a highly accurate model fit. The minimal dispersion confirms the strong predictive power of the model, consistent with the adjusted R² of 0,97 and the 97,6% deviance explained. This validates the model's robustness in capturing the underlying structure of the data.

Figure 2: Observed vs predicted values of the Economic Value of Insect Pollination per hectare of land

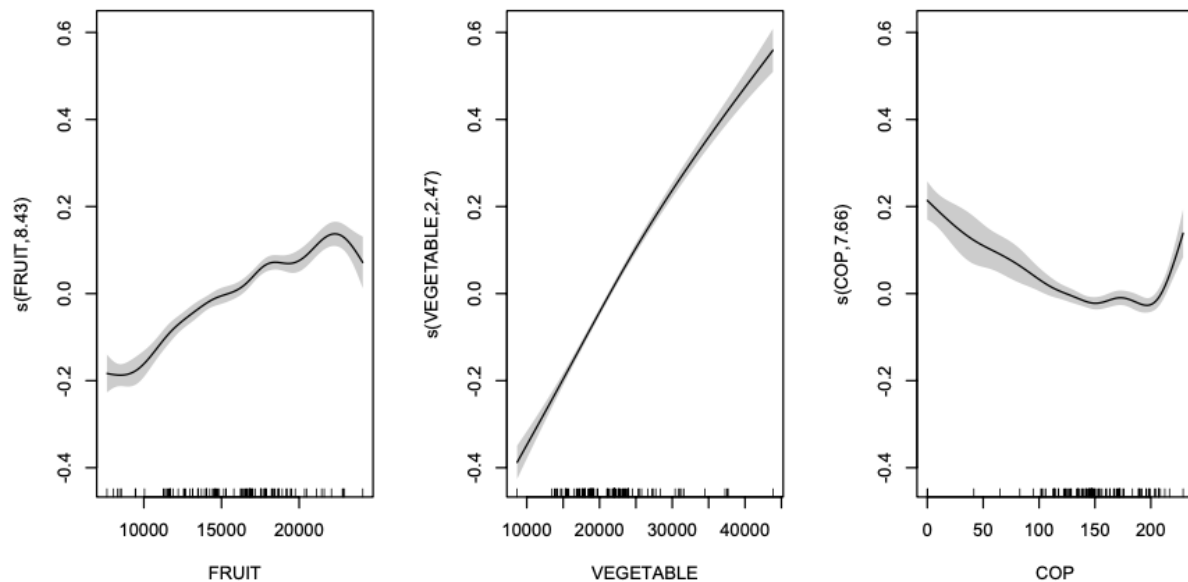


iv. Smooth terms: Non-linear effects of crop categories

Figure 4 below displays the estimated smooth terms for the effects of FRUIT, VEGETABLE, and COP crop areas on the log-transformed EVIP/ha.

- **VEGETABLE:** The strongest effect is observed for vegetables, with a sharp rise in EVIP/ha up to 20000 hectares, before flattening. This supports the finding from the correlation analysis that vegetable area is the primary driver of EVIP/ha, due to the high pollination dependency of many vegetable crops (e.g., melon, pumpkin, tomato).
- **FRUIT:** The effect is positively non-linear, showing a steep increase up to approximately 10000 hectares, followed by a plateau and slight decline beyond 13000 hectares. This suggests a diminishing marginal return of pollination services in fruit-dominant departments beyond a certain threshold.
- **COP:** A negative non-linear effect is observed. EVIP/ha decreases with increasing COP area, then stabilizes. This is consistent with the fact that COP crops (e.g., cereals, rapeseed, oats) have little or no dependence on insect pollination, thus reducing the relative value of pollination services in such areas.

Figure 4: Estimated smooth terms for the effects of Fruit, Vegetable, and COP crop areas on the log- transformed EVIP/ha

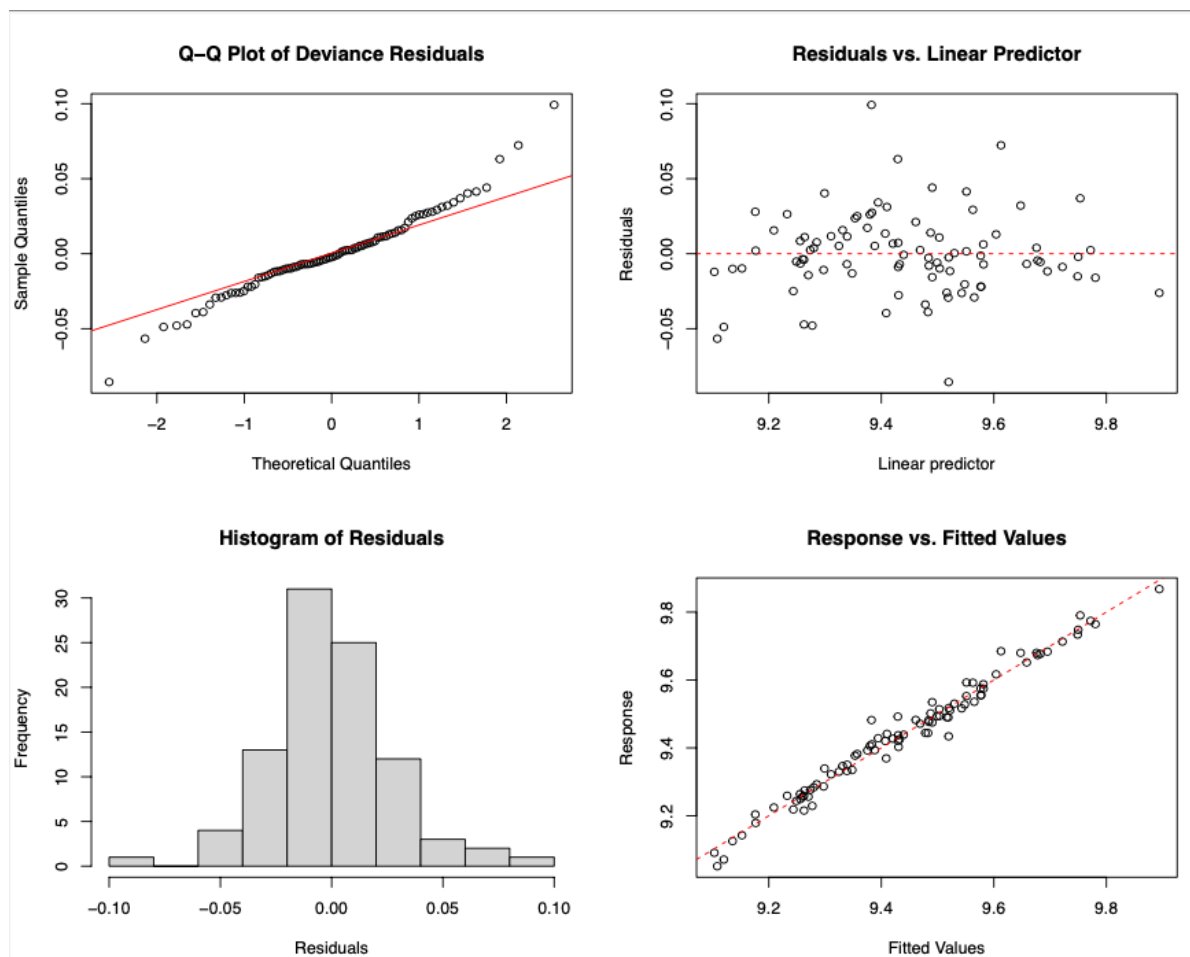


v. Model diagnostics

The figures below (5, 6, 7, and 8) present the standard diagnostic plots used to evaluate the statistical validity of the GAM model:

- QQ plot of deviance residuals (figure 5): The residuals align closely with the reference line, suggesting that the normality assumption is satisfied.
- Residuals vs linear predictor (figure 6): The residuals are randomly scattered around zero, without any visible pattern, confirming the absence of heteroscedasticity or systematic bias.
- Histogram of residuals (figure 7): The distribution appears symmetrical and centered around zero, supporting the normality of errors.
- Response vs fitted values (figure 8): The observed responses closely follow the fitted values along the diagonal, indicating a nice overall model fit.

Figure 3, 6, 7, 8: Model diagnostics



vi. Implications and key insights

The GAM model predicting the Economic Value of Insect Pollination per hectare (EVIP/ha) shows strong reliability, with an adjusted R^2 of 0,97 and 97,6% of deviance explained. The close match between observed and predicted values, as well as the good distribution of residuals, confirms the model's robustness.

Smooth term analysis reveals that vegetable and fruit areas strongly increase EVIP/ha, while the expansion of COP crops reduces it. These non-linear effects highlight the critical role of crop composition in shaping the economic value of pollination services. Overall, the model effectively captures the key ecological-economic relationships across departments.

4. Discussion

In this study, we explored the economic importance of pollinators to agriculture using a high-resolution approach at the departmental level across France. Nationally, we estimate the total Economic Value of Crop Production (EVCP) to be 34,8 billion €, the total Economic Value of Insect Pollination (EVIP) to be 4,19 billion €, the average vulnerability ratio (VR) to be 12%, and the national average EVIP per hectare (EVIP/ha) to be 591,4 €/ha. When compared with the estimates provided by Leonhardt et al. (2013), who valued France's EVIP at $2,1 \pm 0,5$ billion €, our figure is almost twice as high. The contrast is even more striking with EVIP/ha: Leonhardt et al. report an average of 56,12 €/ha, over ten times lower than our estimate. The vulnerability ratio, however, remains comparable, 12% in our study versus $9\% \pm 1\%$ in theirs, which reinforces the reliability of this indicator across methodologies.

Several reasons may explain these differences. First, inflation and global price shifts between 2013 and 2022 have increased the market value of pollination-dependent crops. Second, our methodology leverages finer spatial granularity. The departmental approach

CITE THIS ARTICLE: Blili, Y. ., Nader, E. A. ., Pavez, I. ., Prosperi, P. ., Harbouze, R. ., Kyrgiakos, . L. S. ., Kleisiari, C. . ., Vasileiou, M. ., Angelopoulos, V. ., Vlontzos, G., & Kleftodimos, G. . (2025). Assessing The Economic Impact of Insect Pollination on the Agricultural Sector: A Department-Level Case Study in France. *International Journal of Research in Organic Agriculture*, 1(1), 17-38. <https://ijroa.com/ijroa/article/view/16>

in our study captures heterogeneity in production systems, crop specialization, and local land use, unlike Leonhardt et al.'s country-level averages. This allows for a more accurate assessment of where pollinators contribute the most. Yet, despite methodological and temporal differences, both studies converge on the central insight: if pollinators were to collapse, France would face a double-digit reduction in crop production value. As a matter of fact, Leonhardt et al. (2013) have stated that given the importance of pollination services, the EU might face a monetary loss due to declining numbers of pollinators, although it is unlikely that the world will lose all pollinating insects. This also aligns with international estimates, which showed that even under assumptions of partial substitution and price adjustments, Europe could suffer from a multibillion-dollar economic loss, particularly in fruit-dominated systems (Bauer and Sue Wing 2016).

The VEGETABLE and FRUIT categories are not only economically important but also sinequanone for nutritional security, which explains why the vulnerability is concentrated in these specific crop types. A decline in pollination services would thus affect both economic output and public health (Breeze et al. 2016; IPBES 2016). Moreover, spatial analysis through figures 4 and 5 demonstrates that southern and western French departments are significantly more reliant on pollination, both in terms of EVIP/ha and VR. This is consistent with studies across the EU, where warmer Mediterranean countries reveal both higher pollination value and greater dependency on pollination services, due to their production of fruit and vegetable-dominated systems (Leonhardt et al. 2013). The correlation matrix in figure 1 confirms this connection, revealing that departments with higher EVIP/ha also tend to exhibit greater vulnerability. This suggests that these departments are not only economically reliant on pollination but also structurally exposed to pollinator decline, a concern that Bauer & Wing (2010) further emphasized through macroeconomic simulations, showing how sectoral shocks from pollination loss can amplify regional disparities and threaten rural economic resilience.

Figure 2 highlights departments where pollination services are concentrated. However, it's important to distinguish between ecological risk and economic vulnerability. A department with a high VR doesn't necessarily mean that its pollinator population is most threatened; it means that the local agricultural economy would suffer the most in the case of a pollination service failure. Nonetheless, we note that in orchard-dominated systems, wild bee richness and abundance are often lower than in more heterogeneous landscapes (Bommarco, Marini, and Vaissière 2012). This statement raises concerns for French departments with a high VR where FRUIT cultivation predominates, suggesting the need for targeted ecological assessments and conservation action. This is where the EVIP/ha indicator is useful. It reflects the economic return of pollinators per unit of land. It is going to help inform regional subsidy allocation and design incentive schemes to protect pollination services. This said, departments with a high EVIP/ha should be prioritized in pollinator conservation programs, especially when combined with high VR.

Overall, the multi-indicator framework developed in our study offers a nuanced lens through which policymakers and scientists can identify French departments where pollination services are both economically essential and ecologically at risk. Leonhardt et al. (2013) relied on country-wide averages and coarse resolution data to make national-level assessments, but our approach leverages high-resolution, crop-specific production values and recent department-level data to provide a much finer and more accurate spatial diagnosis. This level of detail is important for territorialized policy interventions. In addition to the granularity, our study integrates multiple indicators (total EVIP, EVIP per hectare, and vulnerability ratio), which together offer a multidimensional understanding of how and where pollination matters most.

Moreover, the relevance of our approach is amplified in today's context. Since the study of Leonhardt et al. (2013), agricultural prices have changed due to inflation, market volatility, and the increasing demand for nutrient-dense foods, fruits, and vegetables, for instance, pollinator-dependent crops. Our method provides an updated view and highlights how regions within the same country face disproportionate risks if pollination services continue to decline.

Our findings can also be situated within the broader international literature. For example, Bauer & Sue Wing (2010) simulated the effects of a global pollinator collapse using a computable general equilibrium (CGE) model. Their partial equilibrium approach estimated global direct losses in the crop sector at 138,3 billion \$, while their CGE model, accounting for indirect effects on non-agricultural sectors and price changes, estimated total global economic losses at 334,1 billion \$. Interestingly, the direct agricultural loss estimated under general equilibrium was much lower (10,5 billion \$), illustrating that partial approaches may overstate crop-level risks while underestimating broader economic repercussions. Compared to their global-scale and macroeconomic framework, our study adopts a bottom-up approach. The granularity of our data allows us to identify precise hotspots of economic and ecological risk, particularly in vegetable and fruit-dominated French zones. Our study thus complements the CGE model by giving direct and explicit insights for more territorial policies, while also echoing Bauer & Sue Wing's conclusion that pollinator losses, whether direct or systemic, present serious economic threats.

Ultimately, by combining spatial precision with economic and ecological insight, this framework empowers policymakers to make better-informed decisions in a country, France, for instance. It supports the design of targeted agri-environmental schemes, regional incentive structures, and biodiversity strategies that reflect the current territorial value of pollination, ensuring that interventions are scientifically grounded, economically justified, and socially equitable.

5. Conclusions

Our study advances a detailed, spatialized assessment of the economic significance of insect pollination across French departments, using a multi-indicator framework that integrates the total Economic Value of Insect Pollination (EVIP), EVIP per hectare (EVIP/ha), and the vulnerability ratio (VR). Unlike prior national-scale assessments, our departmental approach captures the spatial heterogeneity of crop systems, pollination dependencies, and ecological risks. By doing so, it provides a clearer picture of where pollination services are economically critical and ecologically fragile.

Our results show that the highest total EVIP is found in southern and western departments, especially those with Mediterranean and oceanic climates. These areas are hubs for fruit and vegetable production, crop types that are highly dependent on pollinators and offer high returns per hectare. When looking at EVIP/ha, departments such as Loire-Atlantiques and Finistère emerge as hotspots of pollination efficiency, where small but diverse plots contribute significantly to national pollination value. Meanwhile, the vulnerability ratio (VR) highlights regions like Alpes-de-Haute-Provence and Corrèze, where the agricultural economy, although modest in scale, is disproportionately reliant on pollination services. These areas would suffer the most from pollinator decline, not necessarily due to scale, but due to structural crop dependency. The positive correlation between EVIP/ha and VR underscores a fundamental tension: the regions that benefit most from pollinators are often the most at risk if those services are lost, and this calls for urgent territorial policy responses.

However, our study is not without limitations. First, the modeling framework assumes a static and linear relationship between crop yield and pollination dependency, which doesn't take into consideration adaptive farming practices (shifting crop types or increasing artificial pollination, for example), nor longer-term ecological feedbacks such as pollinator population resilience or land-use shifts. These complex dynamics could alter the actual impact of pollinator decline on crop yields and economic outcomes. In contrast, the Computable General Equilibrium model (CGE), developed by Bauer & Sue Wing (2010), attempts to simulate these adaptive behaviors and systemic feedbacks at a macroeconomic level, providing a complementary but broader-scale understanding of pollination decline effects. Second, the vulnerability ratio (VR) is a valuable indicator of structural exposure to pollination loss, but it doesn't directly measure pollinator health, abundance, or diversity. In reality, these ecological parameters depend on a multitude of factors, such as pesticide use, habitat conversion, and climate change. As a result, a department may exhibit high vulnerability without necessarily having degraded pollinator populations, and vice versa. Third, due to data constraints, the study does not include the full spectrum of crops cultivated in France. Consequently, some minor or region-specific crops were omitted, possibly leading to a slight underestimation of our numbers. Furthermore, some crop types, such as vineyards and tubers, were excluded from the correlation and modeling analyses. This exclusion was specifically methodological. In our dataset, both grapes (vineyard) and potatoes (tuber) have a dependency ratio (RD) of zero. As a result, their associated EVIP, EVIP/ha, and VR values were systematically zero across all departments. Including these values in the correlation matrix and the predictive model would have introduced statistical noise. Their exclusion ensures a cleaner and more interpretable model structure.

Despite these limitations, the framework presented in our analysis offers a tool for territorialized policy planning. By moving beyond national averages and integrating recent, crop-specific economic data, this study updates previous estimates, such as those by Leonhardt et al. (2013).

As for next steps, future research should seek to integrate ecological data on pollinator populations, land-use fragmentation, and pesticide use to better contextualize economic vulnerability. It is also important to couple economic indicators with social and political variables like policy uptake capacity, farmer behavior, and public incentives, which would enhance the operational relevance of decision-making. Additionally, as Bauer & Sue Wing (2010) have demonstrated, the decline of pollinators not only affects the agricultural sector directly, but has consequences that go beyond that. Even modest reductions in pollination services could disrupt food systems and exacerbate regional inequalities. In France, as shown in this study, these risks are unevenly distributed, concentrated in zones where vegetable and fruit production dominate. This calls for the design of agri-environmental policies that are both spatially targeted and economically justified.

In conclusion, safeguarding pollination services in a changing climate and volatile market environment requires more than broad conservation goals. It demands precise, localized, and economically informed actions. The multi-indicator framework developed here lays a foundation for such efforts. It allows for the identification of departments where ecological vulnerability and economic dependency intersect, thereby supporting more effective, regionally adapted, and future-oriented strategies for pollinator protection and sustainable agriculture in France.

Funding: This work was supported by the ALLIANCE Project and has been financed under the European Union's HE research and innovation program under grant agreement N° 101084188.

Credit Authorship Contribution Statement: Yasmine Bili: Writing - original draft, Formal analysis, Methodology, Data curation, Software, Formal analysis, Conceptualization. Elie Abou Nader: Writing - original draft, methodology, Conceptualization. Georgios Klefodimos: Writing – original draft, Writing – review & editing, Conceptualization, Data curation, Supervision, Validation. Iciar

CITE THIS ARTICLE: Bili, Y. ., Nader, E. A. ., Pavez, I. ., Prosperi, P. ., Harbouze, R. ., Kyrgiakos, . L. S. ., Kleisiari, C. . ., Vasileiou, M. ., Angelopoulos, V. ., Vlontzos, G., & Klefodimos, G. . (2025). Assessing The Economic Impact of Insect Pollination on the Agricultural Sector: A Department-Level Case Study in France. *International Journal of Research in Organic Agriculture*, 1(1), 17-38. <https://ijroa.com/ijroa/article/view/16>

International Journal of Research in Organic Agriculture

Volume 1, Issue 1, Year 2025

Pavez: Writing – review & editing, Validation. Paolo Prosperi: Validation. Rachid Harbouze: Validation. Leonidas Sotirios Kyrgiakos: Validation. Christina Kleisiari: Validation. Marios Vasileiou: Validation. Vasileios Angelopoulos: Validation. George Vrontzos: Validation.

Declaration Of Competing Interest: The authors declare no conflict of interest.

Data Availability: Data will be made available on request.

Declaration Of Generative AI And AI-Assisted Technologies in the Writing Process: This article was written entirely by the authors without the use of generative AI or AI-assisted technologies.

References:

1. A. Morse, Roger, and Nicholas W. Calderone. 2000. 'The Value of Honey Bees As Pollinators of U.S. Crops in 2000'.
2. 'Agreste, La Statistique Agricole'. n.d. Accessed 30 May 2025. <https://agreste.agriculture.gouv.fr/agreste-web/disaron/SAA-SeriesLongues/detail>
3. Bauer, Dana Marie, and Ian Sue Wing. 2016. 'The Macroeconomic Cost of Catastrophic Pollinator Declines'. *Ecological Economics* 126 (June):1–13. <https://doi.org/10.1016/j.ecolecon.2016.01.011>.
4. Bommarco, Riccardo, Lorenzo Marini, and Bernard E. Vaissière. 2012. 'Insect Pollination Enhances Seed Yield, Quality, and Market Value in Oilseed Rape'. *Oecologia* 169 (4): 1025–32. <https://doi.org/10.1007/s00442-012-2271-6>.
5. Borges, R C, R M Brito, V L Imperatriz-Fonseca, and T C Giannini. 2020. 'The Value of Crop Production and Pollination Services in the Eastern Amazon'. *Neotropical Entomology* 49 (4): 545–56. <https://doi.org/10.1007/s13744-020-00791-w>.
6. Breeze, Tom D., Nicola Gallai, Lucas A. Garibaldi, and Xui S. Li. 2016. 'Economic Measures of Pollination Services: Shortcomings and Future Directions'. *Trends in Ecology & Evolution* 31 (12): 927–39. <https://doi.org/10.1016/j.tree.2016.09.002>.
7. Cameron, Sydney A., Jeffrey D. Lozier, James P. Strange, Jonathan B. Koch, Nils Cordes, Leellen F. Solter, and Terry L. Griswold. 2011. 'Patterns of Widespread Decline in North American Bumble Bees'. *Proceedings of the National Academy of Sciences* 108 (2): 662–67. <https://doi.org/10.1073/pnas.1014743108>.
8. Carreck, Norman, and Ingrid Williams. 1998. 'The Economic Value of Bees in the UK'. *Bee World* 79 (3): 115–23. <https://doi.org/10.1080/0005772X.1998.11099393>.
9. 'European Commission's Eurostat'. n.d. Accessed 30 May 2025. <https://ec.europa.eu/eurostat>.
10. 'FAOSTAT'. n.d. Accessed 30 May 2025. <https://www.fao.org/faostat/en/#home>.
11. Gallai, Nicola, Jean-Michel Salles, Josef Settele, and Bernard E. Vaissière. 2009. 'Economic Valuation of the Vulnerability of World Agriculture Confronted with Pollinator Decline'. *Ecological Economics* 68 (3): 810–21. <https://doi.org/10.1016/j.ecolecon.2008.06.014>.
12. Gathmann, Achim, and Teja Tscharntke. 2002. 'Foraging Ranges of Solitary Bees'. *Journal of Animal Ecology* 71 (5): 757–64. <https://doi.org/10.1046/j.1365-2656.2002.00641.x>.
13. 'IndexBox Platform'. n.d. Accessed 30 May 2025. https://app.indexbox.io/?_gl=1*10yuic0*_ga*MzM2Mzc1NjE3LjE3NDU1ODUxOTQ.*_ga_6KCVGEDSJE*MTc0NTU4NTE5NC4xLjEuMTc0NTU4NTIxOC4wLjAuMA.
14. IPBES. 2016. 'Summary for Policymakers of the Intergovernmental Science-Policy Platform on Biodiversity and Ecosystem Services on Pollinators, Pollination and Food Production'.
15. Katumo, Daniel Mutavi, Huan Liang, Anne Christine Ochola, Min Lv, Qing-Feng Wang, and Chun-Feng Yang. 2022. 'Pollinator Diversity Benefits Natural and Agricultural Ecosystems, Environmental Health, and Human Welfare'. *Plant Diversity* 44 (5): 429–35. <https://doi.org/10.1016/j.pld.2022.01.005>.
16. Klein, Alexandra-Maria, Bernard E. E Vaissière, Ingolf Steffan-Dewenter, Saul A Cunningham, Claire Kremen, and Teja Tscharntke. 2009. 'Importance of Pollinators in Changing Landscapes for World Crops'. 2009. <https://doi.org/10.1098/rspb.2006.3721>.
17. Leonhardt, Sara Diana, Nicola Gallai, Lucas Alejandro Garibaldi, Michael Kuhlmann, and Alexandra-Maria Klein. 2013. 'Economic Gain, Stability of Pollination and Bee Diversity Decrease from Southern to Northern Europe'. *Basic and Applied Ecology* 14 (6): 461–71. <https://doi.org/10.1016/j.baae.2013.06.003>.
18. Millennium Ecosystem Assessment, ed. 2005. *Ecosystems and Human Well-Being: Synthesis*. The Millennium Ecosystem Assessment Series. Washington, DC: Island Press.
19. Ollerton, Jeff. 2017. 'Pollinator Diversity: Distribution, Ecological Function, and Conservation'. *Annual Review of Ecology, Evolution, and Systematics* 48 (1): 353–76. <https://doi.org/10.1146/annurev-ecolsys-110316-022919>.
20. Ollerton, Jeff, Rachael Winfree, and Sam Tarrant. 2011. 'How Many Flowering Plants Are Pollinated by Animals?' *Oikos* 120 (3): 321–26. <https://doi.org/10.1111/j.1600-0706.2010.18644.x>.
21. Potts, Simon G., Jacobus C. Biesmeijer, Claire Kremen, Peter Neumann, Oliver Schweiger, and William E. Kunin. 2010. 'Global Pollinator Declines: Trends, Impacts and Drivers'. *Trends in Ecology & Evolution* 25 (6): 345–53. <https://doi.org/10.1016/j.tree.2010.01.007>.

CITE THIS ARTICLE: Blili, Y. ., Nader, E. A. ., Pavez, I. ., Prosperi, P. ., Harbouze, R. ., Kyrgiakos, . L. S. ., Kleisiari, C. . ., Vasileiou, M. ., Angelopoulos, V. ., Vrontzos, G., & Kleftodimos, G. . (2025). Assessing The Economic Impact of Insect Pollination on the Agricultural Sector: A Department-Level Case Study in France. *International Journal of Research in Organic Agriculture*, 1(1), 17-38. <https://ijroa.com/ijroa/article/view/16>

International Journal of Research in Organic Agriculture
Volume 1, Issue 1, Year 2025

22. Steffan-Dewenter, Ingolf, Ute Münzenberg, Christof Bürger, Carsten Thies, and Teja Tscharntke. 2002. 'SCALE-DEPENDENT EFFECTS OF LANDSCAPE CONTEXT ON THREE POLLINATOR GUILDS'. *Ecology* 83 (5): 1421–32. [https://doi.org/10.1890/0012-9658\(2002\)083\[1421:SDEOLC\]2.0.CO;2](https://doi.org/10.1890/0012-9658(2002)083[1421:SDEOLC]2.0.CO;2).

vii. Appendices

Table 1: correlation matrix

	FRUIT	VEGETABLE	COP	EVCP_TOT	EVIP_TOT	EVIPHA_TOT	VR_TOT
FRUIT	1	-0,2121	-0,2202	0,0474	0,1302	0,2701	0,2072
VEGETABLE	-0,2121	1	0,0457	0,0952	0,2149	0,8440	0,0697
COP	-0,2202	0,0457	1	-0,0417	-0,1303	-0,1814	-0,2919
EVCP_TOT	0,0474	0,0952	-0,0417	1	0,3832	0,1494	-0,3338
EVIP_TOT	0,1302	0,2149	-0,1303	0,3832	1	0,2842	0,3809
EVIPHA_TOT	0,2701	0,8440	-0,1814	0,1494	0,2842	1	0,2780
VR_TOT	0,2072	0,0697	-0,2919	-0,3338	0,3809	0,2780	1

Table 2: p-value matrix

	FRUIT	VEGETABLE	COP	EVCP_TOT	EVIP_TOT	EVIPHA_TOT	VR_TOT
FRUIT	1	0,0424	0,0350	0,6538	0,2162	0,0092	0,0475
VEGETABLE	0,0424	1	0,6620	0,3612	0,0375	0,0000	0,5044
COP	0,0350	0,6620	1	0,6898	0,2106	0,0802	0,0043
EVCP_TOT	0,6538	0,3612	0,6898	1	0,0001	0,1484	0,0009
EVIP_TOT	0,2162	0,0375	0,2106	0,0001	1	0,0052	0,0001
EVIPHA_TOT	0,0092	0,0000	0,0802	0,1484	0,0052	1	0,0064
VR_TOT	0,0475	0,5044	0,0043	0,0009	0,0001	0,0064	1

Table 3: Dependance ratios table from appendices 1 and 2 from klein and al. (2009)

SOURCE	CROP	DEPENDANCE RATIO	DEPENDANCE RATIO
APPENDIX 1	Okra, gumbo	modest	0,25
	pigeon pea, cajan pea, congo bean	little	0,05
	jack bean, horse bean, sword bean	modest	0,25

International Journal of Research in Organic Agriculture
Volume 1, Issue 1, Year 2025

	chile pepper, red pepper, bell peper, green pepper	little	0,05
	quinoa	no increase	0
	chickpea, gram, garbanzo bean	no increase	0
	watermelon	essential	0,95
	cantaloupe, melon	essential	0,95
	cucumber, gherkin	great	0,65
	pumpkin, squash, gourd, marrow, zucchini	essential	0,95
	guar bean, goa bean	little	0,05
	hyacinth bean, horse-gram, lablab	modest	0,25
	buckwheat	great	0,65
	lentils	no increase	0
	tomato	little	0,05
	velvet bean	unknown	unknown
	Kidney bean, Haricot bean, Lima bean, Adzuki bean, Mungo bean, String bean	little	0,05
	Garden pea, Field pea	no increase	0
	Winged bean, Goa bean	unknown	unknown
	Eggplant, Aubergine	modest	0,25
	Cowpea, Blackeye pea, Blackeye bean	little	0,05
	Bambara beans, Bambara groundnuts, Earth pea	little	0,05
	Kiwifruit	essential	0,95
	Atemoya, Cherimoya, Custard apple	essential	0,95
	Tree- strawberry	modest	0,25
	Breadfruit	unknown	unknown
	Jackfruit	unknown	unknown
	Pawpaw, Indiana banana	essential	0,95
	Carambola, Starfruit	great	0,65
	Papaya	little	0,05
	Bergamot, Chinotto, Citron, Clementine Grapefruit, Kumquat, Lemmon, Lime, Manderine, Orange, Pomelo, Tangerine	little	0,05
	Star apple, Cainito	little	0,05
	Azarole, Azzeruolo	little	0,05
	Longan, Lungan	little	0,05
	Persimmon	little	0,05

CITE THIS ARTICLE: Blili, Y. ., Nader, E. A. ., Pavez, I. ., Prosperi, P. ., Harbouze, R. ., Kyrgiakos, . L. S. ., Kleisiari, C. . ., Vasileiou, M. ., Angelopoulos, V. ., Vlontzos, G., & Kleftodimos, G. . (2025). Assessing The Economic Impact of Insect Pollination on the Agricultural Sector: A Department-Level Case Study in France. International Journal of Research in Organic Agriculture, 1(1), 17-38. <https://ijroa.com/ijroa/article/view/16>

International Journal of Research in Organic Agriculture
Volume 1, Issue 1, Year 2025

	Durian	great	0,65
	Loquat, Japanese plum, Japanese medlar	great	0,65
	Feijoa	great	0,65
	Fig	modest	0,25
	Strawberry	modest	0,25
	litchi, lychee	little	0,05
	apple	great	0,65
	Mammee	modest	0,25
	Mango	great	0,65
	Sapodilla	essential	0,95
	Medlar	unknown	unknown
	Rambutan	little	0,05
	Prickly pear	modest	0,25
	Passion fruit, Maracuja	essential	0,95
	Avocado	great	0,65
	Sapote, Mamey colorado	unknown	unknown
	Plum, Greengage, Mirabelle, Sloe	great	0,65
	Peach, Nectarine	great	0,65
	Sweet cherry	great	0,65
	Apricot	great	0,65
	Sour cherry	great	0,65
	Guava, Guayaba	modest	0,25
	Pomegranate	modest	0,25
	Pear	great	0,65
	Black currant, Red currant	modest	0,25
	Rose hips, Dogroses	great	0,65
	Raspberry, Blackberry, Clouderry, Northern Dewberry, Southern Dewberry	great	0,65
	Elderberry	modest	0,25
	Naranjillo	great	0,65
	Rowanberry	essential	0,95
	Service-apple	modest	0,25
	Hog plum, Mombin	little	0,05
	Tamarind	little	0,05
	Highbush blueberry, Lowbush blueberry, Rabbiteye blueberry, Bilberry	great	0,65

International Journal of Research in Organic Agriculture
Volume 1, Issue 1, Year 2025

	American cranberry, European cranberry	great	0,65
	Table grape, Vine grape	no increase	0
	Jujube	modest	0,25
	almond	great	0,65
	Cashew nut, and Cashewapple	great	0,65
	Peanut, Groundnut	little	0,05
	Brazil nut, Para nut, Cream nut	essential	0,95
	Chestnut	modest	0,25
	Macadamia	essential	0,95
	Mustard seeds	modest	0,25
	Rapeseed, Oilseed rape	modest	0,25
	Turnip rape, Canola	great	0,65
	safflower	little	0,05
	Coconut	modest	0,25
	Oil palm	little	0,05
	Soybean	modest	0,25
	Seedcotton	modest	0,25
	Sunflower seeds	modest	0,25
	Flaxseed	little	0,05
	Olive	no increase	0
	Sesame	modest	0,25
	Broad bean, Faba bean, Field bean, Horse bean	modest	0,25
	Karite nuts, Sheanuts	modest	0,25
	coffee	modest	0,25
	cola nut, kola nut	great	0,65
	cocoa	essential	0,95
	grains of paradise	unknown	unknown
	caraway	modest	0,25
	Coriander	great	0,65
	Cumin	great	0,65
	Cardamom	great	0,65
	Star anise	unknown	unknown
	Fennel seed	great	0,65
	Nutmeg	great	0,65
	Allspice, Pimento	great	0,65

International Journal of Research in Organic Agriculture
Volume 1, Issue 1, Year 2025

APPENDIX 2	Pepper	no increase	0
	Anise	unknown	unknown
	Vanilla	essential	0,95
	Sugar cane	no increase	0
	Maize, Green corn, Sweet corn	no increase	0
	Wheat	no increase	0
	Rice, Paddy	no increase	0
	Potato	increase-breeding	increase-breeding
	Sugar beet	no increase	0
	Cassava	increase-breeding	increase-breeding
	barley	no increase	0
	sweet potato	increase-breeding	increase-breeding
	Cabbage, Cauliflower	increase-seed production	increase-seed production
	Table Grape, Vine Grape	no increase	0
	Onion, Shallots, Welsh onion (green)	increase-seed production	increase-seed production
	Sorghum	no increase	0
	Yam	increase-breeding	increase-breeding
	Millet	no increase	0
	oat	no increase	0
	Carrot	increase-seed production	increase-seed production
	Lettuce, Chicory	increase-seed production	increase-seed production
	Bean dry like Kidney bean, Haricot bean, Lima bean, Azuki bean, Mungo bean, String bean	increase	increase
	rye	no increase	0
	Pineapple	increase-breeding	increase-breeding
	Garlic	increase-breeding	increase-breeding
	Triticale	no increase	0
	Spinach	no increase	0
	Taro (Coco Yam)	increase-seed production	increase-seed production
	Date palm	no increase	0
	Asparagus	increase-seed production	increase-seed production
	bean, green	increase	increase
	Mixed Grain	no increase	0
	Bamboo shoots	no increase	0
	Beets, Chards	no increase	0
	Capers	increase	increase

International Journal of Research in Organic Agriculture
Volume 1, Issue 1, Year 2025

	Cardoons	increase	increase
	Celery	increase	increase
	Chervil	increase	increase
	Cress	no increase	0
	Fennel	increase	increase
	Horseradish	increase	increase
	Sweet marjoram	unknown	unknown
	Oyster plant	increase	increase
	Parsley	increase	increase
	Parsnips	increase	increase
	Radish	increase	increase
	Rhubarb	no increase	0
	Rutabagas, swedes	increase	increase
	Savory	unknown	unknown
	Scorzonera	increase	increase
	Sorrel	no increase	0
	Tarragon	no increase	0
	Watercress	no increase	0
	Babaco	no increase	0
	Mangosteen	no increase	0
	Arracacha	increase	increase
	Arrowroot	increase	increase
	Chufa	no increase	0
	Sago palm	increase	increase
	Oca and ullucu	increase	increase
	Mashua	increase	increase
	Jerusalem artichoke	increase	increase