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Table of contents

| | |
|---|-----------|
| <i>DEDICATION</i> | II |
| ACKNOWLEDGEMENT | II |
| Abstract..... | III |
| Résumé | IV |
| خلاصة البحث | V |
| List of Abbreviations and Acronyms..... | VI |
| List of figures..... | IX |
| List of tables | XI |
| General introduction | 1 |
| Chapter I | 6 |
| Literature Review | 6 |
| I.Literature Review of Cereal Crop Monitoring in Precision Agriculture (PA) | 6 |
| I.1. Precision agriculture concept | 6 |
| I.1.1. Precision agriculture definition | 6 |
| I.1.2. Precision agriculture: the historical development of its meaning | 7 |
| I.1.3. Why precision agriculture? | 10 |
| I.1.4. Advantages and disadvantages of precision agriculture..... | 10 |
| I.1.4.1. The benefits of precision agriculture (advantages) | 10 |
| I.1.4.1.1. The global benefits of precision agriculture..... | 10 |
| I.1.4.1.1. Benefits of PA and its state in Algeria | 11 |
| I.1.4.2. Challenges and barriers to adoption (disadvantages) | 12 |
| I.1.4.2.1. Global challenges | 12 |
| I.1.4.2.1. Challenges to adopt precision agriculture in Algeria | 13 |
| I.2. Generalities on cereal crop | 13 |
| I.2.1. Origins and natural history of wheat | 16 |
| I.2.2. Presentation of the wheat crop | 17 |
| I.2.2.1. Scientific classification and nomenclature | 17 |
| I.2.2.2. Wheat characteristics: plant features and properties | 18 |
| I.2.2.2.1. The wheat grain | 19 |
| I.2.2.2.2. The wheat plant | 20 |
| I.2.2.3. Wheat growth stages: vegetative cycle | 21 |
| I.2.2.3.1. Why do we grow wheat? | 21 |
| I.2.2.3.2. How Does Wheat Grow?..... | 21 |

| | |
|---|-----------|
| I.2.2.4. Wheat requirements..... | 24 |
| I.2.2.4.1. Climatic requirements | 24 |
| I.2.2.4.2. Rainfall requirements | 25 |
| I.2.2.4.3. Soil requirements..... | 25 |
| I.2.3. Wheat production and trade in Algeria | 26 |
| II. Fundamentals and Related Works of Remote Sensing, Geographic Information System (GIS) and Artificial Intelligence (AI) in the Agricultural Field | 28 |
| II.1. Geographic information system in agriculture..... | 28 |
| II.2.1. Geoinformatics in agriculture | 30 |
| II.2.1.1. Introduction to geoinformatics..... | 30 |
| II.2.1.1.1. The historical background, a remarkable history of Geographic Information System (GIS) and geoinformatics..... | 31 |
| II.2.1.1.1.1. Where did it all begin? | 32 |
| II.2.1.1.2. Between GIS and geoinformatics: | 35 |
| II.2.1.2. Geoinformatics in precision agriculture:..... | 36 |
| II.2.2. GIS in agriculture..... | 36 |
| II.2.2.1. Definition of the Geographic Information System or GIS..... | 36 |
| II.2.2.2. GIS functions | 37 |
| II.2.2.2.1. Data capture | 37 |
| II.2.2.2.2. Data management..... | 37 |
| II.2.2.2.3. Spatial analysis..... | 37 |
| II.2.2.2.4. Results presentation | 38 |
| II.2.2.2.5. Agricultural GIS Data: Where It Comes from And How It's Processed? | 39 |
| II.2. Remote Sensing (RS) | 41 |
| II.2.1. Definition of RS | 42 |
| II.2.2. Historical overview | 42 |
| II.2.3. Remote Sensing..... | 44 |
| II.2.3.1. components of Remote Sensing..... | 44 |
| II.2.3.2. Main remote sensing modes of observation..... | 45 |
| II.2.3.2.1. Passive observation mode | 45 |
| II.2.3.2.2. Active observation mode | 46 |
| II.2.3.3. Electromagnetic radiation and the electromagnetic spectrum | 47 |
| II.2.3.4. Remote Sensing images | 48 |
| 1.Resolution's concept..... | 51 |
| 2.Types of remote sensing imagery | 51 |
| II.2.3.5. Operations on Remote Sensing images..... | 55 |

| | |
|---|----|
| II.2.3.5.1. Image acquisition | 56 |
| II.2.3.5.2. Image Pre-processing | 56 |
| II.2.3.5.1.1. Geometric Corrections | 56 |
| II.2.3.5.1.1. Radiometric and atmospheric corrections | 57 |
| II.2.3.5.4. Remote Sensing image processing | 57 |
| 1. Image Enhancement | 58 |
| 2. Image transformation by spectral indices | 58 |
| 3. Image segmentation | 59 |
| II.2.3.6. Applications of RS in agriculture | 61 |
| II.3. Ai for agriculture applications | 62 |
| II.3.1. Historical context of crop yield estimation techniques | 63 |
| II.3.2. Advances in remote sensing and GIS technologies | 63 |
| II.3.3. Application of artificial intelligence in agriculture | 64 |
| II.3.3.1. Definitions of AI | 65 |
| II.3.3.2. AI in agriculture | 65 |
| II.3.3.3. The process of AI adoption in agriculture | 66 |
| II.3.4. Machine learning in agriculture | 67 |
| II.3.3.4.1. Overview of machine learning applications in predicting yields | 67 |
| II.3.3.4.2. Machine Learning models applied in agriculture for yield estimation | 69 |
| Chapter II | 71 |
| Material and Methods: Problematic and resolving methods | 71 |
| II.1. Objectives | 71 |
| II.2. Problematic and motivation | 71 |
| II.3. Study area description | 73 |
| III.4. Machine learning techniques for crop yield estimation | 75 |
| III.4.1. Applications of AI in agriculture | 75 |
| III.4.2. definition of machine learning | 77 |
| III.4.3. Classification methods | 77 |
| III.4.3.1. What is a classification algorithm? | 77 |
| III.4.3.2. ML techniques | 78 |
| III.4.3.2.1. Supervised learning: | 78 |
| III.4.3.2.2. Unsupervised learning | 78 |
| III.4.3.3. Classification | 79 |
| III.4.3.4. Regression | 81 |
| III.5. Methodology and framework description | 82 |

| | |
|---|------------|
| III.5.1. Random Forest Classification..... | 83 |
| III.5.1.1. How RF Classification works..... | 83 |
| III.4.1.2. Flowchart of the Random Forest Classification Algorithm..... | 84 |
| III.5.1.3. Application in land use classification..... | 85 |
| III.5.2. RF regression..... | 86 |
| III.5.2.1. Flowchart of the RF regression..... | 86 |
| III.5.4. XGBoost classification..... | 87 |
| III.5.4.1. XGBoost definition..... | 87 |
| III.4.4.2. XGBoost algorithm for classification..... | 88 |
| CHAPTER III..... | 91 |
| Results and discussions..... | 91 |
| III.1. Experiments..... | 91 |
| III.1.1. Dataset description and preparation..... | 91 |
| III.1.1.1. Ground truth data..... | 91 |
| III.1.1.2. Imagery data..... | 94 |
| III.1.1.2.1. Sentinel-2 images..... | 94 |
| III.3.2.1. Landsat 8 Images:..... | 96 |
| III.1.2. Experimentation..... | 97 |
| III.1.3. Results and analysis..... | 105 |
| CHAPTER IV..... | 107 |
| BUSINESS PLAN..... | 107 |
| IV.1. The Origin of the Idea..... | 107 |
| IV.2. Motivations and Strengths..... | 107 |
| IV.3. Description of the Offer..... | 108 |
| IV.4. Sales Tools..... | 108 |
| IV.5. Project Launch Calendar..... | 109 |
| IV.6. SWOT Analysis..... | 112 |
| General Conclusion..... | 114 |
| References..... | 116 |
| Annexes..... | 127 |

Abstract

This thesis investigates the application of remote sensing and Geographic Information Systems (GIS) for estimating and forecasting wheat yields in the Tipaza region of Algeria. As agriculture faces increasing challenges from climate change and resource limitations, accurate yield estimation becomes critical for enhancing productivity and ensuring food security. Utilizing high-resolution satellite imagery from Sentinel-2 and Landsat 8, this study develops a comprehensive methodology that includes land use mapping, classification, and yield forecasting. The research begins with the acquisition of satellite images and ground truth data, which are processed using Google Earth Engine and QGIS to create detailed land use maps. Classification algorithms, specifically Random Forest and XGBoost are employed to categorize agricultural practices effectively. The classification results are validated against real yield data and agricultural statistics from the Algerian Directorate of Agricultural Services, ensuring the reliability of the findings. Following validation, Random Forest regression and XGBoost regression techniques are applied to predict wheat yields based on the classified land use data. Additionally, the study compares yield dynamics across other crops, including soft wheat, soybeans, and oats, providing a broader perspective on agricultural productivity in the region. The findings demonstrate that remote sensing technologies can significantly enhance yield estimation accuracy and contribute to sustainable agricultural practices. This research not only advances our understanding of wheat production in Tipaza but also proposes a predictive framework that can be adapted for wider application throughout Algeria. By integrating advanced technologies into agricultural monitoring, this thesis aims to support informed decision-making for farmers and policymakers, ultimately contributing to improved food security and resource management in the region.

Key words: geographic information system, remote sensing, sustainability, yield estimation, Wheat.

Résumé

Cette thèse examine l'application des technologies de télédétection et des systèmes d'information géographique (SIG) pour estimer et prévoir les rendements de blé dans la région de Tipaza en Algérie. Alors que l'agriculture fait face à des défis croissants dus au changement climatique et aux limitations des ressources, une estimation précise des rendements devient essentielle pour améliorer la productivité et garantir la sécurité alimentaire. En utilisant des images satellites à haute résolution provenant de Sentinel-2 et de Landsat 8, cette étude développe une méthodologie complète qui comprend la cartographie de l'utilisation des terres, la classification et la prévision des rendements. La recherche commence par l'acquisition d'images satellites et de données de terrain, qui sont traitées à l'aide de Google Earth Engine et de QGIS pour créer des cartes d'utilisation des terres détaillées. Des algorithmes de classification, tels que Random Forest et XGBoost sont employés pour catégoriser efficacement les pratiques agricoles. Les résultats de la classification sont validés par rapport aux données réelles de rendement et aux statistiques agricoles fournies par la Direction des services agricoles algérienne, garantissant ainsi la fiabilité des conclusions. Après validation, les techniques de régression Random Forest et XGBoost sont appliquées pour prédire les rendements du blé en fonction des données d'utilisation des terres classifiées. De plus, l'étude compare les dynamiques de rendement d'autres cultures, notamment le blé tendre, l'orge et l'avoine, offrant ainsi une perspective plus large sur la productivité agricole dans la région. Les résultats démontrent que les technologies de télédétection peuvent considérablement améliorer l'exactitude des estimations de rendement et contribuer à des pratiques agricoles durables. Cette recherche non seulement fait progresser notre compréhension de la production de blé à Tipaza, mais propose également un cadre prédictif qui peut être adapté à une application plus large à travers l'Algérie. En intégrant des technologies avancées dans le suivi agricole, cette thèse vise à soutenir une prise de décision éclairée pour les agriculteurs et les décideurs politiques, contribuant ainsi à améliorer la sécurité alimentaire et la gestion des ressources dans la région.

Mots clés : Blé, durabilité, Estimation du Rendement, Télédétection, Système d'Information Géographique (SIG)

خلاصة البحث

تتناول هذه الأطروحة تطبيق تقنيات الاستشعار عن بعد ونظم المعلومات الجغرافية لتقدير وتوقع إنتاج القمح في منطقة تيبازة الجزائرية. مع مواجهة الزراعة تحديات متزايدة نتيجة التغير المناخي وقيود الموارد، تصبح تقديرات الانتاج الدقيقة أمرًا ضروريًا لتعزيز الإنتاجية وضمان الأمن الغذائي من خلال استخدام صور الأقمار الصناعية عالية الدقة من Sentinel-2 Landsat-8، تطور هذه الدراسة منهجية شاملة تشمل رسم خرائط استخدام الأراضي، والتصنيف، وتوقع الإنتاج. تبدأ الدراسة بجمع صور الأقمار الصناعية وبيانات الأرض، والتي تتم معالجتها باستخدام GOOGLE EARTH engine و QGIS لإنشاء خرائط تفصيلية لاستخدام الأراضي يتم استخدام خوارزميات التصنيف مثل Random Forest و XGBoost، لتصنيف الممارسات الزراعية بشكل فعال. يتم التحقق من نتائج التصنيف من خلال مقارنة البيانات الحقيقية للإنتاج والإحصاءات الزراعية المقدمة من المديرية العامة للخدمات الزراعية الجزائرية، مما يضمن موثوقية النتائج. بعد التحقق، يتم تطبيق تقنيات الانحدار XGBoost و Random Forest لتوقع إنتاج القمح بناءً على بيانات استخدام الأراضي المصنفة. بالإضافة إلى ذلك، تقارن الدراسة طرق الإنتاج لزراعات أخرى، بما في ذلك القمح اللين والشعير. مما يوفر منظورًا أوسع لإنتاجية الزراعة في المنطقة. تظهر النتائج أن تقنيات الاستشعار عن بعد يمكن أن تعزز بشكل كبير دقة تقديرات الإنتاج وتساهم في ممارسات زراعية مستدامة. لا تساهم هذه الدراسة فقط في تعزيز فهمنا لإنتاج القمح في تيبازة، بل تقترح أيضًا إطارًا تنبؤيًا يمكن تكيفه للاستخدام الأوسع في جميع أنحاء الجزائر. من خلال دمج التقنيات المتقدمة في مراقبة الزراعة، تهدف هذه الأطروحة إلى دعم اتخاذ قرارات مستنيرة للمزارعين وصانعي السياسات، مما يساهم في تحسين الأمن الغذائي وإدارة الموارد في المنطقة.

مفاتيح البحث

الاستشعار عن بعد، القمح، تقدير الإنتاج، نظام المعلومات الجغرافية