

Regulatory Challenges in AI-driven Credit Card Applications

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Abstract: AI is one of the most important and revolutionary steps for the financial industry. The research is about the regulatory challenges in AI-driven credit card application which consists of different factors data privacy and security, fair lending and bias mitigation, transparency requirements, accountability and liability concerns, and cross-border regulatory differences in some of the countries. The design of the research is explanatory and the data collection method is a secondary qualitative and quantitative method. The discussion is also based on different authors to highlight the regulatory compliances such as GDPR to protect the data of customers.

Keywords: AI-driven credit card, GDPR, CCPA, ECOA, Fair Lending laws, Privacy laws

I. INTRODUCTION

A. Background to the Study

AI-driven credit card applications have changed the game for the financial industry and revolutionised personalised credit limits, fraud detection, and automated approvals. There are also regulatory challenges that impact their implementation and adoption. Data privacy is one of the major concerns for AI-driven credit card applications because it consume large amounts of data of consumers which raises concern over data protection [1]. To mitigate this concern, configuration

regulations and compliance with privacy laws are necessary such as GDPR and CCPA. GDPR is the privacy law of Europe and CCPA is the privacy law of the USA.

B. Overview

AI-driven credit card applications are very useful for the financial industry because they revolutionise the industry completely and also help in faster and more efficient credit decisions, improve fraud detection and security, enhance credit risk assessment, and many more. On the other hand, this are also some regulatory challenges that are covered in this research such as data privacy, bias mitigation, transparency, accountability and liability concerns, and cross-border regulatory differences [2]. Through the help of objectives and literature review, the research will be discussed in a detailed manner and the regulatory challenges will be explained.

C. Problem Statement

The financial industry has undergone a revolution in credit risk evaluation by bettering credit risk evaluation, monitoring frauds, as well as providing good customer experience as AI-based credit card applications have widely been used. Despite such advancement, which is indeed happening, it is proceeding with many challenges, particularly regarding issues of compliance, data privacy, algorithmic bias and transparency. Financial institutions have to deal with absurd amounts of regulations

such as GDPR, CCPA, the Fair Lending Laws and so on, while still maintaining the security and explainability of the AI models [3]. This research aims to understand these regulatory challenges as well as tips on how one can balance innovation and compliance while designing AI-enabled credit card applications.

D. Objectives

Aim

To evaluate the regulatory challenges in AI-driven credit card applications and analyse the different strategies to ensure compliance in innovation in the financial sector.

Objectives

1. To analyse the impact of data privacy regulation on AI-driven credit card application procedures.
2. To address the problem of bias mitigation, fairness, and explainability in credit card credit decisions of AI systems as per the finance regulations.
3. To evaluate competing strategies for financial firms in cross border dealings to comply with regulatory difference, without losing their operational efficiency and its nexus to AI.

E. Scope and Significance

The scope of the research to analyse the regulatory challenges created by applications of credit cards by means of AI such as privacy of data, bias of the algorithmic, transparency and trying to overcome the compliance issues related to cross borders. Additionally, the study is also on the regulations such as GDPR, CCPA, ECOA and others and how the AI based Credit Assessment effects the same. The research also looks into how the financial institutions can create fair, explainable, and compliant AI systems that are fair and contagious, and able to function optimally and operationally [4]. It is also important for the research itself as there's a lot that covers important aspects of the study in the landscape of regulatory challenges.

II. LITERATURE REVIEW

Data privacy and security in AI-driven Credit Card applications

According to Faheem (2021), the transformation of AI on credit scoring which entails AI usage in risk assessment models for credit scoring and in the prediction of credit scoring. It also looks into the integrated use of machine learning, processing of live data, and alternative data sources that can enhance credit worthiness of people. AI-driven credit card applications have the concern of data security privacy and security and to figure out these challenges, the study also explores the need for transparent and explainable AI models and data governance by highlighting the laws on different financial organisations. Enforcing laws such as GDPR is one of the major solutions for protecting the data and privacy of customers regarding the issues of data privacy which can be important for the financial institutes to retain their reputation in the market. The findings of the study are the privacy laws are important to protect the data of consumers regarding AI-driven credit card applications.



Figure 2: Biometric Authentication
[9]

According to Patra *et al* (2019), the authentication schemes have also evolved to be multi-modal such as the power spectrum of handwriting and combining fingerprints or validating face and signature to give a better level of assurance to the biometric authentication. The “big data paradigm” enables storing and managing large data efficiently and applying AI is one of the most important factors because it gives another security layer to overall systems. To protect data security by AI and big data, compliances such as GDPR are helpful to protect and secure data.

According to Gomber *et al* (2018), different firms in the financial service industry have been faced with the dramatic and relatively recent emergence of new technology innovations, and process disruptions. The study focuses on the technology innovations that have begun to leverage the execution value associated with payment settlements, blockchain technologies, AI technologies, and also cross-border payment services. The study explores global fintech companies using different AI tools that can navigate different landscapes of regulation in multiple countries. On the other hand, the EU, China, and the US have different AI and privacy laws that can be followed by companies to align AI systems.

III. METHODOLOGY

A. Research Design

This study's research design framework utilizes the collected data to tackle and establish the investigation's trajectory. This study employed an *explanatory research design* which emphasises different factors related to the research which is based on the regulatory challenges in AI-driven credit card applications. This research design helps to find the method that explores something that occurs when limited information is available. This research design aligns with the research of AI-driven credit card applications to increase the understanding of limited information.

B. Data Collection and Analysis

The data collection method that is used for this research is “secondary qualitative and quantitative” data collection. The qualitative data method explores different sources such as industry reports, journals, articles, websites, and many more. On the other hand, the quantitative data method explores the different graphs and charts that can evaluate the information and data related to AI-driven credit card applications. All the data and pieces of information are collected from accurate and valid sources which gives accurate information and data related to research.

C. Case Studies/Examples

Case Study 1: Apple Card (Goldman Sachs): Allegations of Gender Bias

Goldman Sachs issued an Apple Card and faced a lot of backlash in the year 2019 because of gender bias. The women claimed that the company offered low credit limits to women as compared to men with the same financial profile [11]. This case raised concerns under the “Equal Credit Opportunity Act (ECOA)” and “Fair Lending Laws”. The result that occurred with this case is that AI-based credit decisions may be biased sometimes. The outcome of this backlash made Goldman Sachs review their

credit assessment model and later denied it by the company.

Case Study 2: American Express: AI and Data Privacy Regulations

American Express accesses the credit scoring of consumers as per their previous history and behaviour of their transactions and for this, the company uses AI-driven credit scoring models. The main issues regarding this system are data security and the privacy of customers. To manage these risks the company align with regulatory compliances such as GDPR and CCPA regulations which help them to protect the data of their customers and also ensure the brand reputation [12]. The company adopted the explainable AI model and improved communication with customers regarding AI-driven credit assessment by integrating privacy-first AI frameworks. This can secure the data of global consumers.

Case Study 3: ZestFinance and AI Transparency in Credit Scoring

ZestFinance is a credit scoring company which assumes the credit limit through the help of different AI algorithms. The company faced an issue regarding its machine learning scrutiny which is used in assessing creditworthiness. The company also follow the regulations and laws to protect the data such as GDPR and “Consumer Financial Protection Bureau (CFPB)” guidelines to make the use of AI decisions which are non-discriminatory [13]. The company adopted the explainable AI techniques to enhance the transparency in credit decisions which are followed through the help of different compliances.

IV. RESULTS

A. Data presentation

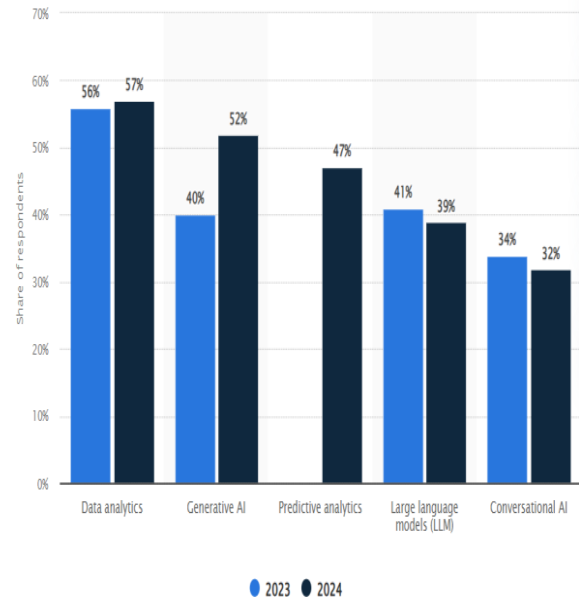


Figure 3: AI workloads in financial services worldwide

[14]

AI is generally used in financial services in different forms such as data analytics, generative AI, predictive analysis, large language models, and conversational AI. In the year 2022, data analytics was used by financial services by 56% and in the year 2022 used by 57%. On the other hand, Generative AI was used in the year 2022 was 52% and in the year 2022 was 40%. Predictive analytics was used by the financial sector in the year 2022 was 47% which shows the overall dependence of financial sectors on predictive analytics [14].

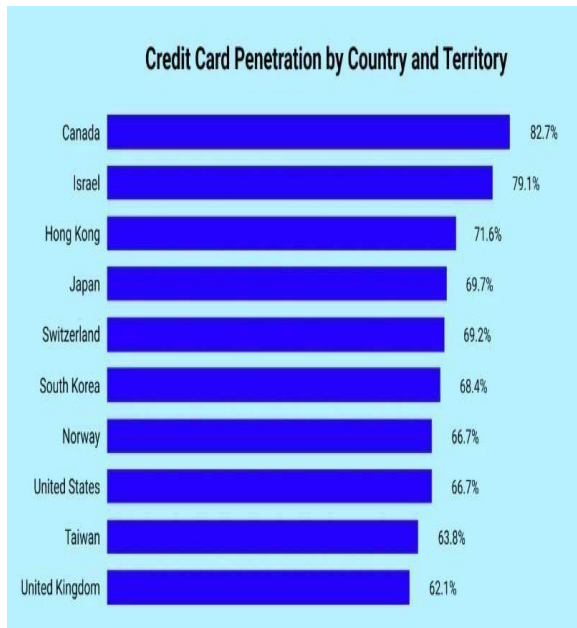


Figure 4: Credit Card Penetration by Country [15]

The most used credit card by country is Canada and 82.7% of payments have been down from the credit cards which shows that consumers are highly dependable on the credit cards for their payment. On the other hand, Israel ranks second and 79.1% of the payment has been done through credit cards [15]. These results highlight that users in Canada and Israel are highly dependent on credit cards for payment.

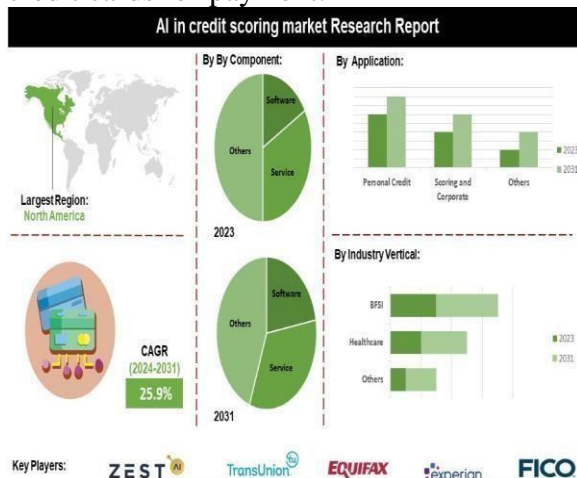


Figure 5: AI in credit scoring market [16]

Figure 5 highlights the use of AI in the credit scoring market that consists of multiple

strong players in the market such as Zest, TransUnion, Equifax, Fico, and many more. The CAGR is valued at 25.9% from the year 2022 to 2031 [16]. The key players introduce AI highly on their scoring methods such as by transaction behaviour of the customers, transaction history, and many more.

B. Findings

The findings of the study highlight the importance of AI in the financial sector or industry. AI is used in financial sectors in different forms to enhance their operations such as using credit cards for scoring purposes through the help of transaction history and consumer behaviour. Canada and Israel are the top two countries where the payment method is highly dependable through the help of credit cards. On the other hand, key players such as Zest, Fico, and many more use AI in credit scoring which helps them to know the limit barriers. The credit-scoring market is also expected to grow by 25.9% [16].

C. Case Study Outcomes

Case Study	Key Findings	Relevance
Case Study 1: Apple Card (Goldman Sachs): Allegations of Gender Bias	AI showed gender bias, violating ECOA & Fair Lending Laws [11].	Fair AI models and bias mitigation in credit decisions.
Case Study 2: American Express: AI and Data Privacy Regulations	Adopted explainable AI & GDPR/CCPA compliance.	Data security, transparency, and compliance in AI credit models [12].
Case Study 3:	Lack of AI explainability	Transparent AI models

ZestFinance and AI Transparency in Credit Scoring	y	
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Table 1: Case Study Outcomes

(Source; Self-developed)

Table 1 highlights the key findings and relevance of the case studies. The key findings for case study 1 cover the ECOA and fair lending laws and the relevance of constructing fair AI models. On the other hand, to mitigate data security America expresses compliance with GDPR. Lastly, ZestFinance lacks AI explainability and focuses on transparent AI models to improve customer services.

D. Comparative Analysis

<i>Authors</i>	<i>Focus Area</i>	<i>Key Findings</i>	<i>Limitations</i>
[5]	AI-driven credit scoring and data privacy [5].	GDPR helps protect consumer data	AI risks beyond privacy concerns
[6]	AI in data security for credit card applications	Privacy laws such as GDPR & CCPA enhance cybersecurity	Lacks empirical data
[7]	Bias, fairness, and explainability in AI credit decisions [7].	AI bias can impact financial risk	Lacks real-world case studies

[8]	AI algorithms in credit decisions	Detects fraud through pattern analysis	Lack of AI implication decisions
[9]	AI and biometric authentication	Multi-modal authentication	Limited focus on AI-driven credit application
[10]	Global AI regulations and cross-border compliance	AI adoption varies by region	Lack of detailed comparison of privacy laws [10].

Table 2: Comparative Analysis

(Source: Self-developed)

Table 2 highlights the presentation of different authors related to research. Some author focuses on GDPR and other compliances that help to mitigate the risk of data security and some of the author focuses on AI bias which can impact the risk of finance. On the other hand, the discussion is also based on cross-border compliance meanwhile, lacks a detailed comparison of laws related to privacy.

V. DISCUSSION

A. Interpretation of Results

AI is one of the most used tools in financial sectors that helps to enhance the efficiency of the finance sector. AI-driven credit card application also comes with different regulatory challenges such as data privacy and security compliances, fair lending and bias mitigation, explainability and transparency requirements, accountability and liability concerns, and lastly cross-border regulatory compliance. On the other hand, the financial sector heavily relies on AI to

measure credit score by observing transactional behaviour and transactional history. The AI credit scoring market is also growing at a CAGR of 25.9% [16]. Some of the key players using AI to measure credit scores such as Zest, Fico, and many more.

B. Practical Implications

The research focuses on the regulatory challenges in AI-driven credit card applications which consist of different factors to solve these issues. Implementing transparent and fair processes in AI can evaluate the mitigation process such as by combining with different regulations such as GDPR, CCPA, and Fair Lending Laws that protect the data of consumers. American Express is the best example of implementing AI explainability and robust fraud detection to comply with GDPR to enhance their services [12]

C. Challenges and Limitations

The challenge and limitations of the secondary data collection method consist of different factors. Lack of data reliability, outdated information and data, lack of informative insights related to research and many other things [17]. Lack of proper real-world examples also missed some important aspects of different privacy laws in cross-border compliances can weaken the research. The cross-verification and ensuring the credibility of different sources is also important to address the limitations of research.

D. Recommendations

The research should be more transparent and explore the explainable AI model's transparency in different credit decisions and regulatory compliances. On the other hand, the study should also focus on the interviews and case studies of financial institutions that help to identify the limitations of secondary data in future studies [18]. AI security framework should be added in the research because this will help develop robust data protection models.

VI. CONCLUSION AND FUTURE WORK

The regulatory challenges are centred around the issue of data privacy, bias, transparency and cross-border compliance and focus on this research. Although the credit risk assessment is assisted by the AI, the legal, ethical and many more parts of the rights of AI are different. The compliant AI model has demands of explainability, compliance, and fairness, which financial institutions have trust and their obligation to comply with various rules around the world.

Future studies can be done using AI security frameworks, as well as real-world case studies and the policies which can be used for the collection of primary data. Therefore, the effort to boost the adoption of responsible AI in financial services can be enhanced by paying more importance to the role of adaptive compliance capabilities built on evolving laws and AI bias mitigation techniques.

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