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## Revealing Physicians Referrals from Health Insurance Claims Data<sup>☆</sup>

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### ABSTRACT

Health insurance companies in Brazil have their data about claims organized having the view only for service providers. In this way, they lose the view of physicians' activity and how physicians share patients. Partnership between physicians can be seen as fruitful, when they team up to help a patient, but could represent an issue as well, when a recommendation to visit another physician occurs only because they work in same clinic. This work took place during a short-term project involving a partnership between our lab and a large health insurance company in Brazil. The goal of the project was to provide insights (with business impact) about physicians' activity from the analysis of the claims database. This work presents one of the outcomes of the project, i.e., a way of modeling the underlying referrals in the social network of physicians resulting from health insurance claims data. The approach considers the flow of patients through the physician-physician network, highlighting connections where referrals between physicians potentially occurred. We present the results from the analysis of a claims database (detailing 18 months of activity) from the health insurance company we partnered with. The main contribution presented in this paper is the model to reveal *mutual referrals* between physicians. Results show the proposed model reveals underlying characteristics of physicians' activity from real health insurance claims data with multiple business applications.

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### 1. Introduction

Health insurance costs are a main issue of concern in almost every country in the world as budget constraints impact directly on the quality of the service. As a result, health insurance companies have been extensively trying to reach a trade-off between offered services and costs as a way to meet budget constraints.

One way for health insurance companies to address those issues is to better understand the complex relationships among the diverse participants of the healthcare systems, including patients, physicians, hospitals, and other service providers. To support this quest, healthcare insurance companies and other health service providers have often a wealth of data from their own operations at their disposal, especially transactional data.

In the case of health insurance companies, an important piece of transactional data involves the claims presented by their ecosystem of providers. In the present work, a claim represents a report from a physician or a healthcare service provider to a health insurance company requesting some form of fee related to a patient's

consultation with a physician, a clinical exam, or a medical procedure. Even though claims data may vary, it generally contains at least the ID of the healthcare professional involved in the procedure (it may also be a group of professionals), the ID of the patient, the type of procedure, and time information related to the event. It may include other types of information such as location of the service, pre-authorization codes, etc.

Traditionally the analysis of claims data is based on applying statistics and Data Mining methods to the individual elements of the system (physicians, service providers, patients) or to the set of claims. However, healthcare is often provided by collaborative teams of physicians, nurses, and technicians which are connected to each other by often strong professional relationships. Physicians that refer patients to other physicians have clear preferences about who they want to team up with for specific procedures and often are involved in master-apprentice structures. Physicians also have preferences for specific service providers such as hospitals and clinical analysis laboratories. Those recommendations could be good for building patient trust or indicate a fraud when this is not the patient's will. Similarly, patients establish bonds of trust and reliance with specific physicians or group of physicians.

In practice, mining claims is difficult because claims are paid to a wide variety of providers, such as hospitals, clinics, or even

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1 physicians registered as small companies. A single patient can con- 67  
 2 sult a physician through all those channels. It could be even diffi- 68  
 3 cult to know who exactly is taking care of a given patient, since 69  
 4 that there are cases in which the physician ID used in a claim is 70  
 5 from a professional registered in the health care provider system, 71  
 6 but the one taking care of the patient is an unregistered physician. 72  
 7 Moreover, claims contain no information about referrals and of- 73  
 8 ten the connections among service provider team members are not 74  
 9 recorded explicitly in claims. It is also well known that claims data 75  
 10 is riddled with errors and unreliable information. Despite all those 76  
 11 difficulties, we show in this paper that meaningful and reliable in- 77  
 12 sights about the flow of patients through the network of physicians 78  
 13 and how physicians refer each other can be inferred from claims 79  
 14 data. 80

15 This study took place during a partnership, in a short-term 81  
 16 project, between our lab and a major Brazilian health insurance 82  
 17 company involving the analysis of their claims database. The main 83  
 18 challenge the insurance company brought to our team was to iden- 84  
 19 tify physicians that excel at medicine by using the claims database. 85  
 20 After decomposing this challenge considering the health insurance 86  
 21 workflow, the following components were identified as key factors 87  
 22 for outstanding professionals, defined by the health insurance com- 88  
 23 pany: physicians referred by peers, relative importance in the net- 89  
 24 work of physicians, and returning behavior of patients. The need 90  
 25 for modeling referrals emerged during interactions with the health 91  
 26 insurance company staff as a way of identifying physicians that 92  
 27 excel in a certain specialty and are referred by their peers. These 93  
 28 interactions occurred weekly, fomenting discussions between our 94  
 29 team and subject matter experts, IT specialists, analytics team, and 95  
 30 the superintendent of the health insurance company. Hence, this 96  
 31 work aims at presenting a way of modeling mutual referrals in 97  
 32 a physician–physician network, which connects to the hypothesis 98  
 33 studied in this work: *H1) It is possible to identify underlying physi- 99  
 34 cians' referrals from claims data.* 100

35 The main contribution of this work is a way of modeling mutual 101  
 36 referral patterns in the physician–physician network. The method 102  
 37 can therefore be used by health insurance companies to bet- 103  
 38 ter manage the physicians they have businesses with, nurturing 104  
 39 the experience of registered physicians and inviting unregistered 105  
 40 physicians that collaborate with registered ones. It can also be used 106  
 41 to support patients to receive more integrated care from a group 107  
 42 of physicians and service providers. 108

43 This paper is organized as follows: section 2 describes the 109  
 44 related work, section 3 details the database analyzed, section 4 110  
 45 presents the proposed model for highlighting mutual referrals, sec- 111  
 46 tion 5 discusses the obtained results, and section 6 concludes and 112  
 47 points to future works. 113

## 48 2. Related work 114

49 Healthcare data is heralded as the key element in the quest 115  
 50 to improve efficiency and reduce costs in healthcare systems [1]. 116  
 51 This trend is becoming more pronounced as multi-scale data gen- 117  
 52 erated from individuals is continuously increasing, particularly due 118  
 53 to new high-throughput sequencing platforms, real-time imaging, 119  
 54 and point-of-care devices, as well as wearable computing and mo- 120  
 55 bile health technologies [2]. 121

56 In healthcare, data heterogeneity and variety arise as a result 122  
 57 of linking the diverse range of biomedical data sources available. 123  
 58 Sources of quantitative data (e.g., sensor data, images, gene arrays, 124  
 59 laboratory tests) and qualitative data (e.g., diagnostics, free text, 125  
 60 demographics) usually include both structured and unstructured 126  
 61 data, normally under the name of Electronic Health Records. Addi- 127  
 62 tionally, the possibility to process large volumes of both structured 128  
 63 and unstructured medical data allows for large-scale longitudinal 129  
 64 studies, useful to capture trends and to propose predictive mod- 130  
 65 els [3]. 131  
 66 One of the most useful and commonly used datasets are claims 132

67 studies, useful to capture trends and to propose predictive mod- 67  
 68 els [3]. 68

69 One of the most useful and commonly used datasets are claims 69  
 70 databases. Claims data records are often rich in details, as they 70  
 71 describe important elements of the events taking place around 71  
 72 the healthcare professional and the patient, e.g., timestamps, ge- 72  
 73 ographical location, diagnosis codes, associated expenses, among 73  
 74 others. The use of claims data in healthcare studies has been scru- 74  
 75 tinized in [4] and [5], providing a set of good practices and outlin- 75  
 76 ing the shortcomings of claims-based research. 76

77 Social network analysis has proven to be a useful analysis tool 77  
 78 in this context, allowing for insights difficult to reach by traditional 78  
 79 descriptive statistics as presented in [5]. For instance, social net- 79  
 80 work analysis has been used to study comorbidity, the simultane- 80  
 81 ous presence of two chronic diseases or conditions in a patient. By 81  
 82 structuring diseases as a network, it is possible to quantify some of 82  
 83 the aspects of the complex interactions between conditions in the 83  
 84 different patient populations. A number of studies have focused on 84  
 85 extensive claims datasets to examine and understand comorbidity 85  
 86 networks. In [6], the authors study a diffusion process on a com- 86  
 87 morbidity network to model the progressive spreading of diseases 87  
 88 on a population depending on demographic data. In [7], the au- 88  
 89 thors study how a given chronic disease (diabetes) correlates with 89  
 90 age and gender, spanning almost 2 million patients from an en- 90  
 91 tire European country. Such comorbidity networks have also been 91  
 92 proposed as models to understand the connection between genetic 92  
 93 and environmental risk factors for diseases [8]. 93

94 Beyond clinical purposes, claims data have also been studied 94  
 95 to understand the complex interactions of different organizational 95  
 96 structures and management relationships involved in patient care 96  
 97 processes. For instance, in [9], temporal patterns in electronic 97  
 98 health records were modeled in order to present useful informa- 98  
 99 tion for decision-making. The authors developed a data represen- 99  
 100 tation for knowledge discovery so as to extract useful insights on 100  
 101 latent factors of the different processes involving a patient, aiming 101  
 102 at improving workflows. 102

103 Another important trend is the understanding of the relation- 103  
 104 ship among healthcare professionals, in particular the physicians. 104  
 105 In [10], the authors apply social network analysis to mine networks 105  
 106 of physicians which might be used to improve the designation 106  
 107 of middle-sized administrative units (accountable care organiza- 107  
 108 tions). Sauter et al. [11] use social network analysis to understand 108  
 109 networks of healthcare providers which share patients, providing 109  
 110 insights in the interplay between general practitioners, internal 110  
 111 specialists, and pediatricians. Also, the network structure of differ- 111  
 112 ent healthcare providers taking care of a given individual can show 112  
 113 important variability of the healthcare system [12]. 113

114 Social networks have also been used to understand the state of 114  
 115 coordination of healthcare actors. In [13], the authors describe a 115  
 116 complex network approach applied to health insurance claims to 116  
 117 understand the nature of collaboration among physicians treating 117  
 118 in-hospital patients and to explore the impact of collaboration on 118  
 119 cost and quality of care. Also, in [14], the authors study the social 119  
 120 network structure in hospitals among healthcare professionals to 120  
 121 understand which variables affect patient care efficiency measures. 121  
 122 The idea is further developed in [15] from a statistical point of 122  
 123 view in a medium-sized number of hospitals, through the analysis 123  
 124 of temporal patterns and costs. 124

125 The medical referral system in the Canadian healthcare system 125  
 126 is studied in [16], where the authors map and analyze the network 126  
 127 between general practitioners and specialists. In [17], the authors 127  
 128 describe the condition of the basic medical insurance for urban 128  
 129 and rural residents in China, then they demonstrate that social net- 129  
 130 work analysis can be used in the health insurance claims data to 130  
 131 support better understanding of patients transfers among hospi- 131  
 132 tals. 132

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