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Social Capital or Education: What Matters Most to Cut Time to Diagnosis?

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Abstract

Time to diagnosis, defined as the time span from first symptoms to final diagnosis, has received little if no attention, although it is perceived as highly variable across conditions, patients and countries and as a key determinant of health prognoses and outcomes. In this paper, we offer one of the first measures of time to diagnosis for four chronic conditions (bipolar trouble, Crohn disease, multiple sclerosis and psoriasis), and analyze the role played by patients education and social networks in explaining time to diagnosis. Adopting a patient's perspective, we use self-reported data from an online open access questionnaire administered to a large French social network of patients with chronic conditions. Duration models are used to explain variations in time to diagnosis. Our findings suggest that social participation and social support indeed reduce the probability of experiencing longer time spans to diagnosis. But contrary to expectations, higher levels of education have the reverse effect. We further analyze these results by identifying differences in patients' health care-seeking behavior: more educated patients tend to consult specialists first, which leads to longer time spans to diagnosis as they are less prone than GPs to referring patients to hospitals for additional tests, when needed. While our social networks findings support WHO's recommendations to enhance individual social capital, results on education provide support for reforms aimed at implementing GP referral systems.

Keywords: Chronic disease, Diagnosis, Education, Health inequalities, Social capital

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1. Introduction

Delays in diagnosis may have detrimental outcomes in terms of health and avoidable health expenditures. Despite innovative diagnostic technologies, and the standardization of diagnosis and treatment protocols, the time elapsed between first symptoms and final diagnosis varies substantially across patients. The French survey Erradiag, conducted in 2016, showed that 25% of patients treated for a rare disease assessed time to diagnosis to be over 5 years¹. Nearly 60% of patients declared that lack of diagnosis led to physical harm, psychological distress, and avoidable medical treatments. Beyond rare diseases, several medical studies conducted on cancer patients have shown that time to diagnosis is negatively correlated to survival time [1, 2, 3]. The pre-diagnosis period has important consequences on prognosis and disease progression and if linked to social status, may play an important role in the construction of social health inequalities. However, it has so far received very little, if no attention in health economics and in public health studies, due to lack of data. In this paper, we study the pre-diagnosis stage of illness, from the patients perspective. Our main variable of interest is time to diagnosis defined as the time elapsed between first symptoms appearance and final diagnosis. Variations in time to diagnosis may reflect supply side effects (the health care systems responsiveness to symptoms' onset), as well as demand side effects, driven by patients characteristics and health care seeking behavior. We describe time to diagnosis among patients treated for four chronic diseases, bipolar trouble, Crohn disease, multiple sclerosis and psoriasis, to then explore socioeconomic factors that may explain differences in time to diagnosis, with an emphasis on individual social capital and education, as these variables have been shown to have important causal effects

¹Erradiag survey conducted by Alliance Maladies rares in 2016. Report available online at: <http://fr.calameo.com/read/003972817bb7d085cce09>

on health status, health behaviors, and health utilization behaviors [4, 5, 6].

We base our research on an online survey administered to patients belonging to a chronic conditions social network. The results show that individual characteristics, among which social capital and education, have strong and significant impacts on time to diagnosis, shedding a new light on the construction of social health inequalities. The paper is organized as follows. Section 2 reviews the literature on diagnosis work-up, and on the channels through which social capital and education may influence time to diagnosis. Section 3 introduces the questionnaire and the data. Section 4 presents the statistical methods, section 5 presents our main results which we are discussed in section 6.

2. Literature Review

This study aims at investigating how education, social capital and individual preferences may affect time to diagnosis and how the pre-diagnosis period can contribute to the construction of social inequalities in health. In this section, we review the existing literature on time to diagnosis and the links between health and health care utilization.

The time elapsed between first symptoms and first medical encounter varies with symptoms specificity and illness severity. If the symptoms are nonspecific², the patient may interpret them as transient episodes of tiredness or anxiety. On the contrary, severe symptoms, sometimes combined with disabilities, will spur patients' decision to seek care, thereby reducing time to first contact with a health care professional[7]. After the patients first medical encounter, a thorough examination is critical in obtaining a full diagnosis. Health care professionals must choose the optimal diagnosis strategy, weighing the benefits

²A nonspecific symptom does not allow a unique disease identification, because it either affects multiple organs at the same time, or the body as a whole, such as pain or fatigue.

and costs associated with additional tests, in particular patients' costs (invasive procedures, anxiety) [8]. The time elapsed from the first medical encounter to the full diagnosis will also vary according with the nature of the symptom (specific or not) and the disease stage and form: illnesses such as multiple sclerosis or lupus, are characterized by relapsing-remitting forms, for which symptoms may suddenly disappear. In complex or rare diseases, diagnosis is harder to obtain and errors are more frequent. In cases where symptoms are common to more than one disease, health professional may fail to identify the proper illness. From the results of the survey conducted in 2012 by the French Observatory for rare diseases³, 90% of health care professionals lack knowledge on rare diseases. Moreover, due to increasing specialization, health care professionals, when faced with multiple conditions, may fail to adopt a global approach to patients health. Indeed, while increased specialization may have positive effects on health care and health outcomes for diseases within the area of specialty, it has been shown to lead to coordination failure when multiple medical disciplines are involved [9].

Patients characteristics have been shown to influence health care seeking behaviour. An extensive literature has established the link between education and health, often referred to as the health education gradient. Results indicate that more educated people have healthier lifestyles, a better health status, and a higher life expectancy [10, 11, 12]. They show higher competencies to gain access, understand and use information in ways that promote and maintain good health often referred to as health literacy [13, 14]. The more educated tend to adopt healthier life styles, adhere and comply to medical decision and treatment, and reap the benefits of improved medical technology [15, 16] and health

³Survey conducted by the French Observatory for rare diseases in 2012. Report available online at:
http://www.maladiesraresinfo.org/assets/pdf/Rapport_Observatoire_maladies_rares_2012.pdf

campaigns [17, 18]. Education is associated with better access to both health services and financial support [19, 20]. Furthermore, education is associated with better health related behavior such as lower cigarette consumption and higher levels of physical exercise [21]. An additional year of schooling appears to reduce average daily cigarette consumption by 1.6 for men and 1.1 for women and to increase physical exercise on average by 17 minutes [22, 23].

Moreover, socio-anthropological studies show that more educated individuals also have different perceptions of their body and their health care needs and they seem to experience and report pain differently [24]. These results explain why less educated people may underuse health care services, even when provided freely [25].

Studies on patient-doctor interaction suggest that the social proximity between patients and doctors influence diagnosis process, health care provision and information sharing [26, 27, 28]. Balsa and McGuire analyzed the way in which interactions between doctors and patients may contribute to social disparities and suggested that patients relative positions affect doctors interpretations and decisions during the diagnosis process. Besides, studies conducted before acute coronary crises showed that more investigations had been undertaken for better educated patients [29, 30, 31]. Patients education therefore appears to affect both health care seeking behaviors and health care professionals responses. The pre-diagnosis time span hence appears crucial in the construction of social health inequalities.

Beyond education, social relationships and networks may also play an important role in time to diagnosis inequalities. Social capital, developed by [32, 33, 34], the definition of which has evolved over time, is of growing interest for different fields, from sociology to political science to economics and epidemiology. In the health economics literature, social capital is either assessed

at the collective level (society, communities) as “features of the social organizations such as trust, norms, and networks that can improve the efficiency of society by facilitating coordinated actions” [34], or at the individual level.

In this paper, we refer to individual social capital: “individuals social characteristics that enable private returns via interaction with others that can be accumulated or transmitted across generations” [35, 36, 37]. Intensive interactions offer patients privileged channels of information transmission, the opportunity to share past experiences on health facilities, health services, health professionals, and this may both reduce patients informational costs [38] and provide social/financial support [39]. Rocco et al. presented 4 different mechanisms that may account for the link between social capital and health: (1) Social capital may provide easier access to health relevant information as a result of more intense social interaction. (2) Social capital may facilitate the provision of informal health care and psychological support. (3) Social capital may facilitate peoples lobbying efforts to obtain health-enhancing goods and services. (4) Social capital may induce rational people to reduce their risky behavior by increasing the expected value of life [36]. In the time span to diagnosis, returns to social capital may enhance patients’ health-seeking behavior and facilitate interactions within health care professionals.

The role of individual preferences in decision-making has been widely investigated in economics [40, 41]. Attitudes toward risk [42], and time preferences, i.e. the preference for immediate over delayed satisfaction [43], capture part of individual heterogeneity and are useful concepts for understanding and predicting economic behavior. Various approaches have been suggested to elicit individual preferences but relatively little attention has been paid to their contribution to health and health care seeking disparities. Studies suggest that individuals with lower time preferences or less risk adverse individuals are more likely to

undergo screening [44, 45] and provide evidence that time preferences and risk aversion are correlated with risky behaviors such as smoking. Their results also show a reduction of the educational gradient in smoking after controlling for these characteristics, supporting their role as partial mediators. Regarding access to health care, results from [44] suggest that individuals with lower time preferences tend to delay care seeking, and be less proactive during the diagnosis making period because they underestimate the future losses associated with delayed access. Similarly, risk averse individuals may fear the diagnosis and be less proactive during the diagnosis making period, and experience longer time span to diagnosis. None of these hypotheses have been empirically tested, to the best of our knowledge, and such is the aim of our paper.

While many studies examined the isolated effect of education, social capital and individual preferences on health and health behavior, very few have considered these variables together. Yet, the literature provides evidence that these social characteristics exhibit strong associations. Verba et al. [46] show that more educated people engage more often in collective activities while Putnam and Helliwell [47] show that education levels increase levels of trust, one of the commonly used proxy for social capital. Regarding individual preferences, the less educated are more likely to engage in riskier behaviors [48].

3. Data and Methods

3.1. Data

To understand the role of patients' preferences in explaining time to diagnosis, we must adopt a patient's perspective and rely on patients' reported outcomes. Although a growing field of interest, no information is jointly collected on patients' reported outcomes (PROMs) and patients' characteristics. Our study is therefore the first to link these two dimensions. It is based on pa-

tients' assessments of time elapsed from first symptoms to final diagnosis. This information was collected from an online survey conducted on a French social network dedicated to chronic patients called Carentity⁴), between May and July 2015. The response rate was estimated to 23%, with no significant differences between diseases. This response rate should be considered as a lower bound as the questionnaire was sent to all patients included in the whole database, including those who were not active for more than six months, which overestimates the pool of potential respondents. As individual health characteristics are not systematically recorded in Carentity, it was not possible to explore which factors determined individual participation.

Data collected are self-reported and retrospective. We checked patients' response consistency by comparing values for age, age of symptoms' onset, date of first symptoms and final diagnosis date. 659 (78%) completed questionnaires passed the aforementioned coherence tests. Since the study focus is the impact of education and social capital on time to diagnosis, patients aged less than 18 years old at the time of first symptoms were excluded. While reducing the sample size to 503 patients, this enables us to concentrate on patients with autonomous behavior in health management and to exclude patients diagnosed at birth or in their childhood.

The questionnaire draws from the main French health and insurance interview survey⁵. Questions on the pre-diagnosis period were designed to describe patients experience from symptoms appearance to final diagnosis. The questionnaire was piloted on a sub-sample of 21 patients.

The sample is not designed to be nationally representative and suffers from

⁴Carentity is a social network dedicated to patients with chronic conditions (see the UK's version at <https://www.carentity.co.uk/who-we-are>)

⁵The French health, health care and insurance surveys were conducted by the Institute for Research and Information in Health Economics from 1988 till 2014 and they now form the basis of the European Health Interview Survey.

the usual selection biases to be expected from with patients' participation to social networks, both in terms of social and health characteristics. Previous work conducted on the Carenity patients' database to explore its representativity suggests that compared to nationally representative patients, Carenity's sample has less seniors and more women, but displays no significant differences in geographical representativity [49]. Besides, the survey being administered online, with an open access and no control on respondents environment, we suspect patients with expanded pre-diagnostic periods to be more prone to share their experience online. The verbatim collected in the open question "What has been your experience of the health care system from symptoms appearance to final diagnosis?" shows that those who responded were likely to have experience poorly integrated care.

Regarding memory bias, the literature suggests that the period preceding final diagnosis is usually well remembered by patients [50, 51], all the more if the illness entails a substantial and durable change in social and professional life. Such should be the case for patients suffering chronic conditions, particularly those selected for our analysis: multiple sclerosis, bipolar disorder, psoriasis, and Crohns disease. Our choice favoured heterogeneous conditions, both in terms of severity and progression, for which clinical diagnoses are difficult to establish, given the non specificity of symptoms. Another key criterion was the fact that these conditions had severe impacts on functional health and patients daily lives, as well as their relative importance in the database, to ensure a large enough sample size.

Multiple sclerosis is an inflammatory disease of the central nervous system characterized by sensation disorders and mobility impairments. The diagnosis is based on imaging analysis such as MRI scan. Prognosis is unpredictable, with long remission periods [52].

Bipolar disorders are characterized by maniac phases or recurring hypomania and major depressive episodes. Addictions, isolation, divorce or unemployment are commonly associated with bipolar disorders. The risk of suicide is 15 times higher than in general population in France[53]. There is a ten-year lag, on average, between first symptoms and treatment initiation [54].

Psoriasis is a long-lasting autoimmune disease characterized by patches of abnormal skin. Diagnosis is typically based on the signs and symptoms and difficult to establish. Psoriasis is associated with an increased risk of psoriatic arthritis, lymphomas, cardiovascular disease, Crohns disease, and depression. Psoriatic arthritis affects up to 30% of individuals with psoriasis [55].

Crohns disease is a type of inflammatory bowel disease (IBD) characterized by inflammatory and remission phases. Symptoms include abdominal pain, diarrhea, fever and weight loss. Diagnosis is based on the addition of several tests including biopsy and bowel wall examination. Delayed diagnosis may lead to severe complications [56].

Variables used in the analysis

The variable *Time To Diagnosis* (in months) captures the time elapsed between first symptoms to final diagnosis. It is calculated from respondents' assessment of first symptoms and final diagnosis dates.

The variable *Education* is based on the highest educational attainment reported by patients, grouped into 3 categories: no diploma or technical degree, baccalaureate (corresponding to the high school final degree in France), and undergraduate level or more.

We used two variables to proxy the level of social capital: *Social Participation* is a dummy variable that measures patients participation to collective

activities such as voluntary-charity work, training courses, sport-social clubs, religious organizations, and political-community organizations. It is one of the most commonly used variable to capture individual social capital [57, 58]. *Social Support* is a dummy variable stating if the respondent was able to rely on support from family or friends during the pre-diagnostic period. This question was tailored to the particular context of the pre-diagnostic period and to the measurement of the pre-diagnostic social capital level. Information on symptoms severity was also collected to control for clinical symptoms using a categorical variable *Symptoms Severity* corresponding to mild, moderate or severe symptoms.

Individual preferences were collected using standardized questions developed and validated by IRDES for the French Health Healthcare and Insurance Survey 2008. The *Time Preference* and *Risk Aversion* variables were derived from scores assessed by respondents on a 0 to 10 scale. These scales were defined in the INSEE-DELTA survey [59, 60] which aimed at measuring time and risk preference measures [61, 62]. From the scores, we constructed dichotomous variables (0,1,2/3-10 for time preference and 0-7/8-10 for risk aversion, following Jusot and Khlal [45]).

The *First medical encounter* may either be a GP, a specialist outside hospital, a hospital specialist, and A & E Departments, since the French system still allows patients to freely choose between these four entry points, despite the 2004 reform, aimed at introducing GP referral.

Table 1 displays respondents socioeconomic characteristics by type of disease. Table 2 reports the distribution of time to diagnosis by disease and displays large discrepancies between and within illnesses. As shown in table 1, women represent the majority of the sample (77%): this feature is consistent with the over-representation of women in online social networks ([63]), and the

higher incidence of multiple sclerosis [64] and bipolar trouble [65] among women. Carenity respondents were on average more educated and younger than the general population [63, 49], which corresponds to general findings on social media participation rates. Looking at the distribution of time to diagnosis, we observe very skewed distributions toward 0 (less than 1 month of time to diagnosis) with very large extreme values (with a maximum value of 20 years of time to diagnosis for psoriasis and 35 years for bipolar trouble). We do observe large variations in time to diagnosis within and between each of the four chronic conditions.

Table 1: PATIENTS' SOCIODEMOGRAPHIC CHARACTERISTICS

	Medical conditions				
	Psoriasis Col %	Multiple sclerosis Col %	Bipolar trouble Col %	Crohn disease Col %	Total Col %
Gender					
Male	36.6	19.9	24.3	17.5	23.3
Female	63.4	80.1	75.7	82.5	76.7
Age group					
18-34	4.9	12.4	8.7	33.8	13.7
35-44	15.9	22.1	26.1	27.5	22.9
45-54	31.7	33.2	39.1	18.8	32.0
>55	47.6	32.3	26.1	20.0	31.4
Couple					
No	42.7	37.6	50.4	37.5	41.4
Yes	57.3	62.4	49.6	62.5	58.6
Education					
No diploma	40.2	32.3	20.0	35.0	31.2
Baccalaureate	14.6	22.1	20.9	12.5	19.1
Bac+2 and +	45.1	45.6	59.1	52.5	49.7
Social participation					
No	82.9	75.2	82.6	82.5	79.3
Yes	17.1	24.8	17.4	17.5	20.7
Social support					
No	46.3	38.1	53.9	40.0	43.3
Yes	53.7	61.9	46.1	60.0	56.7
N	82	226	115	80	503

Source: Authors data.

Table 2: TIME TO DIAGNOSTIC WORK-UP BY DISEASE (IN MONTHS)

Disease	N	Mean	SD	p25	p50	p75	Min	Max
Psoriasis	82	35.5	71.1	2	4	34	0	366
Multiple sclerosis	226	37.3	64.1	3	10	46	0	386
Bipolar trouble	115	90.5	90.8	7	68	150	0	420
Crohn disease	80	24.6	44.0	3	6.5	26.5	0	239
Total	503	47.2	73.6	3	12	57	0	420

Source: Authors data.

Table legend: Time to diagnostic is defined as the time span between symptoms appearance and final diagnostic, expressed in months. Statistics displayed are respectively: (1) Number of observations; (2) Mean ; (3) Standard deviation; (4) First quartile; (5) Median; (6) Third quartile; (7) Minimum; (8) Maximum.

3.2. Methods

We studied the links between time to diagnosis (in months) and our two main variables of interest (education, social capital) accounting for demographics, risk and time preferences, and severity differences between respondents. We used a Cox proportional hazard model [66] to analyze time to diagnosis, using Breslow method for ties. Time to diagnosis was used as our time scale (in months). This semi-parametrical procedure enabled us to compute the baseline hazard function $h(t)$ which is the failure probability - here, the diagnosis in the next step, given that diagnosis was not available at time t , without any restrictive distributional assumptions. The Cox model assumes that covariates are multiplicatively related to the hazard (ie. proportional hazard assumption) but this hypothesis appears to be refuted for Age Class, as young patients experienced on average shorter time spans to diagnosis compared to older patients in our sample.

We therefore used an extension of the Cox model to deal with non proportional hazards by stratifying over the covariate *Age Class* that does not satisfy the proportional hazard assumption. The extension allows for multiple strata with distinct baseline hazard function but common values for the coefficient vector. Our model specifies 4 strata corresponding to the 4 categories of *Age Class*

(18-34 years old; 35-44 years old; 45-54 years old; >55 years old) as displayed in Table 1.

The model for each strata is written:

$$h_g(t) = h_{0g}(t)exp[\beta_1X_1 + \dots + \beta_pX_p]$$

with $g \in \{1, 2, 3, 4\}$, strata defined from *Age Class*

Using this specification, we measure the effect of both education and social capital on the probability of survival, which here represents the probability to remain without diagnosis (the event of the survival analysis being 'obtaining the final diagnosis'). The fully adjusted models include controls for gender, marital status and symptoms severity. We added controls and clustered errors for medical conditions to account for heterogeneity across diseases. The vector X hence includes the following variables: *Education*, *Social Participation*, *Social Support* and the aforementioned control variables: gender, marital status, symptoms severity and medical condition.

4. Results

Table 3 reports estimates for the models (1), (2), (3). The results from Cox stratified models show that social capital, as measured by social participation and social support, is significantly associated with shorter time to diagnosis. Patients reporting social support or social participation exhibit hazard ratios superior to 1, meaning that social support significantly reduces the probability to have a longer time to diagnosis (Table 3 Model (1), HR respectively 1.21, $p < 0.05$ and 1.27, $p < 0.000$). This finding is robust to the inclusion of control variables accounting for time and risk preferences (Table 3 Model (2) and Model (3)).

More educated patients have a higher probability to experience longer time

Table 3: ESTIMATED COEFFICIENTS AND 95% CONFIDENCE INTERVALS FOR COVARIATES' INCIDENCE ON TIME TO DIAGNOSTIC WORK-UP USING COX MODEL STRATIFIED BY AGE CLASS.

	(Model 1)	(Model 2)	(Model 3)
Female	-0.138 (-1.64)	-0.148 (-1.91)	-0.123 (-1.78)
Couple	0.0506 (0.49)	0.0403 (0.40)	0.0835 (0.87)
Education (Ref=No diploma)			
Baccalaureate	-0.0237 (-0.89)	-0.0471*** (-3.98)	-0.0476 (-0.64)
Bac+2 and +	-0.0548** =(-2.74)	-0.0714** (-3.13)	-0.0601 (-1.55)
Social support	0.189* (2.34)	0.199** (2.82)	0.221*** (3.31)
Social participation	0.240*** (4.93)	0.244*** (4.81)	0.268*** (4.01)
Disease (Ref=Psoriasis)			
Multiple sclerosis	-0.0659 (-0.63)	-0.0560 (-0.56)	-0.143 (-1.26)
Bipolar trouble	-0.660*** (-30.95)	-0.640*** (-18.13)	-0.715*** (-19.25)
Crohn disease	0.0469 (0.85)	0.0650 (1.15)	-0.0507 (-0.63)
Symptoms' severity (Ref=Mild)			
Moderate	-0.0397 (-0.17)	-0.0662 (-0.26)	-0.0338 (-0.15)
Severe	-0.0654 (-0.39)	-0.101 (-0.51)	-0.0915 (-0.47)
Risk aversion		-0.147 (-0.90)	-0.159 (-0.87)
Pref. for present		-0.0551* (-1.96)	-0.0706*** (-5.42)
First medical contact (ref=GP)			
Hospital specialist			-0.0871 (-1.15)
Ambulatory specialist			-0.513** (-2.65)
Emergency service			-0.126 (-0.50)
Observations	503	503	503

t statistics in parentheses

* $p < 0.05$, ** $p < 0.01$, *** $p < 0.001$

to diagnosis compared to patients without diploma (Table 3, Model (1), HR 0.95 for individuals with a College degree, $p < 0.01$). After controlling for individuals risk aversion and time preferences (Table 3, Model (2)), coefficients for *Education* remain constant and significant. Preference for the present appears to significantly increase the probability of longer time to diagnosis. Table 3, Model (2), HR 0.95, $p < 0.05$), while the coefficient for risk aversion remains insignificant.

Consulting a specialist first, rather than a GP, increases significantly the probability of having a longer time to diagnosis (Table 3, Model (3), HR 0.60, $p < 0.01$). Moreover, after controlling for type of first medical encounter, the coefficient for *Education* becomes insignificant, providing evidence that education was previously impacting the probability of having a longer time to diagnosis when choosing specialists as first medical encounter.

This result is robust to the inclusion of control variables accounting for individual risk and time preferences. It is also robust to the exclusion of the 3% extremes values of our sample, which reduces the maximum value of time to diagnosis from 420 months to 262 months (see table 4 in Appendix, Model (4)). We also remove from the sample recently diagnosed individuals, respectively in the last year or the last 2 years, to avoid the risk of wrong diagnoses. The results are robust to this robustness check (see table 4 in appendix, Model (5)). Finally, we test the hypothesis of proportional hazard for our stratified Cox model: proportionality of hazard was not rejected significantly ($p > 0.67$), supporting our chosen specification.

Stronger social participation and social support significantly reduce time to diagnosis (Table 3, Model (3): HR respectively 1.25, $p < 0.001$; 1.31, $p < 0.000$). Social interactions may allow patients to gather health information as well as information from other patients, thereby reducing time to diagnosis. Social

capital also facilitates access to social and financial support.

Regarding individual preferences, only strong preferences for the present were associated with a higher probability of longer time to diagnosis (Table 3, Model (3): HR 1.25, $p < 0.001$). This result on time preferences is in line with previous findings [44]: patients with a stronger preference for the present are substantially more focused on current well-being and may be less proactive in the diagnosis-seeking process. Such patients may undervalue future losses associated with postponing a medical consultation or a less proactive behavior during the pre-diagnosis period. They may also favor procrastination, i.e. value ignorance over awareness, postponing examinations they fear may reveal an unknown pathology [67].

Higher levels of education increase the probability of longer time to diagnosis (Table 3, Model (1), HR on college degree: 0.95, $p < 0.01$). Even after controlling for risk aversion and preference for the present (Table 3, Model (2)), the coefficients for *Education* remain constant and significant. We identify three possible channels to account for this result:

- 1- More educated patients tend to challenge medical doctors assessments [68, 69], and may seek second opinions. Besides, doctors may behave differently when patients are more educated, knowing that they are more proactive and may challenge their diagnosis [70]. Recent evidence indeed shows that doctors have implicit biases which influence their response to patients' demands [71]. As a result, doctors may attempt to gather more evidence before giving their final diagnosis, to further reduce the probability of diagnosis errors.

- 2- Educated patients may retrospectively assess symptoms appearance earlier, and differences in biases regarding self-reported dates may partly drive the results on education. More educated patients have been shown to have higher expectations about their health status [72, 73] and socioeconomic variations in

pain tolerance have also been established [24]

3- In France, patients have the freedom to choose between GPs and specialists, who are both paid on a fee-for-service basis. Although gate keeping was implemented in 2004, it is not binding [74], and some patients continue accessing specialist care directly, despite higher out-of-pocket expenditures. As more educated patients consult specialists more readily than GPs, this freedom of choice may have induced social disparities in diagnostic trajectories, due to differences in individual preferences, health literacy, and economic resources [75, 76]. Ambulatory care specialists, unlike GPs, tend to delay patients referring to hospitals or emergency services [77, 78, 79], which can in turn increase time to final diagnosis.

No information was collected on patients or doctors actual behavior during the study period, which does not allow further investigations channels 1 and 2. But based on the available data, we can further document the third hypothesis. Results suggest differences in health care-seeking behavior as the coefficient on *Education* loses significance (Table 3, Model (3): HR respectively 0.94, $p > 0.1$) while the coefficient associated with specialists becomes significant (Table 3, Model (3) : HR 0.60, $p < 0.01$): the more educated patients tend to first consult specialists, which is correlated with a longer time to diagnosis, since specialists are less likely than GPs to refer patients to hospitals for additional tests, when needed.

Our study suffers from several limitations, First, since questionnaires were kept reasonably short to ensure good response rates and avoid excessive burden for respondents, we did not include questions on disease history as well as symptoms evolution. As a result, controls for disease severity lack precision as it is based on patients perception of symptoms severity in the disease early phase, to measure time to diagnosis. Moreover, beyond education, no other

socio-economic characteristics of patients were collected, such as incomes.

To address possible reverse causality issue, we chose to remove patients with symptoms occurring before 18 years old, which ensures inclusion of patients with autonomous behavior in health decisions. Yet, this method only partially answers the endogeneity issue since the time to diagnosis may have implications on educational achievements. Finally, we lacked data to model and estimate the determinants of patients decision to consult, which would contribute towards understanding patients decisions to seek care and their disease management strategies.

5. Conclusion

The full time span to diagnosis, from first symptoms to final diagnosis, has received very little attention since most of the data collection starts after the first medical encounter. Only anecdotal evidence has been gathered on variations in time to diagnosis, indicating that more educated patients or those with strong social networks have benefited from shorter time to diagnosis. What is well documented, on the other hand, is that delays in obtaining the final diagnosis have a detrimental impact on health prognoses, outcomes and more generally on quality of life. Using a patients' network for our data collection, and adopting a patient's perspective, we consider four diverse chronic conditions (bipolar trouble, Crohn disease, multiple sclerosis and psoriasis), and offer a first measure of this time to diagnosis. Despite data restrictions, we are able to document large variations between patients and between conditions.

Our results provide evidence that the first medical encounter plays an important role in explaining time to diagnosis. Patients seeking GPs first have shorter time spans, compared to those seeking specialists for their first medical

encounter. This may be due to the fact that specialists are less prone than GPs to referring patients to hospitals for additional tests, when needed. These results provide support for public policies aimed at establishing coordinated and streamlined health care pathways, with GPs as gatekeepers.

Regarding social networks (social participation and social support), we find that they reduce the probability of experiencing longer time spans to diagnosis and support WHO's recommendations to enhance individual social capital.

Higher levels of education, for their part, seem to increase the probability of longer time spans. We further analyze this unexpected result by identifying differences in patients' health care-seeking behavior and find that it is the more educated patients who tend to consult specialists first, leading to longer time spans. Two alternative channels are offered for this result on education, based on 1) specialists' implicit biases regarding educated patients or 2) differences in patients' ability to identify and report dates of first symptoms. None of these two alternative channels can be identified using our data, but they are plausible and partly contribute to our education result.

While our results on social networks are likely to apply to most developed countries' health care systems, those on education may not hold for health care systems where gate keeping is fully implemented. Carrying out similar studies in countries where higher education or stronger social networks may influence time to diagnosis through other channels would shed light on these issues. Our results point to the important role of the pre-diagnosis period in the construction of differentiated access to diagnosis. Yet delays in time to diagnosis, not solely patients' responsibility, are an under-documented aspect of the fight against health inequalities. Potential leverage for their reduction must be further explored in academic research to refine health care practices and health care system's reforms.

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Appendices

Table 4: ROBUSTNESS CHECKS: MODEL (4): COEFFICIENTS FOR MODEL (3) WITHOUT 3% TOP VALUES AND MODEL (5) DISPLAYS COEFFICIENTS FOR MODEL (3) RESTRICTING FOR PATIENTS DIAGNOSED WITHIN THE 2 PREVIOUS YEARS

	(Model 4)		(Model 5)	
Female	-0.0976	(-1.53)	-0.158	(-1.93)
Couple	0.101	(0.88)	0.0606	(0.64)
Education (Ref=No diploma)				
Baccalaureate	-0.0946	(-1.17)	-0.0504	(-0.47)
Bac+2 and +	-0.0587	(-0.85)	-0.0520	(-1.07)
Social support	0.154***	(4.74)	0.248***	(4.05)
Social participation	0.174*	(2.52)	0.278***	(3.87)
Disease (Ref=Psoriasis)				
Multiple sclerosis	-0.355***	(-5.17)	-0.135	(-1.11)
Bipolar trouble	-1.037***	(-11.01)	-0.695***	(-18.67)
Crohn disease	-0.374***	(-4.13)	-0.0562	(-0.66)
Symptoms' severity (Ref=Mild)				
Moderate	-0.0149	(-0.09)	-0.0821	(-0.35)
Severe	0.000401	(0.00)	-0.139	(-0.75)
Risk aversion	-0.0877	(-0.51)	-0.219	(-1.27)
Pref. for present	-0.236***	(-3.42)	-0.0420*	(-2.43)
First medical contact=1	0	(.)	0	(.)
Hospital specialist	-0.0672	(-0.83)	-0.120	(-1.38)
Ambulatory specialist	-0.413*	(-2.55)	-0.536**	(-2.95)
Emergency service	-0.000556	(-0.00)	-0.148	(-0.54)
Observations	489		474	

t statistics in parentheses

* $p < 0.05$, ** $p < 0.01$, *** $p < 0.001$

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