

A Service of

ZBW

Leibniz-Informationszentrum Wirtschaft Leibniz Information Centre for Economics

Kutasi, Kristóf et al.

Working Paper Understanding hesitancy with revealed preferences across COVID-19 vaccine types

KRTK-KTI Working Papers, No. KRTK-KTI WP - 2022/07

Provided in Cooperation with: Institute of Economics, Centre for Economic and Regional Studies, Hungarian Academy of Sciences

Suggested Citation: Kutasi, Kristóf et al. (2022) : Understanding hesitancy with revealed preferences across COVID-19 vaccine types, KRTK-KTI Working Papers, No. KRTK-KTI WP - 2022/07, Hungarian Academy of Sciences, Institute of Economics, Centre for Economic and Regional Studies, Budapest

This Version is available at: https://hdl.handle.net/10419/256920

Standard-Nutzungsbedingungen:

Die Dokumente auf EconStor dürfen zu eigenen wissenschaftlichen Zwecken und zum Privatgebrauch gespeichert und kopiert werden.

Sie dürfen die Dokumente nicht für öffentliche oder kommerzielle Zwecke vervielfältigen, öffentlich ausstellen, öffentlich zugänglich machen, vertreiben oder anderweitig nutzen.

Sofern die Verfasser die Dokumente unter Open-Content-Lizenzen (insbesondere CC-Lizenzen) zur Verfügung gestellt haben sollten, gelten abweichend von diesen Nutzungsbedingungen die in der dort genannten Lizenz gewährten Nutzungsrechte.

Terms of use:

Documents in EconStor may be saved and copied for your personal and scholarly purposes.

You are not to copy documents for public or commercial purposes, to exhibit the documents publicly, to make them publicly available on the internet, or to distribute or otherwise use the documents in public.

If the documents have been made available under an Open Content Licence (especially Creative Commons Licences), you may exercise further usage rights as specified in the indicated licence.



WWW.ECONSTOR.EU



KRTK-KTI WORKING PAPERS | KRTK-KTI MŰHELYTANULMÁNYOK

Understanding hesitancy with revealed preferences across COVID-19 vaccine types

KUTASI KRISTÓF – KOLTAI JÚLIA – SZABÓ-MORVAI ÁGNES –

RÖST GERGELY – KARSAI MÁRTON – BIRÓ PÉTER – LENGYEL

BALÁZS

KRTK-KTI WP – 2022/07

March 2022

https://kti.krtk.hu/wp-content/uploads/2022/03/CERSIEWP202207.pdf

KRTK-KTI Working Papers are distributed for purposes of comment and discussion. They have not been peer-reviewed. The views expressed herein are those of the author(s) and do not necessarily represent the views of the Centre for Economic and Regional Studies. Citation of the working papers should take into account that the results might be preliminary. Materials published in this series may be subject to further publication.

A KRTK-KTI Műhelytanulmányok célja a viták és hozzászólások ösztönzése. Az írások nem mentek keresztül kollegiális lektoráláson. A kifejtett álláspontok a szerző(k) véleményét tükrözik és nem feltétlenül esnek egybe a Közgazdaság- és Regionális Tudományi Kutatóközpont álláspontjával. A műhelytanulmányokra való hivatkozásnál figyelembe kell venni, hogy azok előzetes eredményeket tartalmazhatnak. A sorozatban megjelent írások további tudományos publikációk tárgyát képezhetik.

INSTITUTE OF ECONOMICS, CENTRE FOR ECONOMIC AND REGIONAL STUDIES EÖTVÖS LORÁND RESEARCH NETWORK (ELKH) BUDAPEST, 2022

<u>ABSTRACT</u>

Many countries have secured larger quantities of COVID-19 vaccines than their populace is willing to take. This abundance and variety of vaccines created a historical moment to understand vaccine hesitancy better. Never before were more types of vaccines available for an illness and the intensity of vaccine-related public discourse is unprecedented. Yet, the heterogeneity of hesitancy by vaccine types in certain segments of society has been neglected so far, even though factual or believed vaccine characteristics and patient attributes are known to influence acceptance. In this paper, we address this problem by analysing acceptance and assessment of five vaccine types using information collected with a nationally representative survey (N=1000) at the end of the third wave of the COVID-19 pandemic in Hungary, where a unique portfolio of vaccines were available to the public in large quantities. Our special case enables us to quantify revealed preferences across vaccine types since one could evaluate a vaccine unacceptable and even could reject an assigned vaccine to wait for another type. We find that the source of information that respondents trust characterizes their attitudes towards vaccine types differently and leads to divergent vaccine hesitancy. Believers of conspiracy theories were significantly more likely to evaluate the mRNA vaccines (Pfizer and Moderna) unacceptable while those who follow the advice of politicians evaluate vector-based (AstraZeneca and Sputnik) or whole-virus vaccines (Sinopharm) acceptable with higher likelihood. We illustrate that the rejection of non-desired and re-selection of preferred vaccines fragments the population by the mRNA versus other type of vaccines while it generally improves the assessment of the received vaccine. These results highlight that greater variance of available vaccine types and individual free choice are desirable conditions that can widen the acceptance of vaccines in societies.

JEL codes: I12

Keywords: vaccination, COVID-19, revealed preferences

<u>Kutasi Kristóf</u>

Rice University, Department of Economics, Houston TX, 77005-1827, USA <u>Koltai Júlia</u>

Centre for Social Sciences, Computational Social Science - Research Center for Educational and Network Studies, Budapest, 1097, Hungary

Eötvös Loránd University, Faculty of Social Sciences, Budapest, 1117, Hungary

Central European University, Department of Network and Data Science, Vienna, 1100, Austria

Szabó-Morvai Ágnes

Eötvös Loránd Research Network, Centre for Economic and Regional Studies, Health and Population Lendület Research Group, Budapest, 1097, Hungary

Debrecen University, Department of Economics, Debrecen, 4032, Hungary Röst Gergely

University of Szeged, Bolyai Institute, Szeged, 6722, Hungary

Karsai Márton

Central European University, Department of Network and Data Science, Vienna, 1100, Austria

Alfréd Rényi Institute of Mathematics, Budapest, 1053, Hungary

Biró Péter

Eötvös Loránd Research Network, Centre for Economic and Regional Studies,

Mechanism Design Lendület Research Group, Budapest, 1097, Hungary

Corvinus University of Budapest, Department of Operations Research and Actuarial Sciences, Budapest, 1093, Hungary

Lengyel Balázs

Eötvös Loránd Research Network, Centre for Economic and Regional Studies,

Agglomeration and Social Networks Lendület Research Group, Budapest, 1097,

Hungary

Corvinus University of Budapest, Corvinus Institute for Advanced Studies, Budapest, 1093, Hungary lengyel.balazs@krtk.hu

A vakcinaelutasítás elemzése a COVID-19 vakcinák iránti kinyilvánított preferenciákon keresztül

KUTASI KRISTÓF – KOLTAI JÚLIA – SZABÓ-MORVAI ÁGNES – RÖST GERGELY – KARSAI MÁRTON – BIRÓ PÉTER – LENGYEL BALÁZS

<u>ÖSSZEFOGLALÓ</u>

A COVID válság során sok ország szerzett be nagyobb mennyiségű vakcinát, mint amennyit várhatóan fel tudnak használni. A felhalmozott készletek mérete és a vakcinák sokfélesége alkalmat ad arra, hogy a vakcina elutasítás jelenségét mélyebben megértsük. Soha ezelőtt nem állt rendelkezésre ennyiféle vakcina ugyanarra a betegségre, és soha nem volt ilyen széleskörű és színes a vakcinákkal kapcsolatos közbeszéd. A vakcina elutasítás heterogenitása mindezidáig kevés figyelmet kapott, noha a fogyasztók bizonyos karakterisztikái, és az egyes vakcinák vélt vagy valós tulajdonságai bizonyítottan befolyásolják a vakcinák elfogadását.

Cikkünkben ezt a kérdést járjuk körül egy 1000 fős reprezentatív minta segítségével, amelyet Magyarországon, a COVID-19 járvány harmadik hullámának végén készítettünk, amikor az országban a világon egyedülállóan széles skálája volt elérhető a vakcináknak. Ez a speciális helyzet lehetővé tette, hogy értékeljük vakcinánként a kinyilvánított preferenciákat, mivel lehetőség volt a felajánlott vakcina visszautasítására és egy másik típusú oltás kivárására. A megkérdezettek mindegyik vakcina típust értékelték, függetlenül a ténylegesen számukra felajánlott típustól.

Az elemzésünkben azt találjuk, hogy a megkérdezettek által megbízhatónak tartott információ forrás típusa meghatározza az egyes vakcina típusokhoz fűződő preferenciáikat. Ez a különféle vakcina típusokkal kapcsolatos preferenciák és elutasítás heterogenitásához vezet. Akik a konspirációs elméletekben hisznek, azok sokkal nagyobb valószínűséggel értékelték az mRNA típusú vakcinákat (Pfizer és Moderna) elfogadhatatlannak. Akik a politikusok véleményére hallgatnak inkább, azok a vektor alapú (AstraZeneca és Sputnik) illetve a teljes vírus típusú (Sinopharm) vakcinákat nagyobb valószínűséggel tartották elfogadhatónak.

Az elemzésben bemutatjuk, hogy a nem kívánt vakcinák elutasításának és egy eltérő típusú vakcina választásának lehetősége megosztotta a társadalmat az mRNA és az egyéb vakcinák között. Ugyanakkor növelte a végül elfogadott vakcina értékelését. Ezek az eredmények azt sugallják, hogy a szélesebb választási lehetőség és az egyéni szabad vakcinaválasztás kívánatos feltételek lehetnek, és ezek javítják a vakcinák elfogadását a társadalmakban.

JEL: I12

Kulcsszavak: oltás, COVID-19, kinyilvánított preferenciák

Understanding hesitancy with revealed preferences across COVID-19 vaccine types

Kristóf Kutasi¹, Júlia Koltai^{2,3,4}, Ágnes Szabó-Morvai^{5,6}, Gergely Röst⁷, Márton Karsai^{4,8}, Péter Biró^{9,10,+}, and Balázs Lengyel^{11,12,+,*}

¹Rice University, Department of Economics, Houston TX, 77005-1827, USA

²Centre for Social Sciences, Computational Social Science - Research Center for Educational and Network Studies, Budapest, 1097, Hungary

³Eötvös Loránd University, Faculty of Social Sciences, Budapest, 1117, Hungary

⁴Central European University, Department of Network and Data Science, Vienna, 1100, Austria

⁵Eötvös Loránd Research Network, Centre for Economic and Regional Studies, Health and Population Lendület Research Group, Budapest, 1097, Hungary

⁶Debrecen University, Department of Economics, Debrecen, 4032, Hungary

⁷University of Szeged, Bolyai Institute, Szeged, 6722, Hungary

⁸Alfréd Rényi Institute of Mathematics, Budapest, 1053, Hungary

⁹Eötvös Loránd Research Network, Centre for Economic and Regional Studies, Mechanism Design Lendület Research Group, Budapest, 1097, Hungary

¹⁰Corvinus University of Budapest, Department of Operations Research and Actuarial Sciences, Budapest, 1093, Hungary

¹¹Eötvös Loránd Research Network, Centre for Economic and Regional Studies, Agglomeration and Social Networks Lendület Research Group, Budapest, 1097, Hungary

¹²Corvinus University of Budapest, Corvinus Institute for Advanced Studies, Budapest, 1093, Hungary

*Corresponding author: lengyel.balazs@krtk.hu

+these authors jointly supervised this work

ABSTRACT

Many countries have secured larger quantities of COVID-19 vaccines than their populace is willing to take. This abundance and variety of vaccines created a historical moment to understand vaccine hesitancy better. Never before were more types of vaccines available for an illness and the intensity of vaccine-related public discourse is unprecedented. Yet, the heterogeneity of hesitancy by vaccine types in certain segments of society has been neglected so far, even though factual or believed vaccine characteristics and patient attributes are known to influence acceptance. In this paper, we address this problem by analysing acceptance and assessment of five vaccine types using information collected with a nationally representative survey (N=1000) at the end of the third wave of the COVID-19 pandemic in Hungary, where a unique portfolio of vaccines were available to the public in large quantities. Our special case enables us to quantify revealed preferences across vaccine types since one could evaluate a vaccine unacceptable and even could reject an assigned vaccine to wait for another type. We find that the source of information that respondents trust characterizes their attitudes towards vaccine types differently and leads to divergent vaccine hesitancy. Believers of conspiracy theories were significantly more likely to evaluate the mRNA vaccines (Pfizer and Moderna) unacceptable while those who follow the advice of politicians evaluate vector-based (AstraZeneca and Sputnik) or whole-virus vaccines (Sinopharm) acceptable with higher likelihood. We illustrate that the rejection of non-desired and re-selection of preferred vaccines fragments the population by the mRNA versus other type of vaccines while it generally improves the assessment of the received vaccine. These results highlight that greater variance of available vaccine types and individual free choice are desirable conditions that can widen the acceptance of vaccines in societies.

1 Introduction

The fast development of vaccines against the SARS-COV-2 virus will go to history books as an outstanding scientific achievement^{1,2}. Unprecedented efforts and investments resulted vaccines of various types³: established techniques yielded vector-based and whole-virus vaccines while cutting-edge technologies were pioneered for mRNA vaccines. However, mostly due to misinformation on vaccines, and uncertainties about their effects on health, large populations remained hesitant against vaccination^{4–7}, threatening the efficiency of vaccination plans^{8–12}. Previous literature on vaccine hesitancy has considered the individual act of taking a vaccine an uncertain decision because risks and benefits are difficult to compare¹³. Hence, besides

demographic and socio-economic characteristics that correlate with hesitancy^{14–16}, personal beliefs are thought to be critical as well^{17, 18}. Consequently, the trusted source of information is an important factor: those who trust public authorities and scientists are generally less hesitant^{4,19,20}, while believers of conspiracy theories are more hesitant about the vaccines^{21–23}.

Despite the intensive public discourse about the efficacy²⁴ and health risks of specific vaccines²⁵, how the level of hesitancy differs across vaccine types is unclear. It makes vaccine comparison difficult that most countries apply only a few selected vaccines. Yet, individuals might not be hesitant against all types of vaccines; instead, might be open for vaccines that use certain technologies or come from certain countries²⁶. However, such insights were gained so far only by investigating hypothetical vaccines²⁷. Therefore, to increase vaccine acceptance, it is important to better understand how individual characteristics, and especially how the trusted source of information, can explain the hesitancy about particular vaccine types.

In this paper, we compare the acceptance and assessment of five vaccine types in Hungary. This country applied the greatest diversity of vaccines in the third wave of the COVID-19 pandemicⁱ. Besides the centralized European Union (EU) purchase, which included the Pfizer-Biontech, Moderna, and AstraZeneca vaccines, the Hungarian government imported Sputnik from Russia and Sinopharm from China. (Although Janssen was also available in Hungary from May 2021, only a small share of population received it. We excluded this vaccine from the analysis.)ⁱⁱ Although the application of Sputnik and Sinopharm put Hungary among the quickest countries globally in the early vaccination period and vaccination rate was still outstanding among Central and Eastern European countries at the beginning of the fourth waveⁱⁱⁱ, the vaccination plan generated an intensive public debate. The government strongly communicated the safety and efficacy of Sputnik and Sinopharm²⁸; while mRNA vaccines were rated high in society²⁹. As hesitancy was high in the beginning of the vaccination campaign¹⁹ (53% in January 2021 and 20% in May 2021²⁹) a rigid vaccine assignment could have potentially decreased people's willingness to vaccinate. Thus, individuals were allowed to reject the assigned vaccine and wait for the preferred one.

This historically unique situation enables us to contribute to the ongoing discussion in two ways. First, the globally outstanding diversity of available vaccines in a single country allows us to compare hesitancy across vaccines by technologies (mRNA, vector-based, and whole virus). We are able to identify the individual characteristics that correlate with hesitancy of specific vaccines, focusing on the role of trusted source of information. Second, the special allocation mechanism (i.e. the possibility to reject a vaccine and wait for another) allows us to analyze whether free choice across vaccine types can mitigate hesitancy.

Our empirical analysis is based on a nationally representative survey (N=1000) that we conducted at the end of the third wave of the pandemic when vaccines were already available for everyone^{iv} We collected detailed demographic, socio-economic, and health information of respondents. We asked respondents to identify sources they accept vaccine-related advice from (these include medical doctors, scientists, anti-vaccine propagators, politicians, family etc). They were also requested to evaluate each vaccine and clarify whether their assessment has changed over the third wave. Finally, a list of questions were used to map their COVID-19 infection and vaccination history including the date of potential positive tests, date and type of vaccines received and date and type of vaccines rejected (18% of our vaccinated respondents rejected an assigned vaccine and later took another type). This latter information provides new insights into the nature of vaccine hesitancy through the analysis of revealed vaccine preferences and allows for the comparison of vaccine assessments between rejected and accepted vaccines. From now on, we use the word *evaluate* to refer on whether individuals considered the vaccine types acceptable; *reject* refers to the act when the individual was not willing to take the assigned vaccine; while *accept* is used to describe when a vaccine is taken.

We find heterogeneous vaccine preferences. Most individuals prefer mRNA vaccines but a non-negligible portion (5.5%) of respondents who rejected a vaccine preferred vector-based and whole virus vaccines against mRNA types. The source of vaccine-related advice is an important factor of vaccine hesitancy. We find evidence for divergent attitudes towards vaccine types that can depend on the trusted source of advice. Believers of conspiracy theories are significantly more likely to evaluate the mRNA vaccines unacceptable. Those who follow the advice of national politicians are more likely to evaluate less popular vector-based and whole virus vaccines acceptable, even those that are not approved by the EU. We illustrate that the rejection of non-desired and re-selection of preferred vaccines generally improve the assessment of received vaccines. However, these revealed preferences fragments the population into groups of individuals that either accepts the newest but reject the more established vaccine technologies and vice versa.

The remaining of the article has the following structure. We first introduce our data and illustrate patterns and dynamics of revealed preferences across vaccine types. Then, we analyse determinants of individual hesitancy by vaccine types using statistical techniques. Finally, we investigate vaccine assessment across those who have accepted or rejected the assigned vaccine. Our results call for diversified communication and vaccine allocation strategies to convince the hesitant population. Government communication can be effective in convincing the local population about less popular vaccines. However, misinformation must be fought to mitigate hesitancy towards the newly developed vaccines that might partly originate in uncertainties around new technologies. Although our analysis cannot prove cause and effect, the suggestive evidence can be used to develop a new hypothesis. In the Discussion, we argue that a greater variance of available vaccine types and individual free choice are desirable conditions that can widen the acceptance of vaccines in societies.

2 Results

2.1 Vaccine assignment and revealed preferences

Let us give a general overview about the logistics of the COVID-19 vaccination procedure in Hungary over the third wave of the COVID-19 pandemic, 2021 Spring. A vaccine seeking individual (patient) first had to register online at a government portal^v. The general physician (GP) received the list of patients in its region and contacted them in order of priority, based on chronic illness, age and occupation. Upon contacting a patient, the GP offered a type of vaccine to the patient. The offer followed the guidence of central vaccination plan and depended on patient characteristics (as explained above) and the availability of vaccines. However, the patient could either accept or reject the offered vaccine. In case of a rejection, the GP had to reach out to the patient later, when a different type of vaccine became available. This procedure characterized early vaccination until the vast majority of priority groups were vaccinated. Starting from April, when vaccines became increasingly available, patients could register and select available vaccines through a web-application (first AstraZeneca, then Sinopharm and Sputnik, and later other vaccines as well).

The vaccine allocation procedure in Hungary is especially interesting as patients could choose over the type of vaccination they would like to get. The individual's vaccine selection was in sharp contrast with the allocation mechanism used in other EU countries or in the United States in 2021 Spring where patients found out the assigned vaccine type only when they arrived to the vaccination center. Besides giving floor for various individual preferences, the Hungarian mechanism might have provided acceptable alternatives for those who got scared from the assigned vaccine. For example, the Sinopharm vaccine was assigned initially in Hungary to the older generation, despite it was recommended to a younger population. Similarly, AstraZeneca was assigned for the middle-aged generation, but many became afraid when the connection between the vaccine and thrombosis cases were discussed.

To gain insights into these revealed individual preferences on vaccine types, we have conducted a nationally representative survey data collection on a sample of 1,000 people with CATI (computer assisted telephone interviewing)^{30,31} between May 25 and 31, 2021. Until this date the majority of non-hesitant population has already received at least one dose of vaccine²⁹. The data is representative for the adult Hungarian population by gender, age, education and domicile. For more detailed description of the data collection and survey methods see the Methods Section.

Table 1 describes the vaccination related variables by socio-demographic characteristics of the respondents. Evaluation and rejection of vaccines - which information makes our dataset unique - are presented by sex, education level, chronic illness, and age categories. Here we can distinguish those who rate all vaccine types unacceptable (2.3%) from those who evaluate at least one vaccine unacceptable but evaluate another acceptable (18.2%). Further, we can compare those who accepted the first

	Evaluated all vaccines	Evaluated one vaccine acceptable but	Vaccinated without rejection	Vaccinated after rejection	Non-vaccinated
	unacceptable	another unacceptable			
	(%)	(%)	(%)	(%)	(%)
Men	1.1	7.9	30.7	4.6	11.7
Women	1.2	10.3	32.2	6.2	14.5
University	0.2	3.4	14.4	2.9	4.3
High-school	1.7	11.7	32.7	6.2	14.7
Elementary	0.4	3.1	15.9	1.7	7.2
Not Chronic Illness	1.6	11.8	33.0	6.4	19.9
Chronic Illness	0.7	6.4	29.9	4.4	6.3
Age 20-39	0.9	7.8	16.2	4.3	12.6
Age 40-59	0.8	5.9	20.6	3.4	9.2
Age 60-79	0.5	4.0	23.0	2.7	3.3
Age 80+	0.0	0.2	2.3	0.0	0.3
Pfizer	-	5.0	20.3	5.8	-
Moderna	-	0.8	3.5	1.0	-
AstraZeneca	-	1.1	12.1	1.1	-
Sputnik	-	1.4	12.3	1.3	-
Sinopharm	-	1.5	14.7	1.6	-

Table 1. Percentage of individuals in the sample by major socio-demographic characteristics and COVID vaccination history. Percentages in the last three columns add up to 100% by character groupings (except the rows on specific vaccine types where the last column is not relevant). Individuals are classified by type of vaccine taken in the last rows.

vaccine offered (62.9%) with those who rejected the first offer (10.8%). 26.3% of the sample has not received any vaccine until the data collection. The ratio of non-vaccinated individuals in the total population of the country was around 32% in June 2021. This moderate bias might be due to the tendency that those who are more concerned about the pandemic situation are more willing to take vaccines and participate in related surveys, or, to conform to the public norm, some unvaccinated might not have been admitted it. Apparently, women, the younger cohorts, the lower educated and those with no chronic disease are more likely to reject an offer or a vaccine in general.

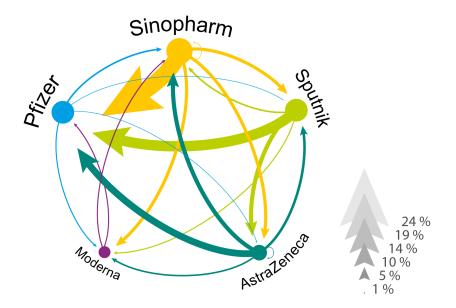


Figure 1. Revealed vaccine preferences by rejection and re-selection. The network of revealed vaccine preferences in Hungary. Vaccines are linked if someone in our data has rejected one vaccine and later accepted another. The direction of the link goes from rejected to accepted and the width of the arrow corresponds to the percentage of individuals who acted accordingly among those who rejected a vaccine.

The data allow us to investigate acceptance and revealed preferences by vaccine types. Table 1 demonstrates that 5% of those respondents who have received Pfizer rejected at least one other vaccine; while these shares are lower in the case of those who received other vaccine types. A similar pattern characterizes our sample according to the rejection of vaccines: 20.3% of all respondents have accepted Pfizer without rejecting it while 5.8% waited for Pfizer after rejecting another. There are fewer individuals in the data who accepted other assigned vaccines without rejection but except Moderna (3.5), which had the shortest supply in the country, AstraZeneca (12.1%), Sputnik (12.3%), and Sinopharm (14.7%) were accepted by large populations without hesitancy. Although a smaller fraction of society than in the case of Pfizer, but few individuals waited for these latter vaccines specifically.

Figure 1 illustrates the network of revealed preferences across vaccine types from our data. We observe a high ratio of individuals, among those who rejected a vaccine, who received Pfizer but rejected Sinopharm (23%), Sputnik (15%) or AstraZeneca (9.7%). However, individuals also accepted these latter vaccines after rejecting other types. We see a few individuals who rejected the most accepted and then accepted a less popular vaccine. For example, 2.2% of the hesitant population have rejected Pfizer to receive Sinopharm. The rules of network generation is described in the Methods section.

Figure 2A demonstrates the dynamics of revealed preferences by vaccines types. Although Pfizer was available early on, individual selection was only possible later in spring when the prioritized groups has all been already vaccinated. Sinopharm could be selected the same time as Pfizer and it was specifically allocated to the older population. AstraZeneca, Moderna and Sputnik could be chosen relatively early compared to the other vaccines. Figure 2B shows that rejection and acceptance happened in two cohorts. Elder patients with chronic illness rejected vaccines early so that they could get vaccinated with a more preferred vaccine type in March. The younger individuals received their first vaccine later and could also select a preferred vaccine. Supporting Information 1 contains further description of the dynamics of vaccination and rejection by age, chronic illness and vaccine types.

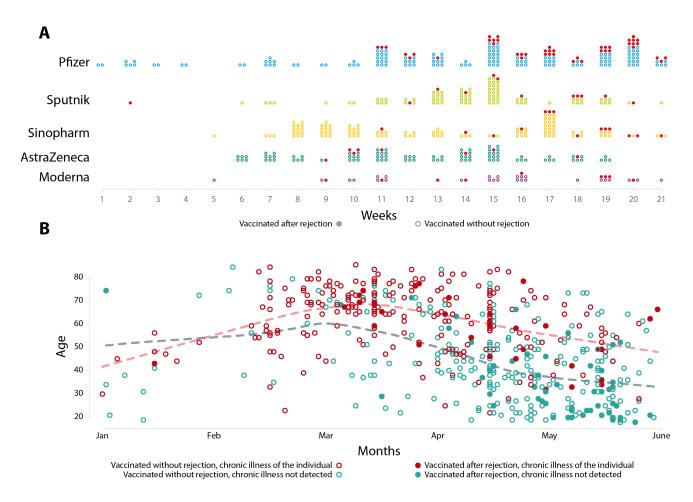


Figure 2. Dynamics of vaccination by vaccine types and re-selection. A Dynamics of vaccination by vaccine types. Dots represent number of vaccines taken by weeks. Dark dots represent those vaccines that were taken after rejecting another one. **B** The dynamics of vaccination by age and chronic illness. The days are identified by first vaccine taken. Those who took a vaccine after rejecting another are depicted with filled markers and those who took the first assigned vaccine are depicted with hollow marker. Dashed lines represent the local averages of vaccinated patients' age (dark is used for patients with chronic illness and bright is used for healthy).

2.2 Trusted information sources characterize groups of vaccine preferences

Let us turn now to the determinants of vaccine hesitancy. Individuals can be influenced in a very complex manner by the public discourse on vaccines. The competing messages of actors can be amplified by the communication channel used (eg. social media) and the mechanisms are difficult to disentangle. To better understand how external influence relates to hesitancy, the respondents were asked to identify their trusted information sources on COVID-19 vaccination, as well as their socio-demographic, health-related information, detailed history of vaccination and any infections. The following sources of trusted information were listed: doctors, scientists, anti-vaccine propagators, politicians, family, friends, journalists, celebrities. The respondents also had to name their most frequently used information channels on pandemic measures such as web news, social media, press, radio, TV, family, friends. Major actors communicated differently about the vaccines. Doctors and scientists argued for vaccination^{vi}, whereas anti-vaccine propagators emphasized the non-effectiveness and the side effects of the COVID-19 vaccines^{vii}. Some medical doctors warned against vaccines that are not approved by the EU^{viii}. The government communication campaigned for the non-EU-approved vaccines to improve their rate of acceptance by the population ^{ix} but a heated debate around vaccine types has fragmented the society ^x.

We apply a variable selection method to decrease the complexity of the problem that both the source of information and the channel of communication (eg. social media) can influence hesitancy simultaneously. The LASSO model (short for least absolute shrinkage and selection operator) enables us to select the most important variables that have high-enough predictive power (the process is described in Methods). Estimation results are reported in Supporting Information 2, Table S1. We find

that communication channels are not important determinants of hesitancy, because all but one (web news) are dropped from the model. We also see that both the demographic characteristics and the sources of trusted information on COVID-19 vaccination are important predictors of the outcome variables. We rely on this result when we select the explanatory variables to the linear probability models.

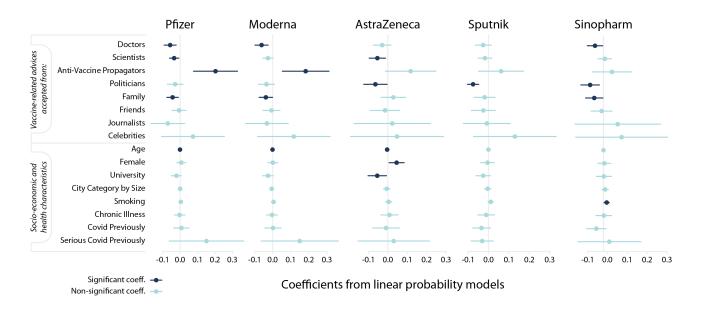


Figure 3. Source of advice taken characterize hesitancy against vaccines differently. Estimation results from linear probability models with robust standard errors. The dependent variable equals 1 if the respondent evaluates a vaccine type unacceptable and 0 otherwise. Markers denote point estimates and 95% confidence intervals are denoted by the bars. Non-significant coefficients are in light blue ($p \ge 0.05$); significant coefficients are in dark-blue (p < 0.05).

Next, we estimate linear probability models (LPM) to analyze the factors that shape the evaluation of each vaccine individually. The dependent variables take value of 1 if the vaccine is evaluated "unacceptable" and value of 0 if it is evaluated either as "acceptable", "fair", "good" or "best". Independent variables include "Advice" indicators that take value 1 if the individual follows the advice of the denoted source. "Age" is a numeric variable and is measured in years, "Female" takes value 1 if the respondent is woman and zero otherwise, "University" equals 1 if the respondent has a university degree or equivalent, "Smoking" and "Chronic Illness" are binary variables as well as "COVID Previously" and "Serious COVID previously". We present the summary statistics in Supporting Information 3, which also contains a correlation matrix and VIF statistics. The latter tests indicate no issues of multicollinearity. The estimation specification can be found in Methods.

Figure 3 illustrates point estimates and 95% confidence intervals calculated from heteroscedasticity-robust standard errors. (For detailed results see Table S3 in Supplementary Information 4.) These results suggest that the factors that affect the rejection decision differ among vaccine types. The most remarkable differences are in the effect of the trusted information sources. Those who trust the advice of anti-vaccine propagators are 18.2 and 20.2 percent more likely to evaluate vaccines made with the newest mRNA technologies (Pfizer and Moderna) unacceptable. On the contrary, trusting doctors' advice decreases the hesitancy probability by 5.7, 6.1, and 5.4 percent for Pfizer, Moderna, and Sinopharm. Trusting scientists reduce the likelihood of evaluating a vaccine unacceptable by 3.5, 2.6 and 5.5 for Pfizer, Moderna, and AstraZeneca. Those who take advice from politicians are 6.4, 7.6 and 8.4 percent more likely to evaluate AstraZeneca, Sputnik and Sinopharm acceptable. Those who trust their family's advice are more likely to evaluate Pfizer, Moderna and Sinopharm acceptable by 4.3, 3.8 and 5.8 percent. The coefficients for the advice of celebrities are either not or only slightly significant. However, the point estimates are large and positive for all types of vaccines.

Socio-demographic characteristics of respondents are found to correlate with vaccine hesitancy in a plausible way. Age has a negative correlation implying that vulnerable, older individuals are less likely to evaluate vaccines unacceptable. Although the coefficient is small in Figure 3, the regression on standardized variables reported in Table S3 (in Supporting Information 4) reveals that age is an important factor of evaluating Pfizer, Moderna and AstraZeneca acceptable. Females are more likely to evaluate AstraZeneca unacceptable, probably due to reported cases of females' blood clot^{xi}. University degree, however, decreases hesitancy against AstraZeneca.

We have carried out various robustness checks that are reported in Supporting Information 4. First, our results hold if we use a logistic regression specification instead of ordinary least squares. Second, our results remain similar after the inclusion of subjective wealth as a control variable. One limitation of our study follows from the fact that the survey was taken when most registered persons have already been vaccinated. This means that "spreading of the alternatives"³² may have affected our results. That is, after a difficult decision making, the valuation of the accepted alternative increases, whereas, that of the rejected alternative decreases. Our results do not change if we apply the models on a subsample of those who had stable vaccine assessment over the period.

In sum, these results suggest diverse vaccine hesitancy across vaccine types. Vaccines using the newest mRNA technologies (Pfizer and Moderna) are likely to be evaluated unacceptable by those who follow anti-vaccine propagators. The large coefficients imply the effectiveness of anti-vaccine propagators. Vaccines using more established vector-based and whole virus technologies are more likely to be evaluated acceptable by those who follow local politicians. This suggests that public communication can make less popular vaccines acceptable for many. We find that the advice of doctors and scientists can mitigate hesitancy against all types of vaccines. Nevertheless, the coefficients of advice taken from doctors, scientists, and politicians are much smaller, half to one-third of the anti-vaccine voices. This difference could explain the forms of communications: the anti-vaccine propagators usually target fears and emotions, whereas doctors, scientists, and politicians tend to use rational arguments.

2.3 Individual selection improves vaccine assessment

In this section, we analyze vaccine assessment of patients sorted by received and rejected vaccines. This enables us to inform policy on how free choice of vaccines can impact vaccine acceptance. We can also group vaccines based on the tendency that individuals find both of them acceptable.

The nationally representative survey asked respondents to rate each of the vaccines based on their personal opinion on a scale from 1 to 5. Figure 4 summarizes the vaccine evaluations in the form of a heat map. A heat map cell corresponds to the average rating of a type of vaccine denoted by the row among a group of patients who got and did not reject any vaccine (Figure 4A), got and rejected any vaccine (Figure 4B), or were first offered and rejected (Figure 4C) the same vaccine denoted by the column. Supporting Information 5 contains summary statistics of vaccine assessments depicted in Figure 4.

In Figure 4A, we consider patients who accepted their first vaccine offer. Notice that Pfizer has the highest average ratings across all columns. This implies that free vaccine choice does not necessarily mean that all patients get the most preferred

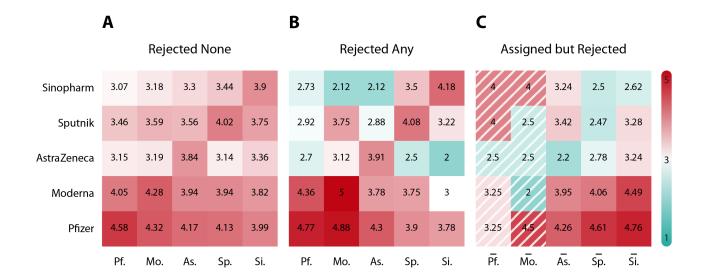


Figure 4. Average vaccine rating by assigned, re-selected and rejected vaccines. A Ratings of those patients who accepted their first vaccine offer. Individuals are grouped by columns of received vaccine. Rows correspond to average rating of a certain vaccine. B Ratings of those patients who rejected at least one assigned vaccine. Individuals are grouped by columns of received vaccine. Rows correspond to average rating of a certain vaccine. C Ratings of those patients who rejected at least one assigned vaccine. Individuals are grouped by columns of received vaccine. Individuals are grouped by columns of rejected vaccine (marked by overline). Rows correspond to average rating of a certain vaccine. The average rating in the first two shaded columns are calculated from few observations only.

vaccine. Instead, individuals make an uncertain decision considering acceptable alternatives and unknown waiting times between the sequential vaccine offers. In the Supporting Information, Table S12 presents the distribution of the most preferred vaccine when the accepted vaccine is not the most highly-rated vaccine. We infer that patients were willing to accept a vaccine other than the preferred Pfizer or Moderna to avoid the long anticipated waiting times. In Table S13, we show that patients had to wait almost 2.76 weeks on average to get their most preferred vaccine if they rejected a less preferred vaccine. In Table S14, we also present evidence that the anticipated waiting times closely follow the actual waiting times on average. That is, if a patient accepts the first vaccine offer then the following conditions must hold; the patient prefers the accepted vaccine to the other vaccines or she has a belief that a more preferred vaccine, which is Pfizer and Moderna with high likelihood, will not be offered anytime soon. We provide an extended analysis in the Supporting Information 6.

Figure 4B illustrates the ratings of those patients who rejected at least one vaccine. We see that the diagonal values in Figure 4B are slightly higher than the ones in Figure 4A. This implies that individuals who rejected any vaccines before value their received vaccine slightly better than those who accepted their first vaccine. Figure 4B also shows a high level of dislike towards AstraZeneca, Sinopharm and somewhat Sputnik when it comes to rating vaccines other than the accepted vaccine.

Comparing the mean values on Figure 4A and 4B, we can see fundamental differences between the rejected none and rejected any groups. Based on the values on the diagonal, those patients who rejected at least one vaccine rated the obtained vaccine higher than those who accepted the first vaccine offer. Looking at the off-diagonal, we see that the rejected any group tends to rate the non-accepted vaccines to be worse. For instance, the mean ratings for AstraZeneca at the off-diagonal are strictly smaller among the patients who rejected a vaccine at least once than among those who accepted their first offer. We also see those patients who received Sinopharm after rejection tend to rate other vaccines worse than those who accepted Sinopharm in their first offer.

Figure 4C highlights the benefits of free choice. We group patients based on the first vaccine offer rejection denoted by the columns. We implicitly assume here that the vaccine registration is not affected by the government's decision to allow free choice. Let us point out the number of observations is small for the Pfizer and Moderna groups with only 6 and 2 observations, respectively. In other words, patients rarely reject Pfizer or Moderna when these vaccines are offered for the first time. Focusing on the AstraZeneca, Sputnik and Sinopharm columns, we see that the values for the assigned vaccines on the diagonal are strictly smaller than any of alternatives. We can also see a strong preference towards Pfizer and Moderna from the AstraZeneca, the Sputnik and the Sinopharm columns. Figure 4C suggests that free choice significantly increased the efficiency of vaccine allocation because this avoided unnecessary risks of final rejections by enabling re-selecting still acceptable vaccines.

Assessment of vaccines might have dynamically changed as we learned their effects on health antibody production. We report how assessment of vaccines have changed over the third wave by vaccine types in Supporting Information 6. Most of these stay well below 5 percent of the total sample. Only in case of the AstraZeneca did the number of negative changes almost reach 10 percent of the sample, which is most likely due to the strong negative influence of the news about the side effects.

Finally, Supporting Information 7 contains a preference analysis of those individuals who were not vaccinated but evaluated minimum 2 but maximum 3 of the investigated vaccines acceptable. This enables us to eliminate those who cannot accept a specific vaccine to look at co-acceptance among vaccines. We find in Figure S6 that Pfizer is accepted by many who also accept Moderna (this is the most common co-acceptance link), Sputnik and AstraZeneca, but less so with Sinopharm. Yet, the Sputnik-Sinopharm co-acceptance link is the most frequent one that does not include Pfizer. This suggests that a wide and diverse vaccine portfolio including, both the newest technologies and the older techniques can provide good second-best choices in a society of divergent vaccine preferences.

In sum, we find that free choice across COVID-19 vaccine types allows individuals to get an "acceptable" vaccine and reject the unacceptable alternatives. Although mRNA vaccines are the most preferred ones in our sample, some individuals have specifically chosen vector-based and whole virus vaccines. Our results suggest that the society is fragmented into groups of individuals that either prefers the newest vaccine technologies and refuse the more established vaccine technologies or vice versa.

3 Discussion

The COVID-19 virus created an unprecedented challenge is human history. This is the first global pandemic in the digital age, which concerned all countries around the globe. Nations and international alliances, such as the European Union, started financing the development of vaccines in a diverse way, since the fast and reliable supply of effective vaccines seemed the only solution to end the pandemic. A large number of vaccine developments have started simultaneously, led by the producers of the main geopolitical powers (USA, EU, China, Russia and India). So far, several of these vaccines have entered the mass production phase but these use multiple technologies including inactivated whole viruses (eg. Sinopharm), adeno-virus vectors (eg. AstraZeneca, Janssen, Sputnik) and the mRNA technology (eg. Pfizer, Moderna). Most governments decided to use multiple types of vaccines in order to vaccinate the population as soon as possible.

There is a growing consensus that the vaccine diversity, generated by this extraordinary competition between vaccine developers, has several values. First of all, vast capacities have been developed with unprecedented speed when the outcomes - the efficiency of the vaccine - were still very uncertain^{33,34}. Many have also stressed the need for a production at full and increasing capacities; while others highlight that lower efficacy vaccines are better to use than wait for a more efficient one³⁵. Furthermore, individual vaccine preferences may differ by technology and country of origin that can be better handled with a diverse vaccine portfolio²⁶. This latter aspect raises the question whether governments should give the freedom of choice to the people over what type of vaccine they wish to get³⁶.

During the third wave of pandemic in Europe in the first half of 2021, which was also the first period of vaccination under vaccine shortage, most European countries gave no choice in between the available vaccines for the registered people when getting their shots. However scientific studies argued the potential benefits of free choice in the UK³⁷. Survey tests in Germany also suggested strong preferences in between the two most different vaccines available there in April, Pfizer and AstraZeneca, providing experimental evidence that free choice can mitigate vaccine hesitancy³⁸. A recent survey conducted in eight European countries shows the differences in vaccine hesitancy and also the divergent preferences of the people over the available vaccines³⁹. Our paper adds to this line of research by analyzing a representative survey in Hungary that not just ask the preferences of the patients, but collects also information about their actual decisions over accepting or rejecting offers for specific vaccines, thus providing their revealed preferences over the available vaccines.

The available vaccines differ in many characteristics, such as technology, country of origin, effectiveness, side effects, approval status by national and international authorities, etc. These aspects are very hard to measure, weight and compare. Even if experts are conducting the ranking based on sophisticated techniques, the result can be different⁴⁰ depending on the subjective valuations of the experts. It is then not a surprise that the average patients can have very diverse preferences, that can significantly vary across countries³⁹, and can also change over time²⁶, as more scientific information becomes available on the effectiveness of the vaccines in the long run and for new mutations, and more precise estimations on the risky side-effects for different groups of populations. Our survey shows surprisingly diverse preferences, where one of the most relevant aspect identified was the technology, namely mRNA versus vector-vaccine and whole-virus. The other major fragmentation in the preferences was in between over the EU-approved Western vaccines (Pfizer, Moderna, AztraZeneca and Janssen) and the Eastern vaccines (Sputnik and Sinopharm) that were additional vaccine types to the common EU purchase.

Besides describing the surveyed preferences and actual choices over the available vaccines, we aimed to understand the reasons behind the preferences. A novel finding of our study, that was not considered in previous international surveys³⁹, is the effect of information source that respondents trust. We found that people trusting in politicians were more likely to accept less popular vaccines, in our case vector-based (AstraZeneca and Sputnik) or whole-virus vaccines (Sinopharm). Those open to conspiracy theories were more likely to reject mRNA vaccines (Pfizer and Moderna). This shows that the free choice policy in vaccine allocation should be combined with the right communication strategy by the government and other decision makers, where different populations should be targeted with different messages when offered with a range of vaccines.

4 Methods

4.1 Data

The data used in this paper was collected with a CATI telephone survey on a nationally representative sample of the Hungarian adult (18 years or older) population between May 25-31, 2021. The sample size was 1,000, which is equal to the conventional sample size for nationally representative samples in the country. Data was collected both over mobile and landline phones. This campaign was a part of a larger data collection effort, which started in March 2020 and it is still ongoing. The Hungarian Data Provider Questionnaire^{30,31} continuously collects survey data through online questionnaires⁴¹ and is monthly asked on nationally representative phone surveys to follow the behavioural changes of people in terms of their contact dynamics, mobility patterns, self-protection habits, and ask people's opinion related to the COVID-19 pandemic. The sample was selected by a multi-step, proportionally stratified, probabilistic sampling procedure. The sample is representative for the Hungarian population aged 18 or older by gender, age, education and domicile. Minor deviations of the data from population ratios were corrected by iterative proportional weighting after the data collection. After data collection, only the anonymised and hashed data was shared with people involved in the project after signing non-disclosure agreements. The data collection was fully complying with the actual European and Hungarian privacy data regulations and was approved by the Hungarian National Authority for Data Protection and Freedom of Information⁴², and also by the Research Ethic Committee of the Medical Research Council of Hungary (resolution number IV/3073-1/2021/EKU). Informed consent was obtained from all survey participants.

4.2 Availability of Data

For privacy protection we share only the relevant part of the data, which does not allow the reidentification of participants. We exclude sensitive information such as domicile and labels for categorical responses. The data is available at https:

//doi.org/10.5281/zenodo.5575586.

4.3 Methods

In the LASSO model, the dependent variable takes values between 1 and 5, where 1 is "unacceptable" and 5 is "the best"). In models 1-5, we estimate evaluation values by vaccine types. Then, in models 6 to 9, the ratings' variance, mean, minimum, and maximum serve as dependent variables. In model 10, the dependent variable is binary taking the value of 1 if ever rejected a vaccine offer and 0 otherwise. The dependent variable is 1 in model 11, if any vaccine is unacceptable. The lasso coefficients $(\hat{\beta}_{\lambda}^{L})$ minimize the following expression: $\sum_{i=1}^{n} (y_i - \beta_0 - \sum_{j=1}^{p} \beta_j x_{ij})^2 + \lambda \sum_{j=1}^{p} |\beta_j|$. Linear probability models are used to predict rejection of vaccines. These regressions are specified by the formula

Linear probability models are used to predict rejection of vaccines. These regressions are specified by the formula $P(Y = 1|X) = \alpha + \beta X + \varepsilon$, where Y equals 1 if the vaccine is rated unacceptable, α is the intercept, β denotes point estimates, X stand for a vector of independent variables, and ε is heteroscedasticity-robust standard error.

In the heat-maps generation, we used the same vaccine ratings on a scale from 1 to 5, as we mentioned above. We clustered the survey responses based on the acceptance decision of the first vaccine offer. In case of rejection, we grouped individuals based on the accepted vaccine and the first rejected vaccine. Then, we used standard summary statistics tools to find the number of observation, mean and the standard deviation for each of the groups. We report the mean ratings in Figure 3 and the rest of the summary statistics in Supporting Information 5.2.

The link weights of the directed network presented in Figure 1 is generated following the aggregation procedure $W_{ab} = \sum_a L_{ab}$, where L_{ab} equals 1 if the individual has rejected vaccine *a* but later received vaccine *b*. The undirected co-occurrence network presented in Figure S6 follows the rule of $W_{mn} = \sum_m L_{mn,m\neq n}$ where L_{mn} equals 1 if the respondent accepts both vaccines *m* and *n*.

All methods were performed in accordance with the guidelines and regulations of the Hungarian National Authority for Data Protection and Freedom of Information⁴² and were approved by the Research Ethic Committee of the Medical Research Council of Hungary (resolution number IV/3073-1/2021/EKU).

References

- 1. Cohen, J. Vaccine designers take first shots at COVID-19 (2020).
- 2. Graham, B. S. Rapid COVID-19 vaccine development. Science 368, 945-946 (2020).
- 3. Corey, L., Mascola, J. R., Fauci, A. S. & Collins, F. S. A strategic approach to COVID-19 vaccine R&D. <u>Science</u> 368, 948–950 (2020).
- 4. Lazarus, J. V. et al. A global survey of potential acceptance of a COVID-19 vaccine. Nat. Medicine 27, 225–228 (2021).
- 5. Machingaidze, S. & Wiysonge, C. S. Understanding COVID-19 vaccine hesitancy. Nat. Medicine 1–2 (2021).
- 6. Sallam, M. COVID-19 vaccine hesitancy worldwide: a concise systematic review of vaccine acceptance rates. <u>Vaccines</u> 9, 160 (2021).
- Solís Arce, J. S. et al. COVID-19 vaccine acceptance and hesitancy in low-and middle-income countries. <u>Nat. Medicine</u> 27 (2021).
- Wouters, O. J. et al. Challenges in ensuring global access to COVID-19 vaccines: production, affordability, allocation, and deployment. <u>The Lancet</u> (2021).
- 9. Giordano, G. et al. Modeling vaccination rollouts, SARS-CoV-2 variants and the requirement for non-pharmaceutical interventions in Italy. Nat. Medicine 27, 993–998 (2021).
- **10.** Bubar, K. M. et al. Model-informed COVID-19 vaccine prioritization strategies by age and serostatus. <u>Science</u> **371**, 916–921 (2021).
- 11. Matrajt, L., Eaton, J., Leung, T. & Brown, E. R. Vaccine optimization for COVID-19: Who to vaccinate first? <u>Sci. Adv.</u> 7, eabf1374 (2021).
- 12. Priesemann, V. et al. Towards a european strategy to address the COVID-19 pandemic. The Lancet 398, 838–839 (2021).
- 13. Rosselli, R., Martini, M. & Bragazzi, N. L. The old and the new: vaccine hesitancy in the era of the web 2.0. challenges and opportunities. J. Prev. Medicine Hyg. 57, E47 (2016).
- Loomba, S., de Figueiredo, A., Piatek, S. J., de Graaf, K. & Larson, H. J. Measuring the impact of COVID-19 vaccine misinformation on vaccination intent in the UK and USA. <u>Nat. Hum. Behav.</u> 5, 337–348 (2021).
- Paul, E., Steptoe, A. & Fancourt, D. Attitudes towards vaccines and intention to vaccinate against covid-19: Implications for public health communications. <u>The Lancet Reg. Heal.</u> 1, 100012 (2021).

- Sonawane, K., Troisi, C. L. & Deshmukh, A. A. COVID-19 vaccination in the UK: Addressing vaccine hesitancy. <u>The Lancet Reg. Heal.</u> 1 (2021).
- 17. Troiano, G. & Nardi, A. Vaccine hesitancy in the era of COVID-19. Public Heal. (2021).
- Freeman, D. et al. Effects of different types of written vaccination information on covid-19 vaccine hesitancy in the uk (oceans-iii): a single-blind, parallel-group, randomised controlled trial. The Lancet Public Heal. 6, e416–e427 (2021).
- Lindholt, M. F., Jørgensen, F., Bor, A. & Petersen, M. B. Public acceptance of covid-19 vaccines: cross-national evidence on levels and individual-level predictors using observational data. BMJ Open 11, e048172 (2021).
- **20.** Schernhammer, E. <u>et al.</u> Correlates of COVID-19 vaccine hesitancy in austria: trust and the government. J. Public Heal. (Oxford, England).
- **21.** Salali, G. D. & Uysal, M. S. COVID-19 vaccine hesitancy is associated with beliefs on the origin of the novel coronavirus in the UK and Turkey. Psychol. Medicine 1–3 (2020).
- **22.** Murphy, J. <u>et al.</u> Psychological characteristics associated with COVID-19 vaccine hesitancy and resistance in Ireland and the United Kingdom. Nat. Commun. **12**, 1–15 (2021).
- 23. Soveri, A., Karlsson, L. C., Antfolk, J., Lindfelt, M. & Lewandowsky, S. Unwillingness to engage in behaviors that protect against covid-19: the role of conspiracy beliefs, trust, and endorsement of complementary and alternative medicine. BMC Public Heal. 21, 1–12 (2021).
- 24. Lipsitch, M. & Dean, N. E. Understanding COVID-19 vaccine efficacy. Science 370, 763–765 (2020).
- 25. Ledford, H. How could a COVID vaccine cause blood clots? scientists race to investigate. Nature 592, 334–335 (2021).
- **26.** Schwarzinger, M., Watson, V., Arwidson, P., Alla, F. & Luchini, S. Covid-19 vaccine hesitancy in a representative working-age population in france: a survey experiment based on vaccine characteristics. <u>The Lancet Public Heal.</u> **6**, e210–e221 (2021).
- Dror, A. A. et al. Vaccine hesitancy due to vaccine country of origin, vaccine technology, and certification. Eur. J. Epidemiol. 1–6 (2021).
- 28. Goodwin, R. et al. Psychological factors underpinning vaccine willingness in Israel, Japan and Hungary. medRxiv (2021).
- **29.** Koltai, J., Karsai, M., Oroszi, B. & Röst, G. Dynamics of attitudes towards COVID-19 vaccines in Hungary. Work in Progress.
- **30.** Karsai, M., Koltai, J., Vásárhelyi, O. & Röst, G. Hungary in mask/MASZK in Hungary. <u>Corvinus J. Sociol. Soc. Policy</u> (2020).
- **31.** Koltai, J., Vásárhelyi, O., Röst, G. & Karsai, M. Monitoring behavioural responses during pandemic via reconstructed contact matrices from online and representative surveys. <u>arXiv preprint arXiv:2102.09021</u> (2021).
- **32.** Luo, J. & Yu, R. The spreading of alternatives: Is it the perceived choice or actual choice that changes our preference? J. Behav. Decis. Mak. **30**, 484–491 (2017).
- 33. Andreadakis, Z. et al. The COVID-19 vaccine development landscape. Nat. Rev. Drug Discov. 19, 305–306 (2020).
- Price, W. N., Rai, A. K. & Minssen, T. Knowledge transfer for large-scale vaccine manufacturing. <u>Science</u> 369, 912–914 (2020).
- 35. Castillo, J. C. et al. Market design to accelerate covid-19 vaccine supply. Science 371, 1107–1109 (2021).
- Hughes, M. T. et al. Opinion: The importance of offering vaccine choice in the fight against covid-19. Proc. Natl. Acad. Sci. 118 (2021).
- Dal-Ré, R., Stephens, R. & Sreeharan, N. Let me choose my covid-19 vaccine. <u>Eur. J. Intern. Medicine</u> 87, 104–105 (2021).
- **38.** Sprengholz, P., Eitze, S., Korn, L., Siegers, R. & Betsch, C. The power of choice: Experimental evidence that freedom to choose a vaccine against covid-19 improves willingness to be vaccinated. Eur. J. Intern. Medicine (2021).
- **39.** Steinert, J. <u>et al.</u> COVID-19 vaccine hesitancy in eight european countries: Prevalence, determinants and heterogeneity. (2021).
- **40.** Abdelwahab, S. F., Issa, U. H. & Ashour, H. M. A novel vaccine selection decision-making model (vsdmm) for COVID-19. Vaccines **9**, 718 (2021).
- 41. Karsai, M., Koltai, J. & Röst, G. Hungarian data provider questionnaire. https://COVID.sed.hu/ (date of access 2021.09.27).
- 42. Nemzeti Adatvédelmi és Információszabadság Hatóság, https://www.naih.hu (date of access 2020.12.).

Notes

ⁱVaccine diversity is simply calculated by the number of available vaccine types from data at https://ourworldindata.org/covid-vaccinations.

ⁱⁱOfficial names of the vaccines are listed here. **Pfizer-Biontech**: Comirnaty/BNT162b2, COVID-19 mRNA vaccine. **Moderna**: Spikevax/mRNA-1273. **AstraZeneca**: Vaxzevria, Covishield / AZD1222, ChAdOx1 nCoV-19, ChAdOx1-S, AZD2816. **Sputnik**: Gam-COVID-Vac. Sinopharm: **Sinopharm**: BIBP COVID-19 vaccine / Zhong'aikewei, Hayat-Vax. **Janssen**: Ad26.COV2.S JNJ-78436735 Ad26COVS1 VAC31518

ⁱⁱⁱAccording to https://ourworldindata.org/covid-vaccinations, vaccination per capita was much higher in May 2021, when our data was collected, in Hungary than in other Central and Eastern European countries where trust in public authorities is at a similar level.

^{iv}Number of cases can vary among analyses. For a precise consideration of sample sizes, please see Materials and Methods.

^vhttps://vakcinainfo.gov.hu/regisztracio-oltasra

^{vi}Recommendation for vaccination strategy of the Hungarian Academy of Sciences: https://mta.hu/english/recommendation-of-the-hungarian-academy-ofscience-on-the-management-of-the-covid-19-epidemic-in-the-short-and-long-term-110630

vii A summary on anti-vax movement in Hungary: https://www.iribeaconproject.org/our-work-analysis-and-insights/2021-09-16/hungarys-anti-vax-movementalive-and-kicking

viii Scientists, doctors slam Hungarian government criticism of western vaccines: https://www.reuters.com/article/us-health-coronavirus-hungary-vaccinesidUSKBN2CE2C3

^{ix} The Hungarian prime minister is vaccinated with Sinopharm: https://www.euronews.com/2021/02/28/hungary-s-pm-viktor-orban-vaccinated-against-covid-with-chinese-sinopharm-vaccine

^xSurveys https://hungarytoday.hu/hungarian-vaccination-politics-vaccine-preference-coronavirus-survey-willingness/

xihttps://www.reuters.com/business/healthcare-pharmaceuticals/uk-regulator-says-some-evidence-astrazeneca-clots-occur-more-women-than-men-2021-05-06/

Acknowledgements

The authors acknowledge the help of Szabolcs Tóth-Zs in finalizing the figures. Péter Biró gratefully acknowledges financial support from the Hungarian Academy of Sciences, Momentum Grant (No. LP2021-2). Julia Koltai acknowledges funding from the Premium Postdoctoral Grant of the Hungarian Academy of Sciences. The work of Balazs Lengyel was financially supported by Hungarian National Scientific Fund (OTKA K 138970). Agnes Szabo-Morvai was financially supported by the National Research, Development and Innovation Office (OTKA FK141322)

Author contributions statement

J.K., A.Sz., G.R., M.K., P.B. and B.L. conceived the survey data collection, K.K., A.Sz., P.B., B.L. designed the investigation, K.K., A.Sz., and B.L. analyzed the data. All authors wrote and reviewed the manuscript.

5 Additional information

5.1 Accession codes

All codes to analyze data have been written in R (version 4.0.5) and are available at https://doi.org/10.5281/zenodo.5575586.

5.2 Competing interests

All authors claim no conflicting interest.

Supplementary Information

Supporting Information 1: Detailed information on vaccination dynamics

Figure S1 illustrates the time horizon for the vaccination. We can see that Pfizer was the only available vaccine initially, which was most likely offered to medical workers. In the beginning of March, Sinopharm and AstraZeneca became available for mostly chronic ill patients. We see a strict cutoff at the age of 60, where Astrazeneca was offered to patients below 60 and the Chinese Sinopharm to those who are above 60. Chronic ill patients who rejected a vaccine get vaccinated in the end of March. In April, we see that vaccines become widely available for patients under the age of 50. In May, those patients who have rejected a vaccine before and are not chronic ill get finally vaccinated, mainly with Pfizer. Figure S1 suggests that chronic ill and older patients who have not rejected any vaccines in the past tend to get vaccinated sooner.

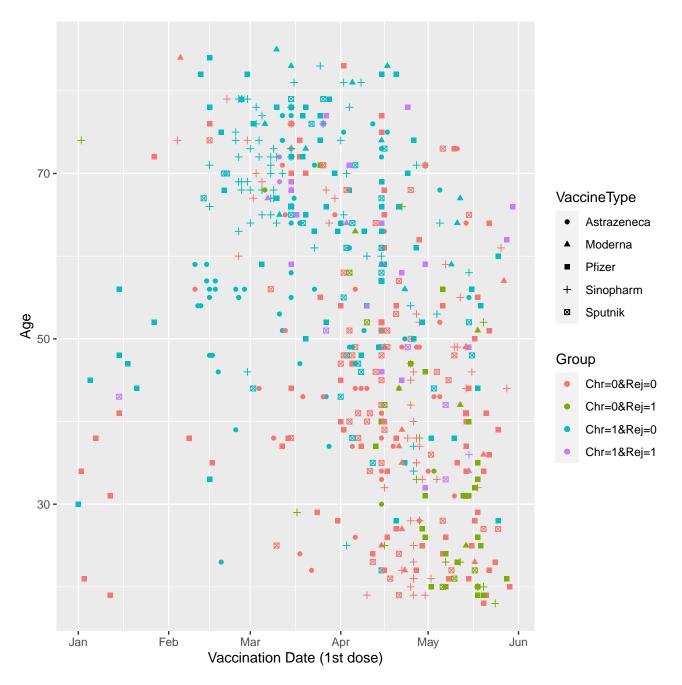
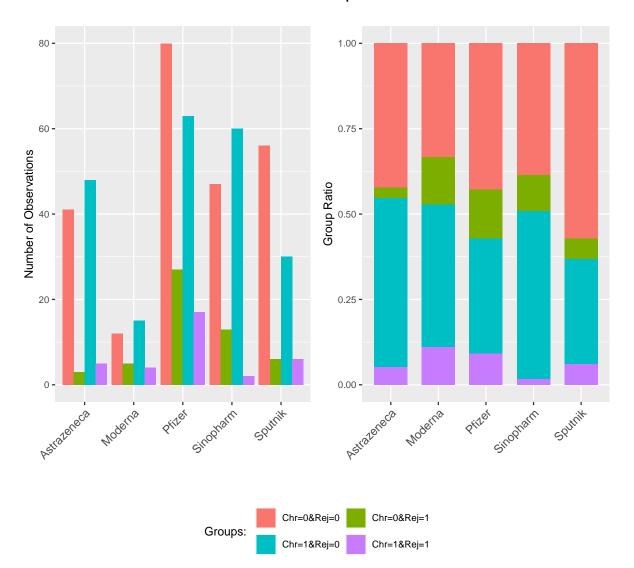


Figure S1. Vaccination date, age, vaccination type, chronic illness and rejection status distribution

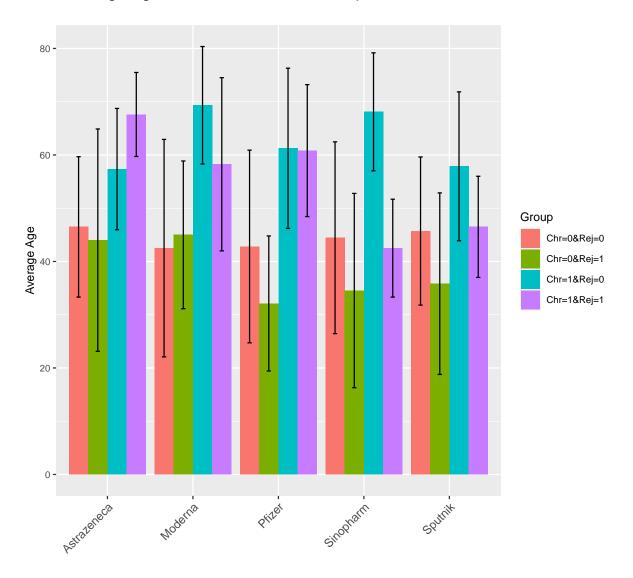
Figure S2 shows the distribution of different vaccine types conditionally on chronic illness and rejection status. Out of the 996 observations, most patients received Pfizer (n = 260), then Sinopharm (n = 162), Sputnik (n = 136), Astrazeneca (n = 122), Moderna (n = 45) and n = 261 people decided not to get vaccinated. We see that only a minority of patients rejected any vaccines. We would like to highlight that Pfizer and Moderna had the highest and AstraZeneca had the lowest ratio of patients who rejected a vaccine before vaccination. We would also like to point out that Sputnik and Pfizer have the two highest not chronic ill ratio of patients. More importantly, Figure S2 suggests that chronic ill patients are less likely to reject a vaccine than patients without any chronic illness.



Within Vaccine Group Distribution

Figure S2. Vaccine type distribution conditionally on chronic illness and rejection status.

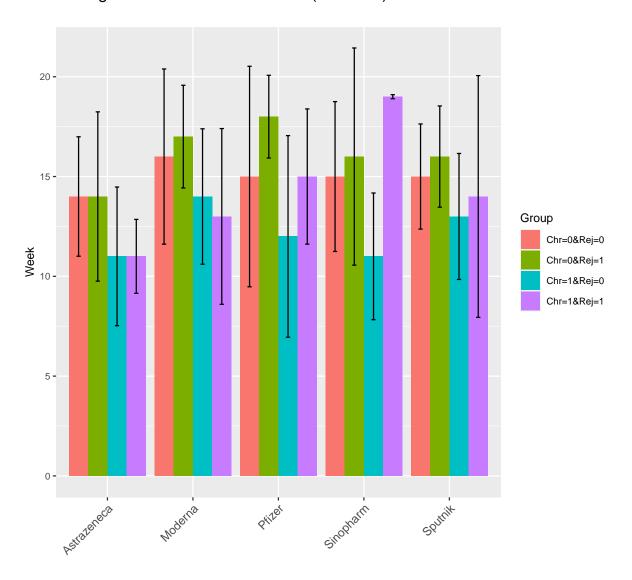
Figure S3 shows the mean and standard deviation of the age distribution conditionally on vaccine type, chronic illness and rejection status. We see that chronic patients tend to be older, as expected. Among patients without chronic illness, Pfizer has the lowest average age, which we believe is a consequence of 2 reasons. First, younger medical workers and school teachers could receive Pfizer sooner, as they were prioritized in the vaccine allocation mechanism by the government. Second, younger patients with better physical health felt less threatened by the pandemic and more inclined to wait for their most preferred vaccine, which was Pfizer in most cases. On the other hand, we see that patients without chronic illness, who received AstraZeneca have the highest average age. A potential explanation could be that elderly patient has a stronger time pressure and hence less likely to wait for alternative vaccines. Figure S3 shows a lot of variation among vaccine types and chronic illness status, which could be explained by prioritization of patients with certain jobs and the amount of time pressure patients experience to wait for an alternative vaccine.



Avearage Age Within Vaccines and Groups

Figure S3. Average age distribution conditionally on vaccine type, chronic illness and rejection status. The black line shows 1 standard deviation.

Figure S4 shows the average week of the first vaccination among vaccine types, chronic illness and rejection status. We see that the average vaccination week tends to increase among those who rejected a vaccine regardless of chronic illness status and vaccination type. Let us point out the large variance in the vaccination date among patients who received Pfizer. This indicates that Pfizer was available for the longest time in Hungary. Also, the smaller values for AstraZeneca suggests that patients were more likely to receive AstraZeneca earlier, in February and early March. Figure S4 is in accordance with Figure S1 as they both show that vaccines became available for the most patient starting from the 14th week and patients with Pfizer have the largest variation in the time of vaccination.



Average vaccination week in 2021 (1st Dose)

Figure S4. Average vaccination week distribution in 2021 conditionally on vaccine type, chronic illness and rejection status. The black line shows 1 standard deviation.

Supporting Information 2: Variable selection for the vaccine rejection analysis

	Pfizer	Moderna	AstraZeneca	Sputnik	Sinopharm	Variance	Mean	Min	Max	Any
	(1)	(2)	(3)	(4)	(5)	(6)	(7)	(8)	(9)	(10)
Age	Х	Х	Х				Х	х		х
Female		Х	Х	Х						
City Cat. by Size	Х	Х	Х	Х						
University		Х		Х		Х		Х		
Smoking		Х	Х	Х	Х					х
Chronic Illness			Х	Х				х		
Acute Illness				Х		Х				
Worked Last Week				Х		Х		х		
Wealth Pre COVID-19		х								
Wealth Now		Х	Х	Х	Х	Х	х	х		х
Adv. from Doct.	х	х	Х	Х		Х	х		х	х
Adv. from Pol.				Х		Х				х
Adv. from Sci.	х	х	Х	Х		Х		х	х	х
Adv. from Anti-vacc.	х	х	Х	Х		Х		х		х
Prop.										
Adv. from Family								х		
Adv. from Friends				Х						
Adv. from Celeb.						х				
Adv. from Journ.			Х					х		
Online News			Х	х	Х	Х		х		х

Table S1. Lasso models on vaccine rejection

Note: Dependent variables: Pfizer and other vaccines: rating (1: "unacceptable"; 5: "best"); Variance/Mean/Min/Max: variance/mean/min/max of the ratings; UnacceptAny=1 if any vaccine rated as "unacceptable". Explanatory variables excluded from each lasso model: SourceSocialNetwork; SourcePress; SourceRadio; SourceTV; SourceFamily; SourceFriends; SourceOther (VaccineX: Takes advice regarding COVID-19 vaccination from X; SourceX: Source of information used to get to know about pandemic-related government measures)

Supporting Information 3: Variable description: summary statistics, correlation, VIF

Variable	Min	1st Qu.	Median	Mean	3rd Qu.	Max	VIF
Advice from Doctors	0.000	0.000	1.000	0.729	1.000	1.000	1.121
Advice from Scientists	0.000	0.000	1.000	0.595	1.000	1.000	1.168
Advice from Anti-vaccine Propagators	0.000	0.000	0.000	0.042	0.000	1.000	1.092
Advice from Politicians	0.000	0.000	0.000	0.065	0.000	1.000	1.236
Advice from Family	0.000	0.000	0.000	0.222	0.000	1.000	1.536
Advice from Friends	0.000	0.000	0.000	0.136	0.000	1.000	1.506
Advice from Journalists	0.000	0.000	0.000	0.017	0.000	1.000	1.586
Advice from Celebrities	0.000	0.000	0.000	0.015	0.000	1.000	1.583
Age	18.00	34.00	48.00	48.86	64.00	85.00	1.542
Female	0.00	0.00	1.00	0.53	1.00	1.00	1.075
University	0.000	.0.000	0.000	0.217	0.000	1.000	1.181
City Cat. by Size	1.000	1.000	2.000	2.296	3.000	4.000	1.073
Wealth Pre COVID-19	1.000	5.000	5.000	5.406	6.000	10.000	1.141
Smoking	1.000	1.000	2.000	2.208	4.000	4.000	1.114
Chronic Illness	0.000	0.000	0.000	0.406	1.000	1.000	1.403
Covid-19 Previously	0.000	0.000	0.000	0.095	0.000	1.000	1.157
Serious Covid-19 Previously	0.00	0.00	0.00	0.01	0.00	.00	1.123

Table S2. Summary statistics and VIF of variables in the linear probability model in Figure 2

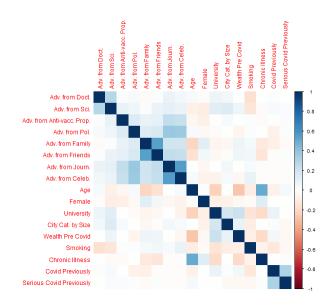


Figure S5. Pearson correlation coefficients of variables in the linear probability model in Figure 2

Supporting Information 4: Alternative specifications of the vaccine rejection analysis

	Depe	endent variable	e: Evaluating a vo	accine unacce _l	otable
	Pfizer	Moderna	Astrazeneca	Sputnik	Sinopharm
	(1)	(2)	(3)	(4)	(5)
Advice from Doctors	-0.057***	-0.061***	-0.029	-0.026	-0.054**
	(0.018)	(0.019)	(0.023)	(0.019)	(0.025)
Advice from Scientists	-0.035**	-0.026^{*}	-0.054**	-0.017	0.008
	(0.014)	(0.014)	(0.022)	(0.017)	(0.021)
Advice from Anti-Vaccine Propagators	0.202***	0.182***	0.117*	0.062	0.054
	(0.064)	(0.066)	(0.066)	(0.056)	(0.064)
Advice from Politicians	-0.029	-0.035	-0.064**	-0.076***	-0.084***
	(0.023)	(0.023)	(0.031)	(0.014)	(0.030)
Advice from Family	-0.043***	-0.038**	0.028	-0.019	-0.058**
,	(0.017)	(0.019)	(0.032)	(0.026)	(0.028)
Advice from Friends	-0.007	-0.007	-0.015	-0.025	-0.012
	(0.020)	(0.024)	(0.039)	(0.030)	(0.034)
Advice from Journalists	-0.071	-0.032	0.023	-0.008	0.089
	(0.049)	(0.060)	(0.100)	(0.058)	(0.138)
Advice from Celebrities	0.073	0.116	0.047	0.129	0.114
	(0.092)	(0.102)	(0.122)	(0.103)	(0.148)
Age	-0.001**	-0.001**	-0.003***	-0.0003	-0.0001
	(0.0004)	(0.0005)	(0.001)	(0.001)	(0.001)
Female	0.006	0.001	0.045**	-0.005	0.005
	(0.013)	(0.014)	(0.020)	(0.017)	(0.021)
University	-0.021	-0.026^{*}	-0.054**	-0.026	0.002
	(0.015)	(0.015)	(0.024)	(0.018)	(0.024)
City Category by Size	0.0002	-0.005	-0.005	-0.004	0.010
	(0.005)	(0.006)	(0.009)	(0.008)	(0.010)
Smoking	0.004	0.005	0.004	0.011	0.020**
-	(0.005)	(0.006)	(0.008)	(0.007)	(0.008)
Chronic Illness	-0.004	-0.004	0.008	-0.011	0.002
	(0.016)	(0.015)	(0.023)	(0.021)	(0.025)
COVID-19 Previously	0.007	0.002	-0.010	-0.034	-0.045
	(0.023)	(0.023)	(0.035)	(0.022)	(0.031)
Serious COVID-19 Previously	0.151	0.149	0.031	-0.031	0.037
	(0.109)	(0.109)	(0.094)	(0.028)	(0.101)
Constant	0.146***	0.166***	0.314***	0.120**	0.095*
	(0.038)	(0.041)	(0.054)	(0.047)	(0.055)
Observations	999	999	999	999	999
R^2	0.095	0.080	0.070	0.026	0.028
Adjusted R^2	0.080	0.065	0.055	0.010	0.012

Table S3. Linear probability models on evaluating a vaccine unacceptable. Results depicted in Figure 3

Note:

*p<0.1; **p<0.05; ***p<0.01

	Depe	ndent variable	e: Evaluating a v	accine unaccej	ptable
	Pfizer	Moderna	Astrazeneca	Sputnik	Sinopharm
	(1)	(2)	(3)	(4)	(5)
Advice from Doctors	-0.130***	-0.130***	-0.041	-0.046	-0.076^{**}
	(0.040)	(0.039)	(0.033)	(0.034)	(0.035)
Advice from Scientists	-0.087^{**}	-0.062^{*}	-0.085^{**}	-0.034	0.013
	(0.034)	(0.034)	(0.034)	(0.034)	(0.033)
Advice from Anti-Vaccine Propagators	0.207***	0.174***	0.076*	0.049	0.035
	(0.066)	(0.063)	(0.043)	(0.045)	(0.041)
Advice from Politicians	-0.036	-0.041	-0.051^{**}	-0.074^{***}	-0.066^{***}
	(0.029)	(0.027)	(0.024)	(0.014)	(0.023)
Advice from Family	-0.092^{***}	-0.075^{**}	0.038	-0.031	-0.077^{**}
	(0.035)	(0.038)	(0.043)	(0.043)	(0.037)
Advice from Friends	-0.011	-0.011	-0.016	-0.034	-0.013
	(0.035)	(0.039)	(0.043)	(0.041)	(0.037)
Advice from Journalists	-0.047	-0.020	0.010	-0.004	0.037
	(0.032)	(0.037)	(0.041)	(0.030)	(0.057)
Advice from Celebrities	0.045	0.067	0.018	0.062	0.044
	(0.057)	(0.059)	(0.047)	(0.050)	(0.057)
Age	-0.089^{**}	-0.086^{**}	-0.200^{***}	-0.019	-0.003
	(0.041)	(0.041)	(0.039)	(0.042)	(0.040)
Female	0.016	0.002	0.072**	-0.010	0.008
	(0.033)	(0.033)	(0.032)	(0.034)	(0.033)
University	-0.044	-0.051^{*}	-0.072^{**}	-0.042	0.003
	(0.031)	(0.029)	(0.032)	(0.029)	(0.032)
City Category by Size	0.001	-0.027	-0.018	-0.017	0.033
	(0.028)	(0.029)	(0.029)	(0.033)	(0.034)
Smoking	0.023	0.028	0.018	0.054	0.081**
	(0.035)	(0.035)	(0.034)	(0.035)	(0.034)
Chronic Illness	-0.009	-0.009	0.013	-0.022	0.004
	(0.039)	(0.036)	(0.036)	(0.040)	(0.040)
COVID-19 Previously	0.010	0.003	-0.009	-0.040	-0.042
	(0.034)	(0.032)	(0.033)	(0.026)	(0.029)
Serious COVID-19 Previously	0.076	0.071	0.010	-0.012	0.012
	(0.055)	(0.052)	(0.030)	(0.011)	(0.032)
Constant	0.0003	0.0003	0.0001	0.0001	0.0003
	(0.030)	(0.031)	(0.031)	(0.031)	(0.031)
Observations	999	999	999	999	999
\mathbb{R}^2	0.095	0.080	0.070	0.026	0.028
Adjusted R^2	0.080	0.065	0.055	0.010	0.012

Table S4. Linear probabilit	y models on evaluating	a vaccine unacceptal	ble with standardized variables

Note:

*p < 0.1; **p < 0.05; ***p < 0.01

Moderna (2) -1.211*** (0.358) -0.682* (0.379) 2.349*** (0.549) -1.228 (1.016) -1.041* (0.554) -0.343 (0.672) -0.544	Astrazeneca (3) -0.298 (0.229) -0.573** (0.236) 0.974** (0.459) -0.939 (0.635) 0.138 (0.278) -0.066 (0.226)	Sputnik (4) -0.381 (0.266) -0.295 (0.271) 0.938 (0.588) -3.267*** (0.844) -0.429 (0.441)	Sinopharm (5) -0.510** (0.217) 0.039 (0.219) 0.495 (0.517) -1.474* (0.834) -0.722**
$\begin{array}{c} -1.211^{***}\\ (0.358)\\ -0.682^{*}\\ (0.379)\\ 2.349^{***}\\ (0.549)\\ -1.228\\ (1.016)\\ -1.041^{*}\\ (0.554)\\ -0.343\\ (0.672) \end{array}$	$\begin{array}{c} -0.298 \\ (0.229) \\ -0.573^{**} \\ (0.236) \\ 0.974^{**} \\ (0.459) \\ -0.939 \\ (0.635) \\ 0.138 \\ (0.278) \\ -0.066 \end{array}$	$\begin{array}{c} -0.381 \\ (0.266) \\ -0.295 \\ (0.271) \\ 0.938 \\ (0.588) \\ -3.267^{***} \\ (0.844) \\ -0.429 \\ (0.441) \end{array}$	$\begin{array}{c} -0.510^{**}\\ (0.217)\\ 0.039\\ (0.219)\\ 0.495\\ (0.517)\\ -1.474^{*}\\ (0.834)\end{array}$
$\begin{array}{c} (0.358) \\ -0.682^{*} \\ (0.379) \\ 2.349^{***} \\ (0.549) \\ -1.228 \\ (1.016) \\ -1.041^{*} \\ (0.554) \\ -0.343 \\ (0.672) \end{array}$	$\begin{array}{c} (0.229) \\ -0.573^{**} \\ (0.236) \\ 0.974^{**} \\ (0.459) \\ -0.939 \\ (0.635) \\ 0.138 \\ (0.278) \\ -0.066 \end{array}$	$\begin{array}{c} (0.266) \\ -0.295 \\ (0.271) \\ 0.938 \\ (0.588) \\ -3.267^{***} \\ (0.844) \\ -0.429 \\ (0.441) \end{array}$	$\begin{array}{c} (0.217) \\ 0.039 \\ (0.219) \\ 0.495 \\ (0.517) \\ -1.474^{*} \\ (0.834) \end{array}$
$\begin{array}{c} -0.682^{*} \\ (0.379) \\ 2.349^{***} \\ (0.549) \\ -1.228 \\ (1.016) \\ -1.041^{*} \\ (0.554) \\ -0.343 \\ (0.672) \end{array}$	$\begin{array}{c} -0.573^{**} \\ (0.236) \\ 0.974^{**} \\ (0.459) \\ -0.939 \\ (0.635) \\ 0.138 \\ (0.278) \\ -0.066 \end{array}$	$\begin{array}{c} -0.295 \\ (0.271) \\ 0.938 \\ (0.588) \\ -3.267^{***} \\ (0.844) \\ -0.429 \\ (0.441) \end{array}$	$\begin{array}{c} 0.039\\ (0.219)\\ 0.495\\ (0.517)\\ -1.474^{*}\\ (0.834) \end{array}$
$\begin{array}{c} -0.682^{*} \\ (0.379) \\ 2.349^{***} \\ (0.549) \\ -1.228 \\ (1.016) \\ -1.041^{*} \\ (0.554) \\ -0.343 \\ (0.672) \end{array}$	$\begin{array}{c} (0.236) \\ 0.974^{**} \\ (0.459) \\ -0.939 \\ (0.635) \\ 0.138 \\ (0.278) \\ -0.066 \end{array}$	$\begin{array}{c} -0.295 \\ (0.271) \\ 0.938 \\ (0.588) \\ -3.267^{***} \\ (0.844) \\ -0.429 \\ (0.441) \end{array}$	$\begin{array}{c} 0.039\\ (0.219)\\ 0.495\\ (0.517)\\ -1.474^{*}\\ (0.834) \end{array}$
$\begin{array}{c} 2.349^{***} \\ (0.549) \\ -1.228 \\ (1.016) \\ -1.041^{*} \\ (0.554) \\ -0.343 \\ (0.672) \end{array}$	$\begin{array}{c} 0.974^{**} \\ (0.459) \\ -0.939 \\ (0.635) \\ 0.138 \\ (0.278) \\ -0.066 \end{array}$	$\begin{array}{c} 0.938 \\ (0.588) \\ -3.267^{***} \\ (0.844) \\ -0.429 \\ (0.441) \end{array}$	$\begin{array}{c} 0.495 \\ (0.517) \\ -1.474^{*} \\ (0.834) \end{array}$
$\begin{array}{c} 2.349^{***} \\ (0.549) \\ -1.228 \\ (1.016) \\ -1.041^{*} \\ (0.554) \\ -0.343 \\ (0.672) \end{array}$	$\begin{array}{c} 0.974^{**} \\ (0.459) \\ -0.939 \\ (0.635) \\ 0.138 \\ (0.278) \\ -0.066 \end{array}$	$\begin{array}{c} 0.938 \\ (0.588) \\ -3.267^{***} \\ (0.844) \\ -0.429 \\ (0.441) \end{array}$	$\begin{array}{c} 0.495 \\ (0.517) \\ -1.474^{*} \\ (0.834) \end{array}$
-1.228 (1.016) -1.041* (0.554) -0.343 (0.672)	-0.939 (0.635) 0.138 (0.278) -0.066	$\begin{array}{c} -3.267^{***} \\ (0.844) \\ -0.429 \\ (0.441) \end{array}$	-1.474^{*} (0.834)
-1.228 (1.016) -1.041* (0.554) -0.343 (0.672)	-0.939 (0.635) 0.138 (0.278) -0.066	$\begin{array}{c} -3.267^{***} \\ (0.844) \\ -0.429 \\ (0.441) \end{array}$	-1.474^{*} (0.834)
(1.016) -1.041* (0.554) -0.343 (0.672)	(0.635) 0.138 (0.278) -0.066	-0.429 (0.441)	(0.834)
-1.041^{*} (0.554) -0.343 (0.672)	0.138 (0.278) -0.066	-0.429 (0.441)	
(0.554) -0.343 (0.672)	(0.278) -0.066	(0.441)	
-0.343 (0.672)	-0.066		(0.363)
(0.672)		-0.461	-0.176
· · · ·	(0.338)	(0.609)	(0.444)
	0.380	0.790	1.221
(1.644)	(1.211)	(0.901)	(1.515)
2.947***	0.910	2.621**	1.347
(0.905)	(1.341)	(1.018)	(1.390)
-0.026**	-0.037***	-0.004	-0.001
(0.013)	(0.008)	(0.009)	(0.007)
-0.006	0.467**	-0.060	0.046
(0.355)	(0.231)	(0.269)	(0.215)
-0.914^{*}	-0.525^{*}	-0.513	0.033
(0.518)	(0.300)	(0.392)	(0.265)
-0.179	-0.086	-0.070	0.105
(0.163)	(0.104)	(0.134)	(0.104)
0.127	0.042	0.176*	0.204**
(0.130)	(0.088)	(0.104)	(0.083)
-0.092	0.103	-0.203	0.017
(0.409)	(0.276)	(0.338)	(0.262)
0.005	-0.090	-0.811	-0.558
(0.592)	(0.394)	(0.667)	(0.459)
			0.395
			(1.235)
			-2.295^{**}
-0.557			(0.582)
	. ,	· /	999
(0.850)			701.087
	999	$\begin{array}{ccc} (0.918) & (1.024) \\ -0.557 & -0.097 \\ (0.850) & (0.550) \end{array}$	$\begin{array}{cccc} (0.918) & (1.024) & (0.828) \\ -0.557 & -0.097 & -1.892^{**} \\ (0.850) & (0.550) & (0.741) \\ \hline 999 & 999 & 999 \\ \end{array}$

Table S5. Logistic regression on evaluating a vaccine unacceptable

Note:

p < 0.1; p < 0.05; p < 0.01

	Depe	endent variable	e: Evaluating a v	accine unacce	ptable
	Pfizer	Moderna	Astrazeneca	Sputnik	Sinopharm
	(1)	(2)	(3)	(4)	(5)
Advice from Doctors	-0.058^{***}	-0.062***	-0.029	-0.027	-0.054^{**}
	(0.018)	(0.019)	(0.023)	(0.019)	(0.025)
Advice from Scientists	-0.034^{**}	-0.026^{*}	-0.054^{**}	-0.017	0.009
	(0.014)	(0.014)	(0.021)	(0.017)	(0.021)
Advice from Anti-Vaccine Propagators	0.202***	0.182***	0.117*	0.061	0.054
	(0.064)	(0.066)	(0.066)	(0.056)	(0.064)
Advice from Politicians	-0.030	-0.037^{*}	-0.065^{**}	-0.079^{***}	-0.088^{***}
	(0.022)	(0.022)	(0.030)	(0.015)	(0.030)
Advice from Family	-0.043^{***}	-0.037^{**}	0.029	-0.018	-0.057^{**}
	(0.017)	(0.019)	(0.032)	(0.026)	(0.028)
Advice from Friends	-0.006	-0.006	-0.014	-0.024	-0.011
	(0.020)	(0.024)	(0.039)	(0.030)	(0.034)
Advice from Journalists	-0.070	-0.030	0.024	-0.006	0.092
	(0.049)	(0.060)	(0.100)	(0.058)	(0.138)
Advice from Celebrities	0.071	0.114	0.046	0.124	0.109
	(0.091)	(0.102)	(0.122)	(0.100)	(0.145)
Age	-0.001^{**}	-0.001^{**}	-0.004^{***}	-0.0004	-0.0003
-	(0.0005)	(0.0005)	(0.001)	(0.001)	(0.001)
Female	0.006	0.0004	0.045**	-0.005	0.004
	(0.013)	(0.014)	(0.020)	(0.017)	(0.021)
University	-0.019	-0.023	-0.053**	-0.021	0.008
	(0.014)	(0.015)	(0.024)	(0.018)	(0.025)
City Category by Size	0.001	-0.005	-0.005	-0.003	0.011
	(0.005)	(0.006)	(0.009)	(0.008)	(0.010)
Wealth pre COVID-19	-0.004	-0.005	-0.003	-0.009^{*}	-0.011
-	(0.004)	(0.004)	(0.006)	(0.005)	(0.007)
Smoking	0.003	0.004	0.004	0.010	0.019**
-	(0.005)	(0.006)	(0.008)	(0.007)	(0.008)
Chronic Illness	-0.005	-0.005	0.008	-0.013	-0.0002
	(0.016)	(0.015)	(0.023)	(0.021)	(0.025)
COVID-19 Previously	0.008	0.003	-0.009	-0.032	-0.043
	(0.023)	(0.023)	(0.035)	(0.022)	(0.031)
Serious COVID-19 Previously	0.148	0.146	0.029	-0.036	0.030
	(0.110)	(0.110)	(0.094)	(0.027)	(0.100)
Constant	0.175***	0.199***	0.332***	0.178***	0.162**
	(0.051)	(0.053)	(0.068)	(0.060)	(0.073)
Observations	999	999	999	999	999
R^2	0.096	0.082	0.070	0.029	0.031
Adjusted R^2	0.081	0.066	0.054	0.013	0.014

Table S6. Linear probability	y models on evaluating a	a vaccine unaccept	table with individual	subjective wealth

Note:

*p<0.1; **p<0.05; ***p<0.01

	Depe	ndent variable	: Evaluating a v	accine unacce _l	ptable
	Pfizer	Moderna	Astrazeneca	Sputnik	Sinopharm
	(1)	(2)	(3)	(4)	(5)
Advice from Doctors	-0.052***	-0.052***	-0.018	-0.022	-0.038
	(0.018)	(0.018)	(0.023)	(0.020)	(0.024)
Advice from Scientists	-0.034**	-0.025^{*}	-0.042^{*}	-0.011	0.006
	(0.014)	(0.014)	(0.022)	(0.018)	(0.022)
Advice from Anti-Vaccine Propagators	0.175***	0.143**	0.121*	0.040	0.012
	(0.065)	(0.064)	(0.073)	(0.055)	(0.061)
Advice from Politicians	-0.026	-0.031	-0.060**	-0.076***	-0.081**
	(0.024)	(0.022)	(0.027)	(0.014)	(0.031)
Advice from Family	-0.041**	-0.033*	0.029	-0.014	-0.055**
2	(0.017)	(0.019)	(0.031)	(0.027)	(0.028)
Advice from Friends	-0.006	-0.005	-0.051	-0.023	-0.027
	(0.021)	(0.024)	(0.035)	(0.032)	(0.032)
Advice from Journalists	-0.068	-0.027	0.090	-0.008	0.119
	(0.050)	(0.059)	(0.089)	(0.058)	(0.162)
Advice from Celebrities	0.082	0.120	-0.102	0.131	0.122
	(0.104)	(0.103)	(0.064)	(0.102)	(0.158)
Age	-0.001**	-0.001^{*}	-0.003***	-0.0002	-0.0001
8-	(0.0005)	(0.0005)	(0.001)	(0.001)	(0.001)
Female	0.009	0.002	0.048**	-0.006	0.013
	(0.013)	(0.014)	(0.020)	(0.017)	(0.021)
University	-0.019	-0.022	-0.045^{*}	-0.017	0.004
	(0.015)	(0.015)	(0.025)	(0.019)	(0.025)
City Category by Size	0.001	-0.005	-0.001	-0.005	0.008
	(0.005)	(0.006)	(0.009)	(0.008)	(0.010)
Smoking	0.004	0.007	0.011	0.012*	0.019**
Smoning	(0.005)	(0.006)	(0.008)	(0.007)	(0.008)
Chronic Illness	-0.004	-0.004	-0.0003	-0.005	-0.003
	(0.016)	(0.015)	(0.023)	(0.021)	(0.026)
COVID-19 Previously	0.012	0.006	-0.029	-0.032	-0.040
	(0.024)	(0.023)	(0.033)	(0.024)	(0.033)
Serious COVID-19 Previously	0.175	0.151	0.079	-0.031	0.039
Serious COVID 19 Treviously	(0.120)	(0.111)	(0.101)	(0.028)	(0.101)
Constant	0.136***	0.143***	0.263***	0.103**	0.093*
Constant	(0.037)	(0.039)	(0.052)	(0.046)	(0.055)
Observations				· /	· /
Observations R ²	962	983	898	951	946
	0.082	0.064	0.073	0.022	0.023
Adjusted R ²	0.067	0.048	0.056	0.005	0.007

Table S7. Linear probility models on evaluating a vaccine unacceptable among those individuals who had stable vaccine assessment over the period

Note:

*p<0.1; **p<0.05; ***p<0.01

Supporting Information 5: Summary statistics for vaccine assessment matrices

Accepted Vaccine	Rated Vaccine	Observations	Mean	Standard Deviation
Astrazeneca	Pfizer	121	4.17	1.03
AstraZeneca	Moderna	121	3.94	1.00
AstraZeneca	AstraZeneca	121	3.84	1.00
AstraZeneca	Sputnik	121	3.56	1.12
AstraZeneca	Sinopharm	121	3.30	1.24
Moderna	Pfizer	35	4.32	1.06
Moderna	Moderna	35	4.28	0.92
Moderna	AstraZeneca	35	3.19	1.04
Moderna	Sputnik	35	3.59	1.18
Moderna	Sinopharm	35	3.18	1.36
Pfizer	Pfizer	202	4.58	0.70
Pfizer	Moderna	202	4.05	1.00
Pfizer	AstraZeneca	202	3.15	1.17
Pfizer	Sputnik	202	3.46	1.19
Pfizer	Sinopharm	202	3.07	1.27
Sinopharm	Pfizer	146	3.99	1.17
Sinopharm	Moderna	146	3.82	1.13
Sinopharm	AstraZeneca	146	3.36	1.23
Sinopharm	Sputnik	146	3.75	1.10
Sinopharm	Sinopharm	146	3.90	1.15
Sputnik	Pfizer	123	4.13	1.08
Sputnik	Moderna	123	3.94	1.07
Sputnik	AstraZeneca	123	3.14	1.09
Sputnik	Sputnik	123	4.02	1.05
Sputnik	Sinopharm	123	3.44	1.17

 Table S8. Summary statistics for Figure 4A: Accepted first vaccine

Accepted Vaccine	Rated Vaccine	Observations	Mean	Standard Deviation
AstraZeneca	Pfizer	11	4.30	1.25
AstraZeneca	Moderna	11	3.78	1.20
AstraZeneca	AstraZeneca	11	3.91	0.83
AstraZeneca	Sputnik	11	2.88	0.83
AstraZeneca	Sinopharm	11	2.12	0.83
Moderna	Pfizer	10	4.88	0.35
Moderna	Moderna	10	5.00	0.00
Moderna	AstraZeneca	10	3.12	0.99
Moderna	Sputnik	10	3.75	1.04
Moderna	Sinopharm	10	2.12	1.36
Pfizer	Pfizer	58	4.77	0.54
Pfizer	Moderna	58	4.36	0.86
Pfizer	AstraZeneca	58	2.70	1.35
Pfizer	Sputnik	58	2.92	1.25
Pfizer	Sinopharm	58	2.73	1.39
Sinopharm	Pfizer	16	3.78	1.20
Sinopharm	Moderna	16	3.00	1.00
Sinopharm	AstraZeneca	16	2.00	0.87
Sinopharm	Sputnik	16	3.22	0.83
Sinopharm	Sinopharm	16	4.18	0.87
Sputnik	Pfizer	13	3.90	1.10
Sputnik	Moderna	13	3.75	1.16
Sputnik	AstraZeneca	13	2.50	1.27
Sputnik	Sputnik	13	4.08	0.79
Sputnik	Sinopharm	13	3.50	1.18

Table S9. Summary statistics for Figure 4B: Rejected at least 1 vaccine; Grouped by accepted vaccine

Assigned Vaccine	Rated Vaccine	Observations	Mean	Standard Deviation
AstraZeneca	Pfizer	28	4.26	1.14
AstraZeneca	Moderna	28	3.95	1.12
AstraZeneca	AstraZeneca	28	2.20	1.08
AstraZeneca	Sputnik	28	3.42	1.10
AstraZeneca	Sinopharm	28	3.24	1.51
Moderna	Pfizer	2	4.50	0.71
Moderna	Moderna	2	2.00	0.00
Moderna	AstraZeneca	2	2.50	0.71
Moderna	Sputnik	2	2.50	0.71
Moderna	Sinopharm	2	4.00	0.00
Pfizer	Pfizer	6	3.25	1.71
Pfizer	Moderna	6	3.25	1.71
Pfizer	AstraZeneca	6	2.50	1.00
Pfizer	Sputnik	6	4.00	0.82
Pfizer	Sinopharm	6	4.00	1.22
Sinopharm	Pfizer	44	4.76	0.43
Sinopharm	Moderna	44	4.49	0.70
Sinopharm	AstraZeneca	44	3.24	1.35
Sinopharm	Sputnik	44	3.28	1.14
Sinopharm	Sinopharm	44	2.62	1.10
Sputnik	Pfizer	26	4.61	0.72
Sputnik	Moderna	26	4.06	1.00
Sputnik	AstraZeneca	26	2.78	1.31
Sputnik	Sputnik	26	2.47	1.12
Sputnik	Sinopharm	26	2.50	1.51

Table S10. Summary statistics for Figure 4C: Rejected at least 1 vaccine; Grouped by assigned vaccine

Supporting Information 6: Dynamics of vaccine assessment

In the nationally representative survey, individuals are asked whether they have changed their opinion about each of the available vaccine types over the past few months. In Table S11, we show the number of observations who declared a negative and positive change, respectively.

	Negative change	Positive change
Pfizer	19	18
Moderna	5	11
AstraZeneca	90	10
Sputnik	17	31
Sinopharm	34	19

Table S11. Number of respondents who changed their attitudes towards vaccines

In Table S12, we consider those individuals who accepted a vaccine that they did not rate the highest among the available vaccines. Based on the accepted vaccine type, we calculated the ratio of individuals who rate an other vaccine the highest. For instance, among the 9 observations who accepted Pfizer and did not rate Pfizer as their most preferred vaccine there were 2 individuals, who rate Moderna as the most preferred vaccine, and hence we obtained 0.22. As individuals were allowed to give the highest rating to multiple vaccines, we see that the sum of the "Best" ratings must be weakly greater than 1. Let us point out that Astrazeneca group has the most observations, who rate another vaccine the highest, and then Sputnik and Sinopharm follows. We can also see a strong preference toward Pfizer among those who did not accept their most preferred vaccine, which suggests the limited supply of Pfizer in the observed period in Hungary.

Table S12. Distribution of the most preferred vaccine types when the accepted vaccine is not the most preferred.

Accepted vaccine	Observations	Pfizer	Moderna	Astrazeneca	Sputnik	Sinopharm
Pfizer	9	-	0.22	0.11	0.78	0.44
Moderna	3	1.00	-	0.00	0.33	0.00
Astrazeneca	40	0.88	0.25	-	0.10	0.10
Sputnik	29	0.93	0.48	0.07	-	0.10
Sinopharm	29	0.79	0.41	0.17	0.17	-

The representative survey asks individuals to estimate the number of weeks that they would be willing to wait to get their most preferred vaccine. In Table \$13, we grouped individuals based on whether they rejected any vaccines and they rate the accepted vaccine the highest. We see that those who accepted a vaccine that they do not rate the highest estimated the expected number of weeks until the best vaccine to be 2.34 and 2.57 weeks on average if not rejected any and rejected at least 1 vaccine, respectively. We see that these estimations are relatively close to 2.76, which is the mean of the actual number weeks that the those individuals waited, who rejected a vaccine and ended up accepting a vaccine that they rate the highest.

Table S13. Expected and actual waiting time distribution based on rejecting any vaccines and rating the accepted vaccine the highest.

Rejected any	Highest rating	Observations	Mean expected	Mean actual	90% CI expected	90% CI actual
			(weeks)	(weeks)	(weeks)	(weeks)
0	0	101	2.34	-	(1.90,2.78)	(-,-)
0	1	526	0.29	-	(0.21,0.36)	(-,-)
1	0	9	2.57	0.83	(0.58, 4.56)	(0.11,1.56)
1	1	99	0.25	2.76	(0.11,0.39)	(2.20,3.32)

In Table S14, we grouped individuals, who accepted a vaccine that they do not rate the highest and individuals who were first offered and rejected that vaccine, and ended up getting a vaccine that they rate the highest. First, notice that the waiting times are not homogeneous among the vaccine types. For instance, the mean expected and the mean actual waiting times for Sinopharm are 3.46 and 3.78 weeks, respectively, while for Pfizer we have only 1.20 and 1.00. Let us also highlight the trend that AstraZeneca has a higher acceptance to rejection $(\frac{40}{25})$ ratio than the one that Sputnik or Sinopharm has $(\frac{29}{23} \text{ and } \frac{29}{41}, \text{respectively})$. This could be explained by that the AstraZeneca group estimated the expected waiting time to be longer than it actually was, while we see the opposite for Sputnik and Sinopharm.

Vaccine type	Observations	Observations	Mean expected	Mean actual	90% CI expected	90% CI actual
	Accept	Reject	Accept	Reject	Accept	Reject
			(week)	(week)	(week)	(week)
AstraZeneca	40	25	2.61	1.94	(1.87,3.36)	(1.47,2.42)
Moderna	3	1	0.00	2.00	(-,-)	(-,-)
Pfizer	9	5	1.20	1.00	(0.09,2.31)	(-0.35,2.35)
Sinopharm	29	41	3.46	3.78	(2.37,4.55)	(2.60,4.96)
Sputnik	29	23	1.61	2.58	(1.04,2.19)	(1.37,3.80)

Table S14. Expected and actual waiting times for accepting and rejecting a vaccine that is not the most preferred.

Supporting Information 7: Co-acceptance of vaccines among non-vaccinated

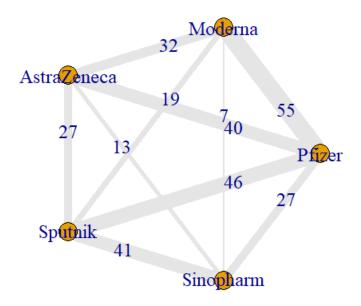


Figure S6. Co-acceptance of vaccines. Edge labels reflect to the number of individuals who accept both vaccines included. We only include those individuals to create this network who accept 2 or 3 vaccines.