Real-world data using mHealth apps in rhinitis, rhinosinusitis and their multimorbidities

Bernardo Sousa-Pinto1,2,3 | Aram Anto4 | Markus Berger5,6 | Stephanie Dramburg7 | Oliver Pfar8 | Ludger Klimek9,10 | Marek Jutel11 | Wienczyslawa Czarlewska12 | Anna Bedbrook4,13 | Arunas Valiulis14 | Ioana Agache15 | Rita Amaral1,2,3 | Ignacio J. Ansotegui16 | Katharina Bastl6 | Uwe Berger6 | Karl C. Bergmann13,17 | Sinthia Bosnic-Anticevich18 | Fulvio Braido19 | Luisa Brussino20 | Victoria Cardona21 | Thomas Casale22 | G. Walter Canonica23 | Lorenzo Cecchi24 | Denis Charpin25 | Tomás Chivato26 | Derek K. Chu27 | Cemal Cingi28 | Elisio M. Costa29 | Alvaro A. Cruz30 | Philippe Devillier31 | Stephen R. Durham32 | Motohiro Ebisawa33 | Alessandro Fiocchi34 | Wytske J. Fokkens35 | Bilun Gemicioğlu36 | Maia Gotua37 | Maria-Antonieta Guzmán38 | Tari Haahpta39 | Juan Carlos Ivancevich40 | Piotr Kuna41 | Igor Kaidashev42 | Musa Khaitov43,44 | Violeta Kvedariene45 | Désirée E. Larenas-Linnemann46 | Brian Lipworth47 | Daniel Laune48 | Paolo M. Matricardi7 | Mario Morais-Almeida49 | Joaquim Mullol50 | Robert Naclerio51 | Hugo Neffen52 | Kristoff Nekam53 | Marek Niedoszytko54 | Yoshitaka Okamoto55 | Nikolao G. Papadopoulos56 | Hae-Sim Park57 | Giovanni Passalacqua58 | Vincenzo Patella59 | Simone Pelosi60 | Nhán Pham-Thi61 | Ted A. Popov62 | Frederico S. Regateiro63,64 | Sietze Reitsma35 | Monica Rodriguez-Gonzales65 | Nelson Rosario66 | Philip W. Rouadi67,68 | Boleslaw Samolinski69 | Ana Sá-Sousa1,2,3 | Joaquin Sastre70 | Aziz Sheikh71 | Charlotte Suppli Ulrik72,73 | Luis Taborda-Barata74 | Ana Todo-Bom63 | Peter Valentin Tomazic75 | Sanna Toppila-Salmi39 | Salvatore Tripodi60,76 | Ioanna Tsiligianni77 | Erkka Valovirta78 | Maria Teresa Ventura79 | Antonio A. Valero80 | Rafael José Vieira1,2,3 | Dana Wallace81 | Susan Waserman82 | Sian Williams83 | Arzu Yorgancioglu84 | Luo Zhang85 | Mihaela Zidarn86,87 | Jaron Zuberbier88 | Heidi Olze88 | Josep M. Antó89,90,91,92 | Torsten Zuberbier13,17 | João A. Fonseca1,2,3 | Jean Bousquet13,17,93

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Scottish Centre for Respiratory Research, Cardiovascular & Diabetes Medicine, Medical Research Institute, Ninewells Hospital, University of Dundee, Dundee, UK

Kymed INNOV, Montpellier, France

Allergy Center, CUF Descobertas Hospital, Lisbon, Portugal

Rhinology Unit & Smell Clinic, ENT Department, Hospital Clinic, and Clinical & Experimental Respiratory Immunology, IDIBAPS, CIBERES, University of Barcelona, Barcelona, Spain

Johns Hopkins School of Medicine, Baltimore, Maryland, USA

Director of Center of Allergy, Immunology and Respiratory Diseases, Santa Fe, Argentina

Hospital of the Hospitaler Brothers in Buda, Budapest, Hungary

Department of Allergology, Medical University of Gdańsk, Gdańsk, Poland

Chiba University Hospital and Chiba Rosai Hospital, Chiba, Japan

Allergy Department, 2nd Pediatric Clinic, University of Athens, Athens, Greece

Department of Allergy and Clinical Immunology, Ajou University School of Medicine, Suwon, South Korea

Allergy and Respiratory Diseases, IRCCS Polyclinic Hospital San Martino, University of Genoa, Genova, Italy

Division of Allergy and Clinical Immunology, Department of Medicine, "Santa Maria della Speranza" Hospital, Battipaglia, and Agency of Health ASL, Salerno, Italy

TPS Production, Rome, Italy

Ecole Polytechnique Palaiseau, IRBA (Institut de Recherche Bio-Médicale des Armées), Bretigny, France

University Hospital 'Sv Ivan Rilski', Sofia, Bulgaria

Allergy and Clinical Immunology Unit, Centro Hospitalar e Universitário de Coimbra, Coimbra and Institute of Immunology, Faculty of Medicine, University of Coimbra, Coimbra, Portugal

Coimbra Institute for Clinical and Biomedical Research (ICBR), Faculty of Medicine, University of Coimbra, Coimbra, Portugal

Pediatric Allergy and Clinical Immunology, Hospital Español de Mexico, Mexico City, Mexico

Hospital de Clínicas, University of Parana, Umuarama, Brazil

Department of Otolaryngology-Head and Neck Surgery, Eye and Ear University Hospital, Beirut, Lebanon

Department of Otorhinolaryngology-Head and Neck Surgery, Dar Al Shifa Hospital, Salmiya, Kuwait

Department of Prevention of Environmental Hazards, Allergology and Immunology, Medical University of Warsaw, Warsaw, Poland

Fundacion Jimenez Diaz, CIBERES, Faculty of Medicine, Autonoma University of Madrid, Madrid, Spain

Usher Institute, The University of Edinburgh, Edinburgh, UK

Department of Respiratory Medicine, Copenhagen University Hospital-Hvidovre, Copenhagen, Denmark

Institute of Clinical Medicine, University of Copenhagen, Copenhagen, Denmark

Department of Immunology and Clinical Immunology, Cova da Beira University Hospital Centre, and UBIAir - Clinical & Experimental Lung Centre and CICS-UBI Health Sciences Research Centre, University of Beira Interior, Covilhã, Portugal

Department of General ORL, H&NS, Medical University of Graz, ENT-University Hospital Graz, Graz, Austria

Allergy Unit, Policlinico Casilino, Rome, Italy

Health Planning Unit, Department of Social Medicine, Faculty of Medicine, University of Crete, Greece and International Primary Care Respiratory Group IPCRG, Aberdeen, Scotland

Department of Lung Diseases and Clinical Immunology, University of Turku and Terveystalo Allergy Clinic, Turku, Finland

Unit of Geriatric Immunology and Inflammation, University of Bari Medical School, Bari, Italy

Pneumology and Allergy Department CIBERES and Clinical & Experimental Respiratory Immunology, IDIBAPS, University of Barcelona, Barcelona, Spain

Nova Southeastern University, Fort Lauderdale, Florida, USA

Department of Medicine, Clinical Immunology and Allergy, McMaster University, Hamilton, Ontario, Canada

International Primary Care Respiratory Group IPCRG, Larbert, Scotland

Department of Pulmonary Diseases, Celal Bayar University, Faculty of Medicine, Manisa, Turkey

Department of Otolaryngology Head and Neck Surgery, Beijing TongRen Hospital and Beijing Institute of Otolaryngology, Beijing, China

University Clinic of Respiratory and Allergic Diseases, Golnik, Slovenia

University of Ljubljana, Faculty of Medicine, Ljubljana, Slovenia

Department of Otorhinolaryngology, Charité-Universitätsmedizin Berlin, Berlin, Germany

ISGlobal, Barcelona Institute for Global Health, Barcelona, Spain

IM (Hospital del Mar Medical Research Institute), Barcelona, Spain

Universitat Pompeu Fabra (UPF), Barcelona, Spain
1 INTRODUCTION

The burden and cost of allergic and chronic respiratory diseases are increasing worldwide, with most economies struggling to effectively respond.\(^1\) Transforming healthcare systems requires strengthened integrated care using organisational health literacy. For this, digital health may be particularly useful, as it may put the patient at the centre of his/her disease management, promote better monitoring and improve patient education. This is particularly true for non-communicable diseases, whose burden is expected to increase in the near future. It is therefore essential to know of the available digital health tools for each disease and how can they be further explored to improve their management.

Digital health is an umbrella term which encompasses eHealth and benefits from areas such as advanced computer sciences (e.g., 'big data' and artificial intelligence), eHealth, as defined by the World Health Organization (WHO),\(^5\) comprises several components including electronic health records, telehealth and mobile health (mHealth). The latter has been defined as a 'medical and public health practice supported by mobile devices, such as mobile phones'.\(^7\) It includes: (i) equipment/connected medical devices, (ii) mHealth services and (iii) mHealth apps.\(^7,8\)

Apps designed for and used in allergic rhinitis (AR) and chronic rhinosinusitis (CRS) may help to better understand these diseases and their management as well as to identify and address some unmet needs. This is particularly important in these chronic diseases which are often trivialised\(^9\) and undertreated,\(^10\) both by patients and healthcare providers. However, these new tools first need to be tested for privacy rules, acceptability, usability and cost-effectiveness. In addition, they should be evaluated for their impact on (i) the digital transformation of health, (ii) healthcare delivery and (iii) health outcomes. Given the potential of mHealth tools to enable the digital transformation of health and care, empowering citizens and building a healthier society,\(^12\) it is of great importance to review apps whose data collection tools (e.g., questionnaires) have been validated for the case study chronic conditions of allergic rhinitis (AR) and CRS.

In the present paper, all apps relevant to AR and CRS management retrieved using a market research tool based on an automatic search process will be presented. However, only apps with peer-reviewed published data for a given disease will be reviewed. The application of these tools/apps will be discussed regarding their potential for identifying disease phenotypes based on real-life direct patient-centred data, diagnosis, management and adherence to treatment, as well as for promoting the digital transformation of health and care.

**Correspondence**
Jean Bousquet, Institute of Allergology, Charité - Universitätsmedizin Berlin, Corporate Member of Freie Universität Berlin and Humboldt-Universität zu Berlin, Berlin, Germany. Email: jean.bousquet@orange.fr

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2 MARKET RESEARCH FOR MHEALTH APPS IN RHINOLOGY

2.1 Identification of mHealth apps

An important challenge for app review studies concerns the lack of automatic standardised search strategies, rendering the identification of potentially relevant apps a time-consuming manual task. Such limitations could be overcome by the development of automatic methods for app screening. Recently, such methods have been described for breast cancer, AR, urticaria and anaphylaxis apps. Automatic methods for app screening also have the advantage of running screening processes more frequently than manual approaches and at an increased speed, and of potentially identifying relevant apps whose name and icon are not obvious.

The method used for the identification of relevant mHealth apps in rhinitis has been described elsewhere. In this review, we will focus on (i) the four apps identified by that study as having associated scientific publications for AR as (ii) two additional apps for which scientific publications were subsequently identified. In brief, an app screening programme capable of performing searches in app stores without any human intervention has been developed for searching for AR apps using JavaScript, a commonly used programming language that allows searches of dynamic content on web pages. The screening programme builds upon two open-source packages.

On the other hand, relevant apps in CRS had not been previously identified. In this study, we used the aforementioned app screening programme to scrape Apple App and Google Play stores for searching for CRS apps, according to the following criteria: (i) search terms: rhinitis, hay fever, rhinosinusitis; sinusitis; (ii) countries: United Kingdom, United States, Germany (since we wanted to have two English-speaking countries plus another one with a different language); and (iii) languages: English, German. The number of results retrieved at each iteration was limited to a maximum of 200 apps for Apple/iOS or Android. After retrieving the search results from all iterations, the programme compared the names of all of the retrieved apps and discarded duplicates. A PubMed search up to November 2021 on the names of the retrieved relevant apps was then carried out to identify published peer-reviewed papers on all such apps.

2.2 Allergic rhinitis and CRS

Using automatic and manual search methods, we identified six relevant apps for AR for which data have been published in the literature: AllergyMonitor, AirRater, MASK-air; Pollen App (patient’s hay fever diary, developed in Austria); Husteblime (a mobile phone health app developed in Germany as a spin-off of Pollen App including the patient hay fever diary) and AllerSearch (Table 1). Galenus Health did not have any published data on AR and could therefore not be considered in this section for AR, but will be discussed in the subsection of CRS apps.

| TABLE 1 | Apps relevant for allergic rhinitis and rhinosinusitis management and with published data on rhinitis |
|-----------------|-----------------------------------------------|
| Countries | Allergy monitor | AirRater | Husteblime | AllerSearch | Pollen | MASK-air | Galenus health |
| Last update | May 2021 | Jan 2021 | Mar 2021 | May 2021 | Dec 2020 | Feb 2021 | Jan 2021 |
| Published methodological or clinical validation | YES | YES | YES | YES | YES | YES | YES |
| List of all medications | YES, by medication and dosage, customised by country | YES, by medication and dosage, customised by country | YES, by medication and dosage, customised by country | YES, by medication and dosage, customised by country | YES, by medication and dosage, customised by country | YES, by medication and dosage, customised by country | YES, by medication and dosage, customised by country |
| Published general data protection regulation (GDPR) | YES | YES | YES | YES | YES | YES | YES |
| | | | | | | | |
Among the 17 apps retrieved for CRS, only one had published data for this disease, namely Galenus Health (with a single identified paper) (Table 2). However, in this paper, the app is labelled as MySinusitisCoach, developed using the MASH-air™ structure. A new app designed by members of the Mayo Clinic has been developed, but solely tested in a pilot study assessing 10 participants. Kagen Air was only presented in a review paper and was therefore excluded.  

2.3 Limitations of the followed approaches

We used a small set of search terms, which may correspond to a restrictive approach. Nevertheless, these terms were chosen by the Allergic Rhinitis and its Impact on Asthma (ARIA) expert group. Moreover, in performing a PubMed search for ‘apps’, ‘mHealth’, ‘eHealth’ AND ‘rhinitis’ or ‘rhinosinusitis’, we did not find any other apps other than the ones included in this review. Only two languages were searched and some apps may exist in other countries (e.g., in Polish, the Apsik and Dzienniki Alergiaka apps are available). However, only one app has been described in articles available in PubMed. An additional limitation is that some apps may not have used the name they are currently using. This is, for example, the case for Galenus Health which was labelled MySinusitisCoach. Finally, we did not search any other medical literature databases and may have missed some apps.

### TABLE 2  Apps retrieved by an automatic search for chronic rhinosinusitis

| App name                                      | Availability |
|-----------------------------------------------|--------------|
| 1. Sinus Infection Symptoms                   | Google       |
| 2. Headache Tracker - Migraine & Headache Log | Google       |
| 3. Ada – check your health                    | Google       |
| 4. Symptomate – Symptom checker               | Google       |
| 5. Migraine and headache diary                | Google       |
| 6. Feeble: symptom and illness tracker        | Google       |
| 7. Headache Log                               | Google       |
| 8. MeMD – Doctor’s Visits Online!             | Google       |
| 9. Allergy, Asthma & Sinus Center             | Apple        |
| 10. SinusMonitor                              | Apple        |
| 11. KagenAir                                  | Apple        |
| 12. Correlate: Health Journal                 | Apple        |
| 13. MigraineMind Migraine Diary               | Apple        |
| 14. iGeoPolen Portugal                        | Apple        |
| 15. PROMinENT - Reporting System              | Apple        |
| 16. Galenus Health                            | Google, Apple|
| 17. Docquity- The Doctors’ Network            | Google       |

3 VALIDATION AND MAIN CHARACTERISTICS OF mHEALTH APPS IN RHINOLOGY

3.1 Allergic rhinitis

3.1.1 AllergyMonitor®

AllergyMonitor® (TPS software production, Rome, Italy) is an online service that was developed in 2009 with the aim of (i) enabling the recording of clinical symptoms, drug use and adherence to allergen immunotherapy (AIT) and (ii) monitoring the efficacy of sublingual or subcutaneous AIT by patients with allergic rhino-conjunctivitis and/or asthma. The system, available to everyone and simple to use, consists of two parts: a patient app (front end) and a website for the attending doctor (back-office). The results can be accessed by the patient and attending physician—as concise reports via a smartphone or computer—in a collaborative setting of blended care. Geolocation is optional.

The download and usage of this app are free of charge and there are no advertisements. It falls under Italian jurisdiction, is CE1 registered and follows the General Data Protection Regulation (GDPR). The Technology Readiness Level (TRL) has been assessed for this app. It is available in 14 countries (TRL9) and in several languages. It contains the Control of Allergic Rhinitis and Asthma Test (CARAT) as well as pollen counts (TRL9).

The quality of the AllergyMonitor® data was checked by estimating the percentage of changes in trends of the trajectories produced by the patients’ data.

Methods for statistical analysis

To account for noise when identifying clusters of homogenous patients, the authors applied the fuzzy k-medoids algorithm to the obtained functional coefficients. By the B-spline basis system, these coefficients allowed continuous smoothing functions to be found, synthesising the general trend of the observed data.

3.1.2 MASK-air®

MASK, the Phase 3 ARIA (Allergic Rhinitis and its Impact on Asthma) initiative, is a Good Practice of DG Santé for digitally-enabled, patient-centred care. It aims to improve the management of AR and asthma multimorbidity in a patient-centred approach and to facilitate shared decision-making. It includes (i) a freely available app (MASK-air®; formerly the Allergy Diary, Android and iOS, operational in 28 countries and 20 languages, (ii) an interoperable electronic decision support system for the support of healthcare professionals in shared decision making, (iii) a web-based interoperable questionnaire for physicians, (iv) the CARAT questionnaire for screening allergic diseases and assessing their control, the European Quality of Life 5 Dimensions (EQ-SD) and the Work Productivity and Activity questionnaire for asthma and (v) a sentinel network for air quality and pollen
seasons. The TRL has been assessed for the MASK-air® app by MASK-air® members (TRL9). Mask-air® is CE1 registered and follows the GDPR. It is in the process of being registered as a Medical Device Regulation Class 2. It operates under French jurisdiction. Geolocation is optional.

Published methodologic assessment of Mask-air®

Following the COncensus-based Standards for the selection of health Measurement INstruments guidelines, internal consistency (Cronbach’s α-coefficient and test-retest), reliability (intraclass correlation coefficients), sensitivity and acceptability of the Mask-air® visual analogue scales (VASs) were shown for global allergy symptoms, nose, eye and asthma. A subsequent study also concluded that VASs display high intra-rater reliability, high test-retest reliability, moderate/large responsiveness and moderate/high concurrent validity. The quality of Mask-air® data was checked by estimating the intra-individual response variability index, a flexible way of detecting insufficient effort responding. The independency of VAS questions (from each other) was confirmed using the Bland and Altman regression analysis. Minimal Important Difference (MID) has been provided for most patient-reported outcome measures (PROMs).

Most Mask-air® analyses were observational and non-interventional. However, there was a clinical trial and a quasi-experimental study.

Acceptability of Mask-air® by patients

Many patients do not understand the needs and benefits of mHealth and may worry about data privacy. A minority may have difficulties in using a mobile phone. On the other hand, many patients over-rely on internet-based information and untested mHealth solutions. Two qualitative studies enabled a better understanding of the patients’ needs and expectations, which permitted an accurate modification of the app. A study in Puglia (Italy) showed that older adults with a low level of education were able to use the Mask-air® app after a short training session.

Overall, Mask-air® aims (i) to strengthen the EU digital single market, (ii) to develop the implementation of digitally-enabled real-life care pathways at the global level with the Global Alliance against Chronic Respiratory Diseases (GARD) and WHO and (iii) to develop a change management strategy in allergic and airway diseases.

3.1.3 | Pollen App: Patient’s hay fever diary

A model of individualised prediction of AR symptoms, named the Patient’s Hay fever Diary (PHD), has been developed in Austria. There was a precise validation of aerobiologic data and of some clinical data in pollen allergic individuals. There were also numerous publications using the symptom data retrieved by this app. The Patient’s Hay fever Diary was first available as a website in 2009 and was later included in Pollen App as well as in Hustebume, a spin-off app. The system allows the documentation of allergy symptoms and the use of medication. It offers a simple comparison of personal symptoms with the regional pollen load for every user. Pollen App is available in eight countries/regions and six languages. The download and usage of this app are free of charge and there are no advertisements. Both systems adhere to the GDPR (Directive 95/46/EC) and collect only a minimum amount of personal data.

Scientific studies using PHD/Pollen App symptom data first under a filtering process to assure the inclusion of exclusively qualitative data (e.g., for more seasons, with a certain number of entries per user) leading to robust results. The datasets have been analysed using not only statistical methods but also computational intelligence methods like Self Organising Maps.

Pollen App consists of three main parts: information, symptom documentation and medical assistance. Information is given concerning the pollen load including a personal allergy risk, the daily pollen load for various aeroallergens, forecast maps based on different models, as well as a dictionary with information on the most important allergic plants. Symptom documentation is made in the pollen diary (Patient’s Hay fever diary) and adapts the forecasts automatically if used (personal pollen information and allergy risk). Medical assistance concerns information on doctors in the vicinity, therapy recommendation situations for no-, low-, medium- and high-risk burden as well as a symptom report (available since 2021) that can be shared with the patient’s doctor.

Forecast data were part of the scientific research besides the exploitation of the symptom data. Nine freely available apps delivering pollen information and pollen forecasts had been tested with a focus on their prediction of the pollen load in the 2016 grass pollen season (Table 3). For six apps, the rates of correct pollen forecasts were around 50%, with Pollen App displaying the highest. Only two apps provided sufficiently accurate forecasts for the “readiness to flower” for grasses.

| App       | Assessed city | Exact rates of correct pollen forecasts | Forecast on the readiness to flower |
|-----------|---------------|----------------------------------------|-------------------------------------|
| Pollen    | Vienna        | 62.9                                    | Yes                                 |
| Biowetter | Vienna        | 31.8                                    | No                                  |
| Pollenwarner | Vienna    | 34.0                                    | No                                  |
| DWD       | Berlin        | 41.1                                    | No                                  |
| Allergiehelfer | Berlin    | 48.6                                    | No                                  |
| Pollenflug | Berlin        | 50.7                                    | No                                  |
| Allergohelp | Berlin       | 45.2                                    | No                                  |
| Pollen news | Basel        | 42.4                                    | Yes                                 |
| Hayfever  | London        | 35.7                                    | No                                  |
A detailed description of the studies assessing PHD data is available in Section 4.1 of this review.

3.1.4 | Husteblume mobile phone health app

Husteblume is a mobile phone health app, developed in Germany, with the aim of facilitating the self-management of pollen-related AR. It includes providing information on the drugs most frequently used by patients with similar profiles. This app is a spin-off of Pollen App. It used Patient’s Hay fever Diary for symptom documentation and the forecasts of Pollen App. A study assessed the usability, changes in quality of life, health literacy and self-efficacy for managing one’s chronic disease. A total of 661 app users were included and 143 were evaluated after the pollen season. The patients using the app for a longer period perceived many subjective improvements, including better information about their allergy, improved quality-of-life and improved coping with their allergy.

3.1.5 | AirRater app

AirRater provides environmental data for patients with AR and allows them to record their symptoms and medication use. This app is available in English and is particularly tailored for Australia. A study assessed AirRater users by means of semi-structured interviews, with most of them indicating that information provided by the app helped them to make decisions and implement behaviours to protect their health.

3.1.6 | AllerSearch

AllerSearch comprises a hay fever daily questionnaire, an assessment of the impact of rhinitis symptoms on work productivity and information on pollen levels. In addition, it is set to implement an artificial intelligence system to determine the degree of rhinitis severity based on photos from the eyes of the patients. A crowd-sourced study using the smartphone app AllerSearch was carried out. In 11,248 subjects (of whom 9041 had AR), demographic factors and symptoms associated with AR were assessed. In addition, using AllerSearch, clusters of patients with allergic rhinitis were obtained based on their presented symptoms.

3.2 | Rhinosinusitis

There is only one mobile app within the defined criteria for CRS, namely mySinusitisCoach, an app for CRS patients. A study reported the cross-sectional evaluation of the data of 626 users of this app. Patient characteristics were analysed as well as the level of disease control based on the VAS global CRS symptom score and on adapted European Position Paper on Rhinosinusitis and Nasal Polyps (EPOS) criteria.

4 | ADDED VALUE OF mHEALTH APPS IN ALLERGY PHENOTYPING

Apps can be useful for generating hypotheses, which frequently need to be confirmed by more ‘classical’ clinical studies. An example of digital health in phenotype discovery was proposed by MASK-air. Multimorbidity in allergic airway diseases was well known, but no data existed regarding the daily dynamics of symptoms. Using the MASK-air app, eight hypothesis-driven patterns were defined based on “Low” and “High” VAS levels. Days with rhinitis alone had the lowest VAS global allergy symptoms. A novel and previously unrecognised extreme pattern of uncontrolled multimorbidity was identified in 2.9% of the days: Rhinitis High - Asthma High - Conjunctivitis High. This hypothesis-generating study was confirmed by classical epidemiologic studies, showing that it is important to consider ocular symptoms in severe asthma and that the severity of individual allergic diseases increases with the number of allergic morbidities. These findings were reinforced using computational analyses suggesting that there are common pathways in multimorbidity. These were confirmed by a genomic approach in the MedALL study (Mechanisms of the Development of Allergy, FP7) and showed a novel whole blood gene expression signature for asthma, dermatitis and rhinitis multimorbidity in children and adolescents.

5 | OPTIMISATION OF THE IDENTIFICATION OF THE POLLEN SEASON AND ROLE OF AIR POLLUTION

The definition of a pollen season determines the start and the end of the time period with a certain amount of pollen in the ambient air. Although common definitions should be used, different pollen season definitions were used for a long time, based on different terms and methods. Recently, suggested pollen season definitions for clinical trials were tested using apps and were applied for the first time to more aeroallergens. Clinical trials with pollen allergic patients need validated, high-quality pollen data and forecasts to yield comparability and adhere to scientific standards.

5.1 | Pollen App (Patient’s hay fever diary)

Using Pollen App, the representativeness of pollen concentrations was assessed for 20 pollen types in 2015–2016 in Vienna for rooftop and ground level. Comparisons were then performed with weather and symptom data. Computational intelligence methods were used to describe similarities and interdependencies, while the random
forest algorithm was used to model symptom data. Most of the examined taxa showed similar patterns (e.g., Betula), while some showed differences at different heights (e.g., the Poaceae family). Some findings contradicted the literature and led to the posing of new hypotheses (e.g., concerning the abundance of Urticariaceae pollen in rooftop and ground levels). Temperature and humidity influenced daily pollen concentrations for most of the taxa. The rooftop trap was adequate when compared with the symptoms, justifying the recommendations concerning the location of a pollen trap and showing the importance of validations using symptom data.

The connection between pollen concentrations and crowd-sourced symptom data provided new insights from daily and seasonal symptom load index data from 2013 to 2017 in Vienna. The Daily Symptom-Load Index (SLI) and pollen concentration data were correlated. This study showed a linear relationship between SLI and pollen concentrations/seasonal pollen index daily but not on a seasonal basis. Cross-reactivity to other pollen types, allergen content and air pollution could play a considerable role.

Quantification of the burden of pollen allergy was performed in Austria and Germany over 10 years using electronically-generated symptom data from PHD. Four different symptom score calculation methods were applied to the datasets. This study did not detect significant differences between the various methods of symptom score calculation. Nasal symptoms determined about 40% of the scores.

Grass pollen-triggered allergic symptoms vary within the season. Symptoms were studied in Vienna (Austria) during the 2014, 2015 and 2016 grass pollen seasons. They were compared with the grass pollen season defined either by grass pollen level data or phenology (grass species determination in the field). The symptom peak of most users was observed in the second section of the grass pollen season (70%), followed by the first section (20%) and the third section (10%). Differences between grass species were found.

5.2 | Disentangling polysensitisation: The @IT.2020 study

An adequate definition of pollen seasons (e.g., regarding their beginning and end) is essential for optimal identification and management in AR patients. A position paper by the European Academy of Allergy and Clinical Immunology (EAACI) proposed pollen season definitions for Northern and Central Europe. In the @IT.2020 multi-centre study, pollen counts for many species were collected over 1 year (2018) in six Mediterranean cities (of four different countries) for seven pollen taxa (Poaceae, Oleaceae, Fabales, Cupressaceae, Parietaria, Ambrosia and Artemisia). The @IT.2020 study showed heterogeneous results between locations in terms of pollen species and duration of pollen seasons. A fragmentation of pollen seasons was found, with high pollen counts being separated by periods of low pollen counts.

In the Mediterranean area, patients with pollen-induced AR are often polysensitised, rendering their assessment complex for aerobiologists and physicians. AllergyMonitor® was used to improve the precision of diagnosing pollen allergy using daily symptom monitoring and graphical representations of airborne pollen data. Unfortunately, diagrams illustrating daily pollen concentrations from many sources in parallel make the interpretation of each of these curves difficult. This problem may be solved by using curves based on the cumulative transformation of pollen data using AllergyMonitor®.

5.3 | Impact of air pollution on rhinitis

Several studies have suggested an interaction between air pollution and pollen exposure, with an impact on allergy symptoms. However, large studies with real-life data have not been available until recently.

In the POLLAR study, associations between ozone and particulate matter with a diameter of <2.5 μm (PM$_{2.5}$) and AR control assessed using MASK-air® were studied during grass and birch pollen seasons as well as outside the pollen season. Pollutant levels were assessed using the System for Integrated modeling of Atmospheric Composition (SILAM) database. Associations between ozone and uncontrolled rhinitis were found to be stronger during the grass pollen season than during the birch pollen season, possibly related to the relationship between ozone and higher temperatures.

Associations between six pollen types and respiratory symptoms were studied using the AirRater smartphone app in Tasmania (Australia) from 2015 to 2019. Associations between daily respiratory symptoms and pollen concentrations were first studied using Poisson regression models, with the case time-series approach designed for app-sourced data. Potentially non-linear and lagged associations were examined with total pollen and six pollen taxa, with adjustments for seasonality and meteorology, and testing for interactions with particulate air pollution (PM$_{2.5}$). Non-linear associations were found between total pollen or individual pollen taxa and respiratory symptoms.

Using Pollen App, associations between symptoms, grass, birch or ragweed pollen levels, air quality and meteorological data (temperature, relative humidity) were studied for the metropolis of Vienna. Only ozone was significantly associated with symptom scores in birch, grass and ragweed pollen seasons. Further analyses in a model with meteorological data showed that the effect estimates of ozone were attenuated, but remained significant for the grass pollen season.

6 | ADVANCES IN DIAGNOSIS

6.1 | Aetiological diagnosis of seasonal allergic rhinitis

The analysis of the AR symptom severity scores during pollen exposure can be used to evaluate the clinical relevance of a patient’s sensitisation to specific pollen. The comparison of symptom severity scores (Rhinitis Total Symptom Score, RTSS, in AllergyMonitor®) or of symptom scores (Pollen App) with pollen concentration data may guide the physician in the choice of the correct AIT composition.
6.2 | Disease severity scores

6.2.1 | Comparison between symptom scores by AllergyMonitor®

Using AllergyMonitor®, 105 children with pollen allergy monitored their daily symptoms for 2 months during the grass pollen season. Six AR severity scores were compared with pollen counts at both population and individual levels: (i) the RTSS, (ii) the Adjusted RTSS calculated based on the last observation, (iii) the Adjusted RTSS calculated based on the worst observation, (iv) the Rhinoconjunctivitis Allergy-Control-SCORE, (v) the Average Combined Score and (vi) the average Adjusted Symptom Score. These disease severity scores tended to provide similar results at population level but often produced heterogeneous slopes in individual patients.

6.2.2 | Finding an optimal combined symptom-medication score

There was an urgent need for a validated combined symptom-medication score (CSMS) in AR, both for clinical practice and clinical trials. The CSMS needed to be developed against a gold-standard tool/measurement that does not simply measure allergy symptoms or use of allergy medications and that assesses, if possible, a variable related to the economic impact of AR. Such tools include, among other endpoints, work productivity and quality-of-life for AR. Only MASK-air® and AllergyMonitor® currently have these capabilities and can be used. However, the latter was missing EQ-5D and Work Productivity and Activity Impairment (WPAI-AS) and therefore MASK-air® was selected. Pollen App and PHD ask for quality of life, but data concerning this aspect have not yet been published.

The results showed that (i) A hypothesis-driven score based on MASK-air® data was highly correlated with all instruments of quality-of-life and work tested, and had high concurrent validity and test-retest reliability; (ii) Several data-driven scores, in particular those based on cluster analyses, had a slightly higher level of correlation with identified endpoints; (iii) These results have been found to be highly reproducible across all tested regions (nine countries).

6.3 | Indication of allergen immunotherapy

The efficacy of AIT depends on the precise identification of the triggering allergen. However, diagnostics based on retrospective clinical history and sensitisation to whole extracts often lead to equivocal results. A study assessed a recently established algorithm for a clinical decision support system (@IT2020-CDSS) for pollen rhinitis and its diagnostic steps (anamnesis, skin prick test or serum-specific IgE, component-resolved diagnosis and real-time digital symptom recording by the AllergyMonitor® eDiary) on doctors’ AIT prescription decisions. After educational training on the @IT2020-CDSS algorithm, 46 doctors (18 allergy specialists and 28 general practitioners) proposed a hypothetical AIT prescription for 10 clinical index cases. Decisions were recorded repeatedly, based on different steps of the algorithm. The combined use of the component-resolved diagnosis and of the AllergyMonitor® eDiary increased the hypothetical AIT prescriptions in both groups. Physicians considered the algorithm useful for the optimisation of classical diagnostic work-up.

7 | EVOLUTION IN MANAGEMENT

7.1 | Medications

mHealth can be used to generate innovative insights into optimising treatment for the improvement of AR control. Two MASK-air® cross-sectional real-world observational studies were undertaken in 22 countries to complement a pilot study and provide novel information on medication use, disease control and work productivity in the everyday life of patients with AR. The four most common intranasal medications containing intranasal corticosteroid (INCS) (including INCS + intranasal antihistamine) and eight oral H1-antihistamines (OAH) were studied. A total of 912 users filled in 112,054 days of VASs up to 2017. The control of days with rhinitis using VAS was (i) similar for ‘no treatment’ and monotherapy with INCS and Azelastine-Fluticasone intranasal formulation (best control), (ii) worse for monotherapy with OAH and (iii) the worst for multiple treatments (co-medication). These observational studies using a very simple daily assessment tool (VAS) on a mobile phone answered questions previously thought infeasible.

7.2 | Allergen immunotherapy

Real-world data are available for AIT in the MASK-air® database. A proof-of-concept study has compared days of participants with AIT versus days of participants without AIT on VAS global allergy symptoms and VAS work. A total of 317,176 days were analysed, of which 11.4% involved AIT users. Lower median VAS global allergy symptoms and VAS work levels were observed for participants under AIT and were compared to the levels on days without treatment, with monotherapy or with polytherapy. This enabled us to better understand the role of AIT in real life (Figure 1). Nevertheless, further studies are required.

8 | ADHERENCE TO TREATMENT IN ALLERGIC RHINITIS

8.1 | Understanding adherence

mHealth may help to better understand adherence to treatment. Following a pilot study in less than 3000 AR patients, an observational cross-sectional study was carried out on all MASK-air® users. The modified Medication Possession Ratio (MPR) and the
Proportion of Days Covered (PDC) approach were used to assess secondary adherence. In total, 6949 users reported at least one VAS data recording, among whom 1195 were included in the analysis of adherence. Of the users, 11.3% were adherent to medication (MPR ≥70% and PDC ≤1.25), 4.2% were partly adherent (MPR ≥70% and PDC = 1.50) and 176 (14.6%) were switchers. On the other hand, 69.1% of the users were non-adherent to medications (MPR < 70%), indicating that adherence to AR treatment is low. This study proposed an approach for measuring retrospective adherence based on an app, representing a novel approach for analysing the behaviour of medication-taking in a real-world setting.

8.2 | Improving adherence

mHealth may improve adherence to treatment in chronic diseases. Children and adolescents (5–18 years) with moderate-to-severe seasonal AR to grass pollen, requiring a daily INCS administration, were recruited in April 2013. Participants were randomised to AllergyMonitor® or to usual care (no diary) and followed up until 15 June 2013. Intra-nasal mometasone use, expressed as both optimal adherence rate and average daily use, was higher in the AllergyMonitor® group than in usual care. Disease knowledge improved among the patients using AllergyMonitor® but not among the controls. However, no differences were observed at baseline and at follow-up visits in the reported severity of disease, nasal flow and quality of life. This was due to an unexpected low temperature and pollen exposure during the observation period.

In another study on AllergyMonitor® in 67 patients, the adherence to daily symptom monitoring remained high (>80%) throughout several weeks when prescribed and thoroughly explained by the treating doctor. Furthermore, app use was associated with improved adherence to symptomatic drugs and AIT.

9 | ASSESSMENT OF THE ECONOMIC BURDEN OF AR AND COST-EFFECTIVENESS OF MANAGEMENT STRATEGIES

Allergic rhinitis is a burdensome condition, with an important impact on work and school productivity.

MASK-air® can be used to quantify this impact, as it includes questions assessing the daily impact of AR symptoms on work productivity (VAS Work) and on school performance. In addition, the WPAI-AS – which quantifies the impact of allergy on work and activities – can be answered optionally in MASK-air®. MASK-air® enables the estimation not only of indirect costs resulting from loss of work productivity, but also of direct costs resulting from AR medication and AIT use. A monthly question asking the user whether he/she had an outpatient visit related to AR during the previous month could help to further improve the estimation of direct costs related to AR.

MASK-air® also includes EQ-5D, whose scores can be converted into utilities (standardised measures of preferences that patients have for health status) for many of the countries where MASK-air® is available. Such a feature may be particularly useful for performing cost-utility analyses, in which interventions are compared regarding their costs and also their effectiveness adjusted for patients’ preferences.

10 | NEXT-GENERATION CARE PATHWAYS FOR THE DIGITAL TRANSFORMATION OF HEALTH CENTRED AROUND THE PATIENT

As an example of chronic disease care, MASK, in collaboration with professional and patient organisations in the field of allergy and airway diseases, proposes real-life care pathways centred around the patient with AR and/or asthma multimorbidity. It uses mHealth to monitor environmental exposure and to sustain Planetary Health. Next-generation guidelines have been proposed to assess the recommendations of Grading of Recommendations, Assessment, Development and Evaluation guidelines in AR and asthma using real-world evidence and real-world data obtained through mobile technology. Moreover, mHealth should be considered in the wider frame of Planetary Health.

11 | CONCLUSIONS

This review has provided several examples of how mHealth apps can play a key role in the scientific investigation and clinical assessment of AR and CRS. While mHealth apps may be a useful complementary tool in the diagnosis and management of patients with AR or CRS, some gaps still merit attention and should be the focus of future studies. Few apps included multimorbidity.

An important limitation of mHealth apps consists of the fact that only a minority of patients use them regularly. Identifying the
patients who most probably use them more regularly would maximise their effectiveness. Such an identification could stem from a simple baseline questionnaire, whose development and validation should be the target of future studies.

On the other hand, connecting daily apps to medications (e.g., inhalers) or diagnostic tests (e.g., spirometry for asthma) may open the possibility of a more personalised monitoring of patients with AR or CRS. Furthermore, personalisation increases motivation to continuously use mHealth and eHealth apps and may improve adherence.

Other issues meriting discussion concern their certification, reimbursement, interoperability and quality control. Only a small fraction of available apps have published scientific results, with the content veracity of the remaining ones pending assessment.

Among the 1500 apps retrieved for AR and the several hundred retrieved for CRS, only a handful were selected for review, including three multilingual apps and two using a single language in AR. That is, while there are several apps claiming to be health-related, only a few have been studied in a relevant manner, prompting the need for some quality control over health-related apps. This may not only concern AR and CRS but also other chronic diseases.

Apps studied in a relevant manner were found to be of interest in the diagnosis, management and cost-effectiveness of AR, as attested by the several examples presented in this review. In CRS, only one app has published results (and in only one single paper). There is an urgent need to validate other apps.

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Sanna Toppila-Salmi reports grants from GSK, personal fees from AstraZeneca, personal fees from ALK Abelló, personal fees from ERT, personal fees from GSK, personal fees from Novartis, personal fees from Sanofi Pharma, personal fees from Roche, outside the submitted work. 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Joanna Tsiiligianni reports grants from Grants from Boehringer Ingelheim, Glaxo Smithkline, Astra ZeneCA and personal fees for advisory boards or speaker bureau from Astra ZeneCA, Chiesi, Novartis, outside the submitted work. Charlotte Suppli Ulrik reports grants and personal fees from AZ, personal fees from GSK, personal fees from Novartis, personal fees from BI, grants and personal fees from Sanofi, personal fees from Orion Pharma, personal fees from Pfizer, personal fees from TEVA, personal fees from Novartis, from the submitted work; Torsten Zubier reports personal fees from AstraZeneca, personal fees from AbbVie, personal fees from ALK, personal fees from Almirall, personal fees from Astellas, personal fees from Bayer Health Care, personal fees from Bencard, personal fees from Berlin Chemie, personal fees from FAES, personal fees from HAL, personal fees from Leti, personal fees from Meda, personal fees from Menarini, personal fees from Merck, personal fees from MSD, personal fees from einer, personal fees from Pfizer, personal fees from Sanofi, personal fees from
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Data sharing is not applicable to this article as no new data were created or analysed in this study.

ORCID
Bernardo Sousa-Pinto https://orcid.org/0000-0002-1277-3401
Oliver Pfarr https://orcid.org/0000-0003-4374-9639
Victoria Cardona https://orcid.org/0000-0003-2197-9767
Cemal Cingi https://orcid.org/0000-0003-3934-5092
Paolo M. Matricardi https://orcid.org/0000-0001-5485-0324
Joaquim Mullol https://orcid.org/0000-0003-3463-5007
Hae-Sim Park https://orcid.org/0000-0003-2614-0303
Sanna Toppila-Salmi https://orcid.org/0000-0003-0890-6686
Maria Teresa Ventura https://orcid.org/0000-0002-2637-4583
Luo Zhang https://orcid.org/0000-0002-0910-9884
Mihaela Zidarn https://orcid.org/0000-0003-0515-5207
Jean Bousquet https://orcid.org/0000-0002-4061-4766

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