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Trials for the development and evaluation of digital therapeutic apps: a scoping review

LAUREN BELL

London School of Hygiene & Tropical Medicine, UK

ABSTRACT

There is a growing recognition that evaluating a digital therapeutic app within a randomised controlled trial (RCTs) presents unique methodological challenges. We consider two of these challenges: the first is that good practise for app development often begins with a ‘Minimum Viable Product’, then the app rapidly evolves from knowledge gained with real-world use and testing. The second challenge is that digital therapeutic apps are often complex interventions, with specific components that are behaviour change interventions. RCTs are a poor fit because they (i) do not allow the app to continually improve from the data gathered during the trial and (ii) results only allow us to understand the effectiveness of the whole app, and not individual components.

There is a need for more agile trial designs that synchronise and balance the two separate goals of learning how to continually improve an intervention, while gathering information to assess overall effectiveness. Researchers have suggested alternative trial designs to be suitable to evaluate apps, including: Sequential Multiple Assignment Randomised Trials (SMARTs) for dynamic treatment regimens; micro-randomisation trials (MRT) for ‘just-in-time’ push notifications; N-of-1 and series of N-of-1 for personalisation of apps; randomised response adaptive trials for allocating more patients the most effective app, Multiple Optimisation Strategy (MOST) framework and Multi-Armed Bandit Models (MABM) for building and optimising apps as complex interventions.

In this presentation, I consider various advantages and disadvantages of each design, how are these trial designs implemented for the development and evaluation of digital therapeutic apps and opportunities to advance trial methodology.

mHealth systems to advance how we understand and approach mental illness

NIRANJAN BIDARGADDI

Flinders University, Australia

ABSTRACT

Fostering population mental health effectively remains a greatest ongoing challenge. A large majority of individuals with common mental disorders never seek professional treatment. Furthermore sustaining engagement with effective interventions has proven challenging with low adherence rates. One potential reason for this is that cognitive and behavioural interventions, despite being effective, require a great deal of energy and focused attention, which is made more difficult by the effect of mood disorders on individuals' motivation and cognitive capacities. Equally at the severe end effective clinical management of individuals with severe mental illnesses outside hospitals remains challenging due to ineffective monitoring systems and fragmented care providers. This talk will present digital resources devised in Australia to overcome many of the barriers experienced by psychiatry and mental health care sector. It covers mhealth approaches to improve access, screen for symptoms early, monitor in real time, and integrate treatments into the fabric of daily lives of patients in a personalised manner. Equally, for more severe mental illness, opportunities afforded by administrative claims data sets to monitor risks of hospitalisation and relapse in real time are outlined.

Mobile AI for personalised healthcare

NGAI-MAN CHEUNG

Singapore University of Technology and Design, Singapore

ABSTRACT

In this talk, I will first discuss our translational research on mobile healthcare using AI, computer vision and smartphones. Then, I will discuss some of our fundamental AI research that is important for mobile healthcare, including deep model compression, generative adversarial networks and few shot learning. I will give an overview of these techniques, highlight their relevance for mobile healthcare, describe the technical challenges and discuss our work to advance state-of-the-art.

ROADMAP for precision and digital health

YING KUEN (KEN) CHEUNG

Columbia University, USA

ABSTRACT

Precision medicine at its core seeks to provide methodologic solutions to the challenge of patient heterogeneity and heterogeneous response to therapy. In developing mobile health applications, for example, how a user interacts with smartphone apps depends on the system recommendations (actions) in a dynamic way. While a SMART (Sequential Multiple Assignment Randomized Trial) allows for unbiased evaluation of the optimal action at each decision point, its use in practice could be limited due to the lack of guidance on how such a trial should be designed and implemented. At the same time, the notions of “N-of-1” and master protocols are very appealing to clinical researchers in precision trials. In this talk, I will compare and contrast the design issues in these two areas. I will also highlight some recent findings on SMART with regard to sample size and adaptation, and activities at the Columbia’s ROADMAP initiative for digital health.

References

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Use of RFID to detect proximity of individuals and their location in closed settings for outbreak modelling

ALEX COOK

National University of Singapore, Singapore

ABSTRACT

Social contacts are one of the main drivers of many infectious diseases, and heterogeneities in contacts influence who is at greatest risk of infection and may guide control efforts. Social contacts can be measured through surveys, such as 24h recall paper-based surveys, but in some conditions it may be possible to measure contacts directly, in particular in closed populations such as schools. In this study we measured contacts in several hundred individuals in such a closed population through RFID tags which we developed. These were either worn by the participant, or wall mounted and geo-referenced. In this talk I will discuss how we sought to analyse these data to discern patterns in contacts.

Personalized data science, small data, and N-of-1 trials

NAIHUA DUAN

Columbia University in the City of New York, USA

ABSTRACT

Data science has emerged in recent decades as an influential discipline to extract knowledge from data, usually applying machine learning technologies to cross-sectional big data obtained from a large number of individuals, such as electronic health records from large clinical organizations. At the same time, personalized data science (PDS) has the potential to expand existing framework for data science in several important directions, for the development and implementation of personalized healthcare. PDS provides tools to facilitate the acquisition and interpretation of individual data (small data) from individual patients acting as citizen scientists, to inform each individual's unique healthcare decisions, both medical and lifestyle, through self-studies such as self-experimentation.

Self-experimentation made important contributions to medicine throughout history, tracing back thousands of years ago to the legendary Emperor ShenNong, who sampled hundreds of herbs to ascertain their therapeutic efficacy, laying the foundation for Chinese Medicine. In recent centuries, self-experimentation by clinical investigators has led to numerous important discoveries, including no less than eight that were awarded the Nobel Prize in Physiology or Medicine, such as the discovery of the bacterium *H. pylori* as a common cause for gastritis and peptic ulcer by Dr. Barry Marshall.

At the same time, self-studies such as self-experimentation by individual patients acting as citizen scientists, such as the tens of thousands of participants in citizen scientist communities such as QuantifiedSelf.com, have the potential to inform their healthcare decisions and improve their health and well-being. However, citizen scientists usually lack practical rigorous tools to conduct scientifically valid self-studies. With the advance in information and communication technologies, especially mobile technologies, PDS has the potential to fulfill this important demand, to assist citizen scientists to design their self-studies, acquire their data, and analyze/interpret their data, to inform their medical or lifestyle decisions.

An intriguing aspect of ShenNong's legend is that he (as a divine being) had a transparent body, and thus could see the effects of different plants and herbs on himself. For mortals, modern technologies in effect begins to render our bodies transparent, approaching ShenNong's divine power. As an example, an auto-titration CPAP device is capable to sense, monitor, record, and transmit the sleep conditions for a patient with sleep apnea throughout a night of sleep, and also makes dynamic adjustments to device settings in real time to improve the patient's sleep condition.

As an illustration of the utility of PDS in personalized healthcare, I will discuss the Personalized Research for Monitoring Pain Treatment (PREEMPT) study, led by Dr. Richard Kravitz, which studied the deployment of personalized (N-of-1) trials to inform therapeutic decisions for patients with chronic musculoskeletal pain. The study developed an mHealth app, Trialist, for patients and their clinicians to design and implement their individual personalized trials, to obtain outcome data, to analyze the data, and to produce reports to inform their therapeutic decisions. The Trialist app allows highly personalized trial design features including the therapeutic options to be compared, the number of crossovers and the duration of each therapeutic episode, etc. The study recruited 215 patients and randomized 108 to the Trialist/personalized trial condition, and 107 to the control (standard care) condition. The primary findings from the study are available in Kravitz et al. (2018 JAMA Intern Med).

Personalized lifestyle interventions – opportunities and challenges

MUELLER-RIEMENSCHNEIDER FALK

National University of Singapore, Singapore

ABSTRACT

Promotion of healthy lifestyles, including physical activity, healthy eating, weight management and abstaining from tobacco use, is key to public health efforts for prevention of NCD. Effective NCD prevention thus requires a rigorous understanding of the patterns and determinants of these lifestyle risk factors in the general population and in patients affected by NCD. Knowledge of distribution and longitudinal trends for risk factors in the population, enables identification of the socio-demographic subgroups, and stages of life (e.g. related to ageing or changes in work status, marital status, or having children) that can be prioritized for preventive interventions. Further personalizing interventions to the individual, based on their specific behavioral risk factors, as well as their cultural, social, psychological, economic and environmental characteristics, is anticipated to enhance long-term behavior change and chronic disease management. The use of evolving technologies and digital health approaches is particularly suitable to overcome limitations of existing approaches and more deeply and continuously characterize individual's health behaviors and contextual factors. Leveraging on these technologies the Ecological Momentary Assessment (EMA) approach can be applied for collection of comprehensive data to advance the current practice of epidemiological and clinical research and to facilitate the development of personalized lifestyle interventions. This presentation provides background for the rationale of personalized lifestyle interventions with a specific focus on physical activity and the application of EMA. It will use examples from research projects conducted in Singapore that are concerned with the development and integration of smartphones, apps, wearables and other technologies and it will illustrate practical and technological challenges in this process.

Dynamic generalized odds-ratio (dGOR): a novel approach to assess dynamic treatment regimes (DTR) with an ordinal outcome

PALASH GHOSH AND BIBHAS CHAKRABORTY

Duke-NUS Medical School, National University of Singapore, Singapore

ABSTRACT

Sequential multiple assignment randomized trials (SMART) are used to construct data-driven optimal treatment strategies for patients based on their treatment and covariate histories in different branches of medical and behavioral sciences where a sequence of treatments are given to the patients; such sequential treatment strategies are often called dynamic treatment regimes (DTR). In the existing literature, the majority of the analysis methodologies for SMART studies assume a continuous primary outcome. However, ordinal outcomes are also quite common in clinical practice; for example, the quality of life is often measured in an ordinal scale (e.g., poor, moderate, good). In this work, first, we develop the notion of dynamic generalized odds-ratio ($dGOR$) to compare two dynamic treatment regimes embedded in a 2-stage SMART with an ordinal outcome. We propose a likelihood-based approach to estimate $dGOR$ from SMART data. Next, we discuss some combinatorial properties of $dGOR$ and derive the asymptotic properties of its estimate. We discuss some alternative ways to estimate $dGOR$ using concordant-discordant pairs and two-sample U -statistic. Then, we extend the proposed methodology to a K -stage SMART. Furthermore, we propose a basic policy search algorithm that uses $dGOR$ to find an optimal DTR within a finite class. A simulation study shows the performance of the estimated $dGOR$ in terms of the estimated power corresponding to the derived sample size. We analyze data from Sequenced Treatment Alternatives to Relieve Depression (STAR*D), a multi-stage randomized clinical trial for treating major depression, to illustrate the proposed methodology. A freely available online tool using R statistical software is provided to make the proposed methodology accessible to other researchers and practitioners.

Some aspects of SMART design: methodological developments and an application in mHealth intervention

SAMIRAN GHOSH

Wayne State University School of Medicine, USA

ABSTRACT

This is really a two part talk. In the First part, we will discuss some cutting edge statistical methodology development for testing non-inferiority (NI) in the SMART setup. NI trials are becoming popular as comparative effectiveness research are becoming mainstream and fewer and fewer block-buster discoveries are made for many legacy disease areas. However, unlike traditional SMART the trial also involves a putative placebo or standard of care arm. In standard Randomized Controlled Trial (RCT) context this gives rise to three-arm NI trial consisting of Reference, Experimental and Placebo arm. However, the situation is slightly more complicated in the SMART-NI context when a third placebo arm is present as defining NI margin is not so straightforward. NI trial has other challenges such as assay sensitivity and constancy which is unique to NI-RCT (as well as for NI-SMART). In this talk we will introduce a “fraction of effect retention” based approach for designing three-arm NI-SAMRT. A detailed sample size and power analysis plan will be also discussed. We will show a NIH supported current ongoing trial for improvement of symptom management among cancer patients where such design could be useful.

In the Second part of the talk we will discuss a real world application of SMART design for Evaluating Implementation of Text Messaging, Cell Phone Support, and Contingency Management in Youth Non-adherent to Antiretroviral Medication. This is a current ongoing NIH funded trial where mobile device based interventions are used for increasing adherence to Antiretroviral Treatment among HIV infected youths. We will discuss the current state of the research and some of the practical challenges we have faced in implementing mHealth intervention in practice.

Hybrid statistical and mechanistic mathematical model guides mobile health intervention for chronic pain

SARA M. CLIFTON^a, DANIEL M. ABRAMS^b, CHAERYON
KANG^c, JINGYI JESSICA LI^d, QI LONG^e, AND NIRMISH
SHAH^f

^a*Department of Mathematics, University of Illinois at Urbana-Champaign, USA*

^b*Department of Engineering Sciences and Applied Mathematics, Northwestern
University, USA*

^c*Department of Biostatistics, University of Pittsburgh, USA*

^d*Department of Statistics, University of California at Los Angeles, USA*

^e*Department of Biostatistics and Epidemiology, Perelman School of Medicine,
University of Pennsylvania, USA*

^f*Department of Medicine, Duke University, USA*

ABSTRACT

Nearly a quarter of visits to the emergency department are for conditions that could have been managed via outpatient treatment; improvements that allow patients to quickly recognize and receive appropriate treatment are crucial. The growing popularity of mobile technology creates new opportunities for real-time adaptive medical intervention, and the simultaneous growth of “big data” sources allows for preparation of personalized recommendations. Here we focus on the reduction of chronic suffering in the sickle cell disease (SCD) community. SCD is a chronic blood disorder in which pain is the most frequent complication. There currently is no standard algorithm or analytical method for real-time adaptive treatment recommendations for pain. Furthermore, current state-of-the-art methods have difficulty in handling continuous-time decision optimization using big data. Facing these challenges, in this study, we aim to develop new mathematical tools for incorporating mobile technology into personalized treatment plans for pain. We present a new hybrid model for the dynamics of subjective pain that consists of a dynamical systems approach using differential equations to predict future pain levels, as well as a statistical approach tying system parameters to patient data (both personal characteristics and medication response history). Pilot testing of our approach suggests that it has significant potential to well predict pain dynamics, given patients reported pain levels and medication usages. With more abundant data, our hybrid approach should allow physicians to make personalized, data-driven recommendations for treating chronic pain.

Interpreting step count activity patterns in an incentivized city-wide health intervention

BRIAN Y. LIM

National University of Singapore, Singapore

ABSTRACT

Activity trackers are being deployed in large-scale physical activity intervention programs, but analyzing them is difficult due to the large data size and complexity. As large datasets of steps become more available, it is paramount to develop analysis methods to more deeply interpret them to understand the variety and changing nature of human steps behavior. In this work, we explored ways to analyze the heterogeneous steps activity data and propose a framework of dimensions and time aggregations to interpret how activity trackers with monetary incentives influence the steps behavior of a city-wide population. We analyzed the daily step counts of 140,000 individuals walking a combined 74 billion steps in 305 days during a city-wide public health campaign. We performed data mining clustering to segment users into 16 types of users with various types of walking behaviors and demonstrate that these clusters help with interpreting how some user increased their steps level. Our findings inform how to scalably interpret large steps data to draw behavioral insights from intervention studies.

Stratified micro-randomized trials with applications in mobile health

SUSAN MURPHY

Harvard University, USA

ABSTRACT

Technological advancements in the field of mobile devices and wearable sensors make it possible to deliver treatments anytime and anywhere to users like you and me. Increasingly the delivery of these treatments is triggered by detections of vulnerability. Two challenges are that vulnerability may be impacted by prior treatment and treatment provided at time t is expected to have an impact on users over a span of time during which subsequent treatments may be provided. Here we discuss our work on the design of a mobile health smoking cessation study in which the above two challenges arose. Multiple online data analysis algorithms are used for detection, for example, of physiological stress as well as forecasting the remaining number of vulnerable times in the day. These algorithms are then inputs into a randomization algorithm that ensures that each user is randomized to each treatment an appropriate number of times per day. The stratified micro-randomized trial involves not only considerations of the randomization algorithm but a precise statement of the meaning of the causal treatment effects along with primary analyses and sample size calculations.

Estimating time-varying causal effect moderation in mobile health with binary outcomes

TIANCHEN QIAN

Harvard University, USA

ABSTRACT

Binary outcome is common in mobile health studies. We focus on estimating the time-varying causal effect moderation for data from micro-randomized trials with binary outcomes. We give the definition of moderated treatment effect in this setting, and provide two estimation methods. One estimation method is for the proximal treatment effect conditional on the entire history; this estimator is semiparametric locally efficient. The other estimation method, based on a weighted and centered least squares approach, is for the proximal treatment effect marginal but conditional only on a subset of variables in history. Both estimators are robust in the sense that they do not require a correct control model. The methods are illustrated by simulation studies and a real data example.

App and system for diet and lifestyle tracking for chronic diseases management and research

JONATHAN STANIFORTH AND ZHAO-YAN MING, ZOE

National University of Singapore, Singapore

ABSTRACT

The prevalence of chronic diseases such as hyperglycemia, hypertension, and hyperlipidemia (3H) becomes a major concern for public health in many countries, including Singapore and China. 3H are strongly lifestyle-related and are preventable health problems that can silently become costly diseases such as stroke, heart disease, and kidney disease. Research has shown that self-management support intervention most frequently resulted in significant improvements in patient-level outcomes; nutritional practices alone can reduce the risk of cardiovascular disease by 60%. Unfortunately, efforts to promote sustained healthy eating habits and exercise have been mostly unsuccessful, largely due to the lack of accurate and reliable ways to log an individual's real time food intake, and other lifestyle data. We aim to tackle the problem of 3H in the population by leveraging state-of-the-art AI technologies to empower patient self-management and to support primary care practitioners using mobile apps and systems. To facilitate the tasks of lifestyle data gathering, analytics, action planning and sharing: the four key AI research technologies underpinning the whole research include: advanced AI lifestyle analytics; private and explainable machine learning; personalized nudging and influence; and incentives for data sharing and analytics.

The personal assistant App together with the backend distributed system will integrate the AI-based technologies and link up AI researchers, experienced primary care, clinical and nutritional researchers, first-line practitioners and systems architects. The infrastructure has the potential to be used to conduct various lifestyle and 3H related large-scale feasibility studies and randomized controlled trials.

Perpetually enhancing technology for human learning and behavior change, through dynamic, personalized, collaborative experimentation

JOSEPH JAY WILLIAMS

University of Toronto, Canada

ABSTRACT

There is a proliferation of websites and mobile apps for helping people learn new concepts (e.g. online courses), and learn how to change health habits and behavior (e.g. websites for reducing depression, apps for quitting smoking). How can we use data from real-world users to rapidly enhance and personalize these technologies? I show how we can build self-improving systems through three applications of MOOClets/AdapComps, a conceptual framework implemented in technology that leverages randomized A/B experiments as tools for collaboration, dynamic enhancement, and adaptive personalization.

First, a novel system that enhanced learning from K12 math problems, by crowdsourcing explanations and using machine learning to automatically experiment to discover the best explanations. Second, a system which enabled three on-campus instructors at Harvard to experimentally investigate which hints and feedback messages students found helpful, enabling more ethical experimentation by dynamically presenting the best conditions to future students. Third, I show how to boost responses to an email campaign in a MOOC, by experimentally discovering how to personalize motivational messages to a user's activity level.

These self-improving systems use experiments as a bridge between designers, social-behavioral scientists and researchers in statistical machine learning. I'll present future directions, such as investigating which reflection prompts help students learn, how to enhance motivation through social-psychological interventions, and how to personalize web-apps that help students set and achieve micro-goals. I will also discuss efforts to bridge education with health behavior change and marketing.

A digital community-based syndromic surveillance system for influenza and other acute respiratory infections in Singapore

PEILING YAP

National Centre for Infectious Diseases, Singapore

ABSTRACT

In Singapore, influenza affects nearly 20-25% of the population. About 1500 people are hospitalized and 588 die each year due to influenza. The World Health Organization (WHO) recommends a comprehensive suite of data streams for influenza surveillance ranging from a broad based community syndromic surveillance, through to district clinics and national hospitals. Traditional surveillance methods usually rely on the symptomatic person visiting a healthcare facility and such systems can be made less efficient by poor health-seeking behaviour and delays in disease notifications. As smart-phone technology gets more pervasive, it allows us an opportunity to address the base of the pyramid of disease surveillance data gathering by making community syndromic surveillance viable through a digital system. Given that Singapore is one of the more digitally connected and mobile-savvy countries in the world, a digital, community-based syndromic surveillance system for influenza can arguably complement the existing national surveillance system, making it more comprehensive and far-reaching into the population. In this presentation, we will showcase mobile applications we have developed, pilot tested and evaluated among working adults in an attempt to build a digital, community-based syndromic surveillance system in Singapore.

A gate-keeping approach to selecting adaptive interventions under general SMART designs

TONY ZHONG

Icahn School of Medicine at Mount Sinai, USA

ABSTRACT

Background: An objective of a sequential multiple assignment randomized trial (SMART) is to identify an optimal adaptive intervention that maximizes the expected value of certain outcomes. While this task can be performed by comparing all possible pairs of interventions with appropriate multiplicity adjustments such as the Bonferroni's correction, the pairwise comparison approach generally suffers substantial loss in power because a SMART typically consists of many adaptive interventions. In addition, there is relatively little discussion on sample size calculation for SMARTs under general settings.

Purpose: To propose an omnibus likelihood ratio test as a gate-keeping test for selecting an optimal adaptive intervention, and to develop a sample size calculation procedure for the test.

Method: We derive the asymptotic distribution of the proposed likelihood ratio test, outline the steps to perform sample size calculation based on the asymptotic results, and evaluate the test's finite sample performance using simulation. The method is illustrated in the context of the CODIACS trial.

Results: The simulation shows that asymptotic approximation performs well for a moderate sample size, and the test achieves the nominal significance level. In addition, the proposed omnibus test has better power than a pairwise comparison approach with Bonferroni's adjustment, with an improvement up to 29 percentage points in our simulation scenarios. When compared to an existing omnibus test based on inverse probability weighting estimation, the proposed test reduces the required sample size by up to 17%.

Limitations: The required sample size by the proposed test increases considerably with the number of stages in a SMART. Large sample size requirement is an inherent difficulty with SMARTs as a result of curse of dimensionality.

Conclusion: The proposed omnibus test is applicable to perform intervention selection under general SMART design settings. The sample size calculation procedure provides a rigorous justification on how large a SMART should be.