Is there a causal relationship between acute stage sensorimotor cortex activity and the development of chronic low back pain? a protocol and statistical analysis plan

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ABSTRACT

Introduction Why some people develop chronic pain following an acute episode of low back pain is unknown. Recent cross-sectional studies have suggested a relationship between aberrant sensorimotor cortex activity and pain persistence. The UPWaRD (Understanding persistent Pain Where it ResiDes) cohort study is the first prospective, longitudinal investigation of sensorimotor cortex activity in low back pain. This paper describes the development of a causal model and statistical analysis plan for investigating the causal effect of sensorimotor cortex activity on the development of chronic low back pain.

Methods and analysis Sensorimotor cortex activity was assessed within 6 weeks of low back pain onset using somatosensory evoked potentials and transcranial magnetic stimulation mapping techniques. Chronic low back pain is defined as ongoing pain (Numerical Rating score ≥1) or disability (Roland Morris Disability Questionnaire score ≥3) at 6 months follow-up. Variables that could confound the relationship between sensorimotor cortex activity and chronic low back pain were identified using a directed acyclic graph and content expertise was used to specify known causal paths. The statistical model was developed ‘a priori’ to control for confounding variables identified in the directed acyclic graph, allowing an unbiased estimate of the causal effect of sensorimotor activity in acute low back pain on the development of chronic pain. The statistical analysis plan was finalised prior to follow-up of all participants and initiation of analysis.

Ethics and dissemination Ethical approval has been obtained from Western Sydney University Human Research Ethics Committee (H10465) and from Neuroscience Research Australia (SSA: 16/002). Dissemination will occur through presentations at national and international conferences and publications in international peer-reviewed journals.

Trial registration number ACTRN12619000021899 (retrospectively registered)

BACKGROUND

Low back pain (LBP) is the most common form of persistent musculoskeletal pain and a leading cause of disability.1 In 2012, the direct healthcare costs of LBP in Australia were estimated at $4.8 billion,2 while in the USA, this figure approaches $50 billion.3 Most of these costs are associated with chronic LBP, that is, pain that has persisted for more than 3 months. It is not understood why up to 40% of people with acute LBP develop chronic LBP.4 Interventions to prevent the development of chronic LBP have not been effective.5 Identifying the causal mechanisms that explain why some people develop chronic LBP may guide the development of targeted treatment and is considered a research priority.6 7

Causes of a health condition are defined as characteristics or events necessary for the condition to occur,8 that is, ‘had the exposure differed, the outcome would have differed’.9 Aberrant sensorimotor cortex activity in the acute stage of LBP is one characteristic

Strengths and limitations of this study

- The causal objective for data obtained from the UPWaRD (Understanding persistent Pain Where it ResiDes) study is made explicit and transparent within this protocol and analysis plan.
- Acknowledging the causal goal of this research can inform scientific discussion of future results.
- Detailed description of confounder selection using a directed acyclic graph is transparently reported ‘a-priori’.
- A causal analysis in observational data can be viewed as an attempt to emulate a hypothetical trial — ‘the target trial’. Currently, it remains challenging to sufficiently define sensorimotor cortex activity as an ‘intervention’.
- There can be no guarantee that a causal model incorporates all confounders.
postulated to have a causal relationship with the development of chronic pain.\textsuperscript{16–12} Cross-sectional studies have shown larger activity and a shift in the S1 representation of the back\textsuperscript{11} and enlarged and shifted M1 representations of the back muscles in chronic LBP compared with pain-free individuals.\textsuperscript{13–14} and these changes are associated with pain, functional impairment and symptom chronicity.\textsuperscript{14–18} Further, preliminary evidence suggests sensorimotor cortex activity in acute LBP is lower in patients with acute clinical LBP than in pain-free controls.\textsuperscript{10} Despite these findings, no study has investigated the causal relationship between sensorimotor cortex activity in acute LBP and the development of chronic pain.

Recent conceptual advances have outlined methods for estimating the causal effect of an exposure on a health outcome using observational data.\textsuperscript{19} Two major considerations when attempting to estimate causal effect include how a particular target trial is emulated with observational data and the appropriate selection of confounding variables.\textsuperscript{20–21} Identification and inclusion of confounding variables in a statistical model is essential to estimate causal effects.\textsuperscript{22} Confounding occurs when an exposure and outcome share a common cause.\textsuperscript{22–24} Data driven identification of confounding variables such as p value based and model-based variable selection methods ignore the causal structure underlying the hypothesis and subsequently do not aid in causal inference.\textsuperscript{15–18} Rather, expert knowledge is required to specify the causal structure.\textsuperscript{25} Causal models can be represented visually using directed acyclic graphs (DAGs).\textsuperscript{9, 19} A DAG provides a graphical representation of a mathematically rigorous method for minimising confounding bias within observational research.\textsuperscript{8} While there can be no certainty a causal model incorporates all known confounders, this approach to identifying confounding bias and developing a causal model makes assumptions explicit and transparent, promoting informed scientific discussion.\textsuperscript{20}

Using data from the UPWaRD (Understanding Where It ResiDes) prospective, longitudinal cohort study, this paper reports the development of a causal model to investigate whether sensorimotor cortex activity (exposure) in the acute stage of LBP has a causal effect on the development of chronic LBP (outcome). First we describe the protocol for data collection and development of a DAG, detailing the explicit assumptions for identification of confounding variables within a causal model of chronic LBP.\textsuperscript{19} Second, we report a prespecified statistical analysis plan in line with the STROBE (Strengthening the Reporting of Observational Studies in Epidemiology) statement.\textsuperscript{20}

**METHODS**

**Design**

The UPWaRD study is a multicentre, prospective, longitudinal, cohort study of people presenting with acute LBP (National Health and Medical Research Council of Australia, Grant ID: 1059116). The study was registered with the Australian and New Zealand Clinical Trials Registry and the full study protocol has been published.\textsuperscript{31} The study enrolled 120 participants with acute LBP, with each participant undergoing a battery of neurophysiological and psychological tests at baseline with follow-up completed at 6 months.

**Patient and public involvement**

Patients and the public were not involved in the design of this protocol and statistical analysis plan. Patient advocacy groups (Chronic Pain Australia, Pain Australia) provided support for recruitment through dissemination of recruitment flyers in newsletters, websites and social media. Individual test results will be provided to participants on request and a summary of the overall outcomes of the study will be available to all participants on completion of the trial.

**Objective**

The primary aim of the UPWaRD study was to determine whether sensorimotor cortex activity, an individual’s capacity for neuroplasticity, and psychosocial features, assessed during an acute episode of LBP could predict 6 month LBP outcome.\textsuperscript{31} As predictive models have different aims to studies investigating causal inference, this paper outlines the statistical analysis plan for using data obtained from the UPWaRD study to investigate whether sensorimotor cortex activity has a causal effect in the development of chronic LBP.

**Hypothesis**

Null hypothesis: Sensorimotor cortex activity in the acute stage of LBP (T1) does not cause chronic LBP at 6 months. The null hypothesis will be rejected if sensorimotor cortex activity demonstrates a significant causal relationship with chronic LBP (pain or disability) at 6 month follow-up (T2).

**Inclusion criteria**

Participants were eligible for inclusion in the study if the following criteria were met:

- 18 years or older and experiencing acute non-specific LBP — defined as pain in the region of the lower back, superiorly bound by the thoracolumbar junction and inferiorly by the gluteal fold\textsuperscript{23}.
- Experiencing a new episode of acute LBP defined as pain present for more than 24 hours and less than 6 weeks’ duration following a period of at least 1 month pain-free;\textsuperscript{24–26}
- Did not have known or suspected serious spinal pathology (for example, fracture, malignancy, inflammatory or infective diseases of the spine; cauda equina syndrome or neurological disorder);
- Did not have a history of previous lumbar spinal surgery (eg, spinal fusion, intervertebral disc replacement);
- Did not report suspected or confirmed pregnancy and/or were less than 6 months’ post-partum;
Figure 1  Directed acyclic graph to identify confounders. Confounding variables (theoretical causal effect on exposure (sensorimotor cortex activity) and outcome (chronic LBP)) are in red circles. Grey circles are variables that were unmeasured in the UPWaRD (Understanding persistent Pain Where it ResiDes) study. LBP, low back pain.

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Jenkins L, et al. BMJ Open 2019;9:e035792. doi:10.1136/bmjopen-2019-035792

- Did not present with suspected radicular pain (dominant leg pain, positive neural tissue provocation tests and/or any two of altered strength, reflexes or sensation for the same nerve root, assessed clinically);
- Were free from the presence of another painful condition (eg, fibromyalgia, neuropathy, rheumatoid arthritis);
- Did not report serious comorbidities affecting sensorimotor function or causing neurological deficit (eg, multiple sclerosis, spinal cord injury);
- Did not report a history of psychological disorders requiring medication for symptom control (eg, major depressive disorder, bipolar disorder, schizophrenia);
- Demonstrated no contraindications for the application of transcranial magnetic stimulation;
- Provided written informed consent to participate and were able to speak and read English.

Outcome variables: pain and disability

Primary outcome
The primary outcome is pain intensity. Self-reported pain scores are determined using the Brief Pain Inventory at T1 and T2. Participants are asked to score their pain intensity on average over the previous week on a 11-point numerical rating scale (NRS: 0='no pain', 10='worst pain imaginable'). Pain intensity scores at T2 will also be dichotomised to determine ‘recovered’ and ‘non-recovered’ participants. A NRS score of 0 will be classified as recovered LBP and a NRS score ≥1 will be classified as chronic LBP.

Secondary outcome
The secondary outcome is disability. Self-reported disability will be determined using the 24-point Roland Morris Disability Questionnaire (RMDQ) at T1 and T2.

Exposure variables: sensorimotor cortex activity
Sensory cortex activity in the acute stage of LBP will be assessed using the peak-to-peak area of the N80 and N150 components of the sensory evoked potential (SEP). Motor cortex activity will be assessed using transcranial magnetic stimulation (TMS) derived map volume of the paraspinal muscles at the L3 and L5 spinal level. These procedures have been outlined in detail in the UPWaRD study protocol. In brief:

SEPs are recorded in response to two blocks of 500 non-noxious electrical stimuli applied via a constant current stimulator (Digitimer, D570H) to the paraspinal muscles 3 cm lateral to the L3 spinous process, ipsilateral to the side of worst LBP. Electroencephalography (EEG) is recorded using gold plated cup electrodes (Digitimer, Reusable Au and Ag EEG Cup Electrodes) positioned over S1 (3 cm lateral and 2 cm posterior to Cz) on the side contralateral to worst LBP and referenced to Fz according to the International 10/20 EEG placement system. The N80 component is thought to represent activity in S1 (between the first major downward deflection of the curve after stimulation and the first major negative peak, N80), the N150 component is thought to represent activity in the secondary sensory cortex (S2) (between the first negative peak, N80, and second negative peak, N150).

L3 and L5 map volume is the measure of total excitability of the corticomotor representation of the paraspinal muscles recorded at L3 and L5 level. Participants undergo a standardised TMS protocol.
mapping procedure. Single-pulse, monophasic TMS is delivered to the M1 contralateral to the side of worst LBP (Magstim 200 stimulator/7 cm figure-of-eight coil; Magstim Co Ltd, Dyfed, UK). The stimulator intensity is set to 100%, with an interstimulus interval of ~5 s. Surface electromyography (EMG) is recorded from the paraspinal muscles with an electrode (silver-silver chloride disposable electrodes; Noraxon USA Inc, Arizona, USA) placed longitudinally, 3 cm lateral to the L3 spinous process and 1 cm lateral to the L5 spinous process, ipsilateral to the side of worst LBP. Five stimuli are delivered over pre-marked scalp sites on a 6×7 cm grid, commencing at the vertex, determined using the International 10/20 System. EMG traces of the five motor evoked potentials recorded at each scalp site are averaged then superimposed over the respective scalp sites to construct a topographical representation of the paraspinal muscle. All TMS data is analysed using MATLAB 7 (The MathWorks, USA).

Identifying sources of confounding

A directed acyclic graph was constructed using DAGitty software to identify all variables that have a plausible, causal effect on the relationship between sensorimotor cortex activity (the exposure) and chronic LBP (the outcome). Figure 1 details all variables included within the DAG. The DAG outlines explicit assumptions made by the investigators, informed by expert opinion and current literature. Table 1 details data collected from the UPWaRD study that can be used to control for the identified confounding variables. Procedures for obtaining these variables are outlined in detail in the UPWaRD study protocol.

**Limitations**

The DAG methodology is not without limitation, establishing the directionality of effects in addition to model misspecifications can result in errors, potentially leading to incorrect inferences. Simpler and more sparse DAGs represent stronger assumptions, as every omission of a variable and its causal pathway represents an assumption of one or more causal null hypotheses. Further, DAGs do not account for the effect of unmeasured confounding. The effect of unmeasured confounding on the study results will be analysed using a sensitivity analysis. A sensitivity analysis determines how important unmeasured confounding would need to be to alter study conclusions.

| Assessment domain | Confounding variable |
|-------------------|----------------------|
| Predisposing factors | 1. Age |
|                    | 2. Sex |
|                    | 3. Previous history of low back pain: Participants are asked the following question: ‘Have you experienced low back pain in the past?’ |
|                    | 4. BDNF genotype: Cheek swabs taken on the day of baseline testing are used to prepare genomic DNA (Isohelix DNA Isolation Kit). |
|                    | 5. Socioeconomic status: Participant postal code is converted into a SEIFA score. |
|                    | 6. Cultural diversity: Participants are asked the following question: ‘How do you define your identity, in ethnic or cultural terms?’ |
| Blood biomarkers   | 7. BDNF serum concentration: Peripheral venous blood is drawn into serum tubes (BD Vacutainer, SST II Advance). BDNF serum concentration is measured using an enzyme-linked immunosorbent assay (ELISA) (Simple Plex Cartridge Kit, Biotrend). |
|                    | 8. Pro-inflammatory cytokines: Serum samples obtained from the UPWaRD study will also be analysed for TNF, IL-1β, IL-6 and CRP. Zero is allocated for values below the test sensitivity. |
| Psychological variables | 9. PCS: Assesses catastrophising thoughts about pain. The PCS includes 13 items, scored on a 5-point scale. |
|                    | 10. DASS-21: Includes three 7-item subscales with higher scores indicating greater depression, anxiety and/or stress. |
|                    | 11. PSEQ: Evaluates the confidence of an individual in their ability to perform a range of functional activities while in pain. A total score between 0 and 60 is calculated, higher scores indicate greater self-efficacy beliefs. |
| Sensitisation      | 12. Local sensitivity: PPT is measured using a hand-held pressure algometer (Somedic, Hörrby, Sweden). The probe (size 1 cm²) is applied perpendicular to the skin until the participant reports the sensation has changed from pressure to pain. PPT is measured three times ipsilateral to the side of the worst LBP, 3 cm lateral to the L3 spinous process, with the average used for analysis. |
|                    | 13. Distal sensitivity: PPT is measured as above on the thumb nail of the hand ipsilateral to the side of the worst LBP. |

BDNF, brain derived neurotrophic factor; CRP, C-reactive protein; DASS-21, Depression, Anxiety and Stress Scale; IL-1β, interleukin-1β; IL-6, interleukin-6; LBP, low back pain; PCS, Pain Catastrophising Scale; PPT, pressure pain threshold; PSEQ, Pain Self-Efficacy Questionnaire; SEIFA, Socio-economic Index for area; TNF, tumour necrosis factor; UPWaRD, Understanding persistent Pain Where it ResiDes.
Sample size

G*Power (V.3.0.10, University of Kiel, Germany) was used to calculate the required sample size for estimating the causal effect of baseline sensorimotor activity on chronic LBP. The minimum sufficient adjustment set identified 16 confounding variables that will be controlled for in the causal model. According to the sample size calculation, 111 participants are required to detect an effect size of 0.2 with 80% power, using an alpha level of 0.05, with 16 confounding variables. This calculation is based on detecting a medium effect for a multiple linear regression.

Missing data

Completeness of data obtained from the UPWaRD study at T1 and T2 will be reported as recommended by the STROBE statement. Cases with missing values will be removed from the data set if follow-up rates are higher than 95% at T2. If missing data exceeds 5%, multiple imputation will be performed. The methods used for combining all reported estimates following multiple imputation will be reported (ie, Rubin’s rule). Where data are missing at random (ie, missing randomly, conditional on covariates), estimates based on multiple imputation are unbiased.

Evaluation of demographics and baseline characteristics

Data analysis will be performed in R (The R Foundation for Statistical Computing, a statistical software). Continuous variables will be presented through centrality measures (mean, median), and dispersion (SD and IQR) according to the distribution, and categorical variables through frequencies and percentages.

Statistical analyses

The primary outcome, pain intensity, will be entered into a linear regression model as the continuous dependent variable. Separate multivariable linear regression models for the exposure variables (N80, SEP component, N150, SEP component, L3 map volume and L5 map volume) will be developed. Confounding variables identified by the DAG will be adjusted for in each linear regression model. Linearity assumptions and model fit will be assessed. The regression coefficient and corresponding 95% CIs will be reported and presented in tabular form as recommended by item number 16a of the STROBE statement. The probability threshold for statistical significance will be set at p<0.05.

To further explore a possible causal effect of sensorimotor cortex activity during acute LBP on the development of chronic LBP, pain intensity scores at 6 month follow-up will be dichotomised into chronic LBP (NRS score, ≥1) or recovered LBP (NRS score, 0). Separate logistic regression models will be created to investigate the causal effect of sensorimotor cortex activity measures and chronic LBP. Adjusted and unadjusted ORs with corresponding CIs will be reported once confounders identified by the DAG are entered into the model. To explore the effect of unmeasured confounding a sensitivity analysis will be performed and reported.

The analysis plan will then be repeated to model the causal effect of sensorimotor cortex activity and the secondary outcome, disability. Any deviations from this protocol will be noted in the final manuscript.

DISCUSSION

This protocol details an ‘a-priori’ reported protocol and statistical plan for investigating causal inference using data derived from the UPWaRD prospective cohort study. A directed acyclic graph is presented for the selection of confounding variables, ensuring analytical transparency. Confounding variables entered into a multivariable regression analysis will determine whether sensorimotor cortex activity in the acute stage of LBP has a causal relationship with the development of chronic LBP.

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Contributors SMS and JHMA acquired funding to undertake this research. LJ, W-JC, VB, CC and ML acquired the original data for this research. SMS, LJ and JHMA formulated the methods and designed the protocol and statistical analysis plan. LJ, SMS and AC drafted the protocol and statistical analysis plan. All authors contributed to revisions and approved the final version of the manuscript.

Funding This work was supported by Grant 1059116 from the National Health and Medical Research Council (NHMRC) of Australia. Funding has been applied to achieve the objectives and outcomes described in this protocol and analysis plan. Funding has been spent only on direct research costs as described within the NHMRC funding agreement. SMS receives salary support from the National Health and Medical Research Council of Australia (1105040).

Competing interests None declared.

Patient consent for publication Not required.

Ethics approval All participants provided written informed consent to participate in the study. Ethical approval was obtained from Western Sydney University Human Research Ethics Committee (H10465) and from Neuroscience Research Australia (SSA: 16/002).

Provenance and peer review Not commissioned; peer reviewed for ethical and funding approval prior to submission.

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