Support for Aboriginal health services in reducing harms from alcohol: 2-year service provision outcomes in a cluster randomized trial

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Abstract

Background and aims: There is a higher prevalence of unhealthy alcohol use among Indigenous populations, but there have been few studies of the effectiveness of screening and treatment in primary health care. Over 24 months, we tested whether a model of service-wide support could increase screening and any alcohol treatment.

Design: Cluster-randomized trial with 24-month implementation (12 months active, 12 months maintenance).

Setting: Australian Aboriginal Community Controlled primary care services.

Participants: Twenty-two services (83,032 clients) that use Communicare practice software and see at least 1000 clients annually, randomized to the treatment arm or control arm.

Intervention and comparator: Multi-faceted early support model versus a comparator of waiting-list control (11 services).

Measurements: A record (presence = 1, absence = 0) of: (i) Alcohol Use Disorders Identification Test—Consumption (AUDIT-C) screening (primary outcome), (ii) any-treatment and (iii) brief intervention. We received routinely collected practice data bimonthly over 3 years (1-year baseline, 1-year implementation, 1-year maintenance). Multi-level logistic modelling was used to compare the odds of each outcome before and after implementation.

Findings: The odds of being screened within any 2-month reference period increased in both arms post-implementation, but the increase was nearly eight times greater in early-support services [odds ratio (OR) = 7.95, 95% confidence interval (CI) = 4.04–15.63, P < 0.001]. The change in odds of any treatment in early support was nearly double that of waiting-list controls (OR = 1.89, 95% CI = 1.19–2.98, P = 0.01) but was largely driven by decrease in controls. There was no clear evidence of difference between groups in the change in the odds of provision of brief intervention (OR = 1.95, 95% CI = 0.53–7.17, P = 0.32).
INTRODUCTION

Indigenous peoples that have been colonized, including Australia’s Aboriginal and Torres Strait Islander peoples, are more affected by alcohol-related harms than general populations [1–4]. Harms of colonization, including intergenerational trauma, combined with introduction of mass-produced alcohol, underpin this disparity [5].

Regular screening for unhealthy alcohol use (drinking above recommended guidelines, including alcohol use disorders) is important for timely detection. Cost-effective brief intervention (BI) can then be provided to those with unhealthy drinking who are not dependent [6]. Clients with dependence can be treated in primary care settings, including with pharmacotherapies or referred to specialist services, if appropriate.

Recent systematic reviews [7,8] found only four studies of implementation strategies to increase the uptake of both screening and the full spectrum of treatment for unhealthy alcohol use [7]. While they consistently showed improvement in screening, their impact on treatment provision was variable. None were conducted in Australian or other Indigenous populations.

This report describes the outcomes of a cluster randomized trial to assess whether a service delivery support model, designed for Aboriginal and Torres Strait Islander Community Controlled Health Services (ACCHS), can produce a sustained increase in uptake of screening and appropriate treatment for unhealthy alcohol use. The first 12 months (active phase) of support resulted in a significant increase in screening in any 2-month period over that time [odds ratio (OR) = 5.52, 95% confidence interval (CI) = 4.31–7.07] [9]. However, there was no significant increase in the odds of BI. That analysis did not assess provision of a broader spectrum of treatment.

Previous studies have shown that longer implementation was associated with better outcomes [10]. In this analysis we examine the effects of the support on screening and provision of alcohol treatment over the full 24 months of implementation (active and maintenance phases), and investigate if the effects of provision of BI over 24 months differed from 12-month results. We hypothesized that over the 24 months there would be an increase in the odds of: (i) screening; (ii) provision of any alcohol treatment; and (iii) provision of BI for unhealthy alcohol use.

METHODS

The full study protocol was retrospectively registered (ACTRN12618001892202) and published [9,11]. This paper was prepared using the Consolidated Standards of Reporting Trial (CONSORT) extension for cluster randomized trials [12].

Study design and recruitment

The study is a cluster randomized trial of 22 ACCHS across Australia, with an equal allocation to treatment (early-support) and waiting-list control arms. During recruitment there were approximately 140–143 ACCHS in Australia [13,14]. Of these, 132 were assessed for eligibility. ACCHS were eligible if they: (i) used Communicare practice software; and (ii) provided care for 1000 or more unique clients per year.

Sample size and randomization

Sample size was calculated to detect 13% increase in treatment for unhealthy alcohol use, as this would require the larger sample than for screening alone (Supporting information, Section S1). Eleven ACCHS were recruited per arm [11]. Randomization of ACCHS, stratified by remoteness (based on the road distance to the nearest urban centre; Table 1) [15], was performed by the study statistician (TD) in SAS statistical software, using coded identifiers to ensure blinded allocation.

Implementation strategy

The 24-month support model (Fig. 1) consisted of eight core components (Supporting information, Table S1) designed to aid routine implementation of alcohol screening and appropriate treatment. Support was delivered to the early-support arm services in two 12-month phases: active (components c1–c8) and maintenance (c4–c8). Waiting-list control services operated as normal and had contact with the research team only when providing data. Following both phases in early support, the waiting-list control arm received their support. Services received $AUD100 each as staff-time reimbursement after each provision of data in phases when they were not receiving support.
Study outcomes

We analysed data collected over 3 years (12 months baseline, 24 months implementation). Services extracted de-identified, routinely collected clinical data from their practice software, Communicare, using SQL commands every 2 months. Clients’ records were matched through client IDs [11]. If a client attended in a 2-month period this resulted in an observation, which included age, gender and outcome variables.

The outcomes were a recording (yes/no) of each of the following in each 2-month period:

- screening with the Alcohol Use Disorders Identification Test—Consumption (AUDIT-C); primary outcome (Supporting information, Section S2);
- any treatment: advice/BI or counselling for unhealthy alcohol use or prescription of naltrexone, acamprosate or disulfiram; secondary outcome; and
- BI: advice/BI for unhealthy alcohol use; secondary outcome.

Analysis

We tested whether the support model improved the clients’ odds of each outcome being recorded at least once in any 2-month period. To account for the effects of clustering by service and client, we used multi-level logistic modelling (‘lme4’ package [16] in R statistical software version 4.0.2 [17]). All models focused on testing the following fixed effects:

- ‘condition’: early support arm (condition = 1); waiting-list controls (condition = 0);
- ‘post-implementation’: whether an observation occurred on or after the start of implementation—taken as the date when service champions returned from the national workshop, 31 August 2017 (yes = 1, no = 0); and
- ‘intervention’: effect of support model, given by the interaction between ‘condition’ and ‘post-implementation’. Interaction represents relative change in the odds for the early support arm when compared to the waiting-list controls, post-implementation.
We tested a range of random effects. Model selection is detailed in Supporting information, Sections S3 and S4. We calculated:

- fixed effects and confidence intervals (Wald estimation);
- changes in odds over time for the early support arm (simple slope analysis) using the delta method (‘car’ package [18,19]); and
- adjusted intraclass correlation coefficients (ICC) to describe the proportion of variability explained by differences between clusters (‘performance’ package [20,21]).

We illustrated the fixed effects by plotting adjusted probabilities (‘ggeffects’ and ‘ggplot2’ packages [22,23]).

Missing data

As routinely collected data were used, we had no ability to detect if there were missing outcome data. When comparing demographic characteristics of arms at baseline we used complete-case analysis, as demographic data could be missing.

Aboriginal involvement and consent

Study methods were designed in consultation with two umbrella Aboriginal community-controlled health organizations (Supporting information, Table S2). The participating ACCHS were involved in refining study design. ACCHS’ custodianship of study data was recognized: consent to participation and data release was sought from each ACCHS from authorized representatives and the board; ACCHS were provided with the results and the manuscript for comment before submitting for publication.

Ethics statement

This study received approval from eight ethics committees in Australian states where the participating services were located (Supporting information, Table S2). Three were Aboriginal-specific committees.

RESULTS

Description of sample

The 22 ACCHS contributed 83,032 client records between 29 August 2016 and 15 August 2019 (Fig. 2, Supporting information, Table S6). From January 2019, one waiting-list control service was unable to provide data due to change in practice software. Gender was missing for six clients. Service and client characteristics at baseline (Table 1) and at the end of the study (Supporting information, Table S6) show that sample composition remained broadly unchanged.
FIGURE 2 Flow diagram of participating services (n = number of services). *One service was unable to provide data from January 2019 onwards as they stopped using Communicare to log AUDIT-C results. Duration of follow-up was 24 months.

TABLE 2 Detailed effects

|                  | ICC (%) | OR     | 95% CI    | Log-odds | SE  | P     |
|------------------|---------|--------|-----------|----------|-----|-------|
| Screening        |         |        |           |          |     |       |
| (intercept)      | 52      | 0.02   | (0.01–0.04)| −3.73    | 0.32| 0.00  |
| Post-implementation | 5.09  | (3.01–8.63)| 1.63     | 0.27    | 0.00|
| Condition (early support) | 0.13  | (0.05–0.31)| −2.04    | 0.44   | 0.00|
| Intervention     | 7.95    | (4.04–15.63)| 2.07     | 0.35   | 0.00|
| Brief intervention | 66    | 0.00   | (0–0)     | −9.15    | 0.71| 0.00  |
| Post-implementation | 0.88  | (0.28–2.71)| −0.13    | 0.58   | 0.82|
| Condition (early-support) | 0.79  | (0.15–4.22)| −0.24    | 0.86   | 0.78|
| Intervention     | 1.95    | (0.53–7.17)| 0.67     | 0.66   | 0.32|
| Any treatment    | 33      | 0.00   | (0–0)     | −6.28    | 0.21| 0.00  |
| (intercept)      |         |        |           |          |     |       |
| Post-implementation | 0.59  | (0.41–0.85)| −0.52    | 0.19   | 0.01|
| Condition (early support) | 1.01  | (0.6–1.69)| 0.01     | 0.26   | 0.98|
| Intervention     | 1.89    | (1.19–2.98)| 0.63     | 0.23   | 0.01|

Intervention = effect of the entire 24-month support model given by the interaction between condition and post-implementation time-period. This interaction represents relative change in the odds for the early support arm when compared to the waiting-list control arm, post-implementation. ICC = intracluster correlation coefficient; OR = odds ratio; CI = confidence interval; SE = standard error.
Outcomes

The odds of screening at baseline were lower for the early support arm than waiting-list controls. The odds of recorded BI and any treatment were negligible in both arms. Unadjusted smoothed rates by service and trial arm (Supporting information, Figs S1–S3) and adjusted ICCs (Table 2) demonstrate great variability in effects among the services. Detailed fixed effects results are presented in Table 2.

AUDIT-C screening

After implementation, the odds of screening increased in both arms, but the increase within the early support arm was much larger (simple slope: OR = 40.48, 95% CI = 17.82–91.97). This resulted in early support increase in odds nearly eight times greater (OR = 7.95, 95% CI = 4.04–15.63, P < 0.001) than controls. Probabilities of AUDIT-C screening adjusted for the effects of the support model are shown in Fig. 3.

Any treatment and BI

We found no clear evidence that the support model increased in the odds of having any treatment recorded in the early support arm (simple slope: OR = 1.12, 95% CI = 0.74–1.68). However, odds reduced significantly for waiting-list controls (OR = 0.59, 95% CI = 0.41–0.85, P = 0.01). The reduction resulted in significantly greater odds of any treatment in early support services (OR = 1.89, 95% CI = 1.19–2.98, P = 0.01) than in controls.

The evidence that the model increased the odds of BI in the early support arm post-implementation was inconclusive (simple slope: OR = 1.71, 95% CI = 0.52–5.56), as was the evidence for the early support arm’s increase in odds when compared to waiting-list controls (OR = 1.95, 95% CI = 0.53–7.17, P = 0.32).

However, adjusted probabilities for both any-treatment and BI remained extremely low (Figs 4 and 5).

DISCUSSION

This support model resulted in increased AUDIT-C screening over the 24 months of support and a higher likelihood of screening than during the 12-month active support phase [9]. However, the high variability in improvement between services [9] persisted until the end of implementation. Given this variability, and the lower baseline screening rates in early support arm, these results must be viewed with caution.

Consistent with the results during 12-month active phase [9], we were not able to show clear evidence that the support model improved the BI rates over 24 months.

The significant change in odds of any treatment in early support services when compared to waiting-list controls was driven mainly by a reduction of recorded treatment provision in controls. This result may indicate that the support prevented a drop-off in treatment provision in the early support arm. However, as the probability of treatment remained extremely low, the increase does not translate to a clinically meaningful result.

Alcohol consumption varies within and between Aboriginal and Torres Strait Islander communities [24]. The design of our model employed tailoring to local conditions, iterative support and using data to drive improvement. Longer duration of implementation (24 months) contrasted with past implementation studies where sites received support over 28 weeks on average [13]. These features may have made it possible for many services to implement improvements despite their highly variable context, ranging from settings where alcohol was freely available to others, where individuals had to drink outside the community.

Recommendations for policy, practice and research

The low implementation rates and high variability in effect sizes indicate that further effort is needed to see more consistent improvements in implementation. At service level, this might include periodic...
CONCLUSIONS

We have demonstrated that our flexible model of support, provided over 24 months, can result in improvement of AUDIT-C screening. Variability in outcomes indicate that the model was not uniformly successful in inducing gains in screening. Effects on treatment rates for unhealthy alcohol use are less clear. The longer duration of support facilitated successful delivery of the support model at all sites and appeared to be a factor in improving outcomes. Our study suggests that multi-faceted implementation strategies, which include data-driven tailoring as well as iterative monitoring and improvement processes, are well suited to the needs of ACCHS because of their flexibility and adaptability to local contexts.

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DECLARATION OF INTERESTS

None.

AUTHOR CONTRIBUTIONS

Monika Dzidowska: Conceptualization; formal analysis. KS Kylie Lee: Conceptualization; supervision. James Conigrave: Conceptualization; data curation; methodology. Timothy Dobbins: Formal analysis; supervision. Beth Hummerston: Data curation. Scott Wilson: Conceptualization. Paul Haber: Conceptualization. Dennis Gray: Conceptualization. Katherine Conigrave: Conceptualization; supervision.

TRIAL REGISTRATION

Australian New Zealand Clinical Trials Registry (ACTRN12618001892202): https://anzctr.org.au/Trial/Registration/TrialReview.aspx?ACTRN = 12 618 001 892 202.

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SUPPORTING INFORMATION

Additional supporting information may be found in the online version of the article at the publisher’s website.

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