

## COMMENTARY

## Introducing the Percent, Number, Availability, and Capacity [PNAC] Spatial Approach to Identify Priority Rural Areas Requiring Targeted Health Support in Light of COVID-19: A Commentary and Application

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During December 2019, a new coronavirus (COVID-19) was identified in Wuhan, China.<sup>1</sup> The spread of COVID-19 since identification has quickly emerged as a global issue with the World Health Organization declaring a pandemic on March 11.<sup>2</sup> To date (March 29, 2020), there have been over 716,000 reported cases across 177 countries/regions, with deaths approaching 34,000.<sup>1</sup> Those who are older than 65 years of age generally face the most extreme consequences of contracting COVID-19. Data from the United States have confirmed that up to 30% of people 65 or older contracting COVID require hospitalization, while up to 10% will die.<sup>3</sup>

Ageing populations in rural and remote communities may be especially vulnerable to the COVID-19 pandemic,

and in part, this is due to the availability and capacity of rural health services. Edwards et al<sup>4</sup> investigated the perspectives that rural hospital decision-makers had in relation to service delivery during a pandemic avian influenza scenario. Findings from 17 hospitals concluded that key issues surround (i) a lack of staff, (ii) the need for coordinated health services, and (iii) operational and facility issues. Similarly, Harrod et al<sup>5</sup> confirmed that staffing constraints also hinder rural hospital performance during a time of crisis. Finally, irrespective of hospital capacity issues existing during crisis, research has confirmed that compared to urban communities, rural communities typically face poorer access to health services.<sup>6</sup> This is especially problematic during a pandemic.

For rural communities, responding to the COVID-19 pandemic requires unique service offerings. Consequently, the most pressing issue that geographically large, developed countries with widely dispersed rural localities (for example, the United States, Canada, and Australia) face in tackling the spread and consequences of COVID-19 surrounds the identification of priority areas where unique service offerings and resources should be dedicated. Spatial methods have been effective toward identifying where health service gaps exist,<sup>6,7</sup> and they have also confirmed that compared to urban regions, rural areas often experience further travel times to essential services.<sup>8</sup> Already, in light of COVID-19, seminal work by Boulos and Geraghty<sup>9</sup> synthesized the extent of spatial applications currently used to track incidences of COVID-19. They also clarify where geographic information system (GIS) methods can inform decision-making and suggest that spatial methods can be applied to identify new sites for health services to address COVID-19 incidences. In the absence of funding and time to develop new sites of service delivery in rural localities, spatial methods can identify rural priority areas where unique service offerings (including the better coordination of services<sup>4</sup>) are necessary.

### **Introducing the Percent, Number, Availability and Capacity (PNAC) Approach**

Priority areas have an overrepresentation of a vulnerable population while also limited health and social services, or capacity within services, to address a health issue. Toward the identification of priority areas, spatial approaches have considered both domains separately: (i) population at-risk and (ii) health service availability and capacity. For example, in the context of people with disability, my previous work<sup>8</sup> has clarified priority areas as ones that meet the criteria of having both a statistically significant number and percentage of people with disability. Conversely, Chandak et al<sup>6</sup> have identified areas requiring improved service offerings as ones that have a poor availability of services.

A spatial approach which identifies priority areas fitting the criteria of having a significant geographic prevalence of a vulnerable population while also poor access to health services can be a powerful method to support COVID-19 service planning efforts, and it can support the evidence-based delivery of resources and/or unique service offerings within rural areas. The proposed Percent, Number, Availability and Capacity (PNAC) spatial approach combines traditional single domain approaches and considers a priority area as one which has a signif-

icantly high percentage and number of vulnerable people, while also considering the availability and capacity of health services for priority areas.

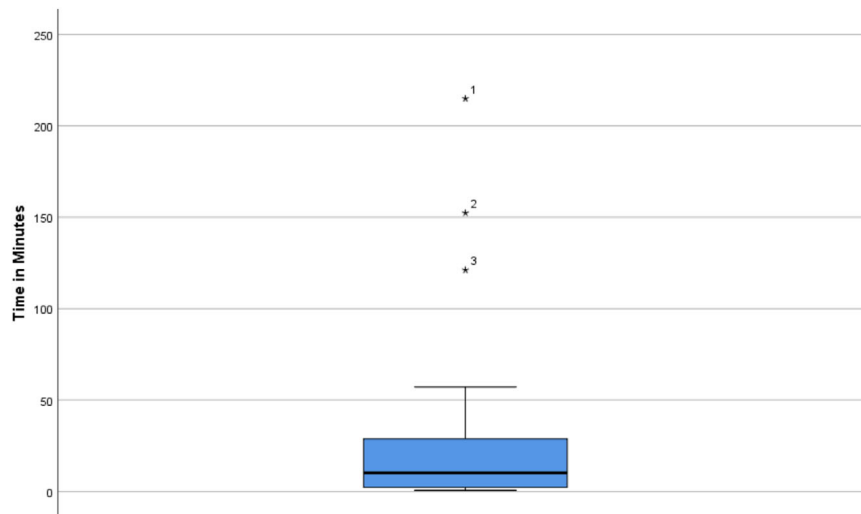
The remainder of this commentary applies the methodology to identify priority rural areas within Queensland, Australia, under the context of COVID-19. Queensland has a large geographic footprint compared to other Australian states,<sup>10</sup> with a considerable proportion of land considered remote or very remote.<sup>11</sup> Additionally, it has a higher proportion of people living outside of the capital city (51.4%) compared to other states in Australia (for example, 35.3% in New South Wales, 24.2% in Victoria, and 22.5% in South Australia).<sup>12</sup> Hence it is an exemplar candidate to illustrate the value the PNAC approach in light of COVID-19.

### **COVID-19 Case Study: Queensland, Australia PNAC Application**

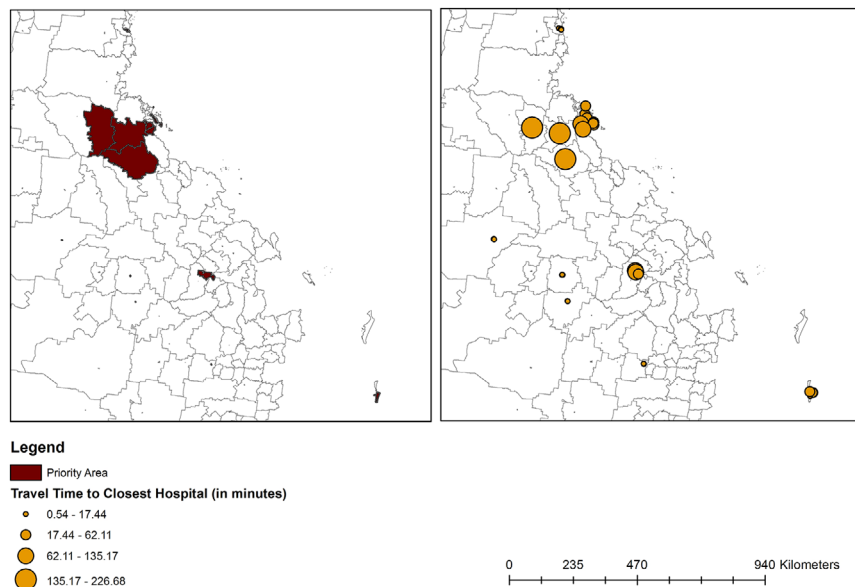
People aged 65 and older were considered the population of interest for this COVID-19 case study, as those 65 and older who contract COVID-19 generally experience the most severe consequences necessitating hospital admission. Data for this case study were identified from a set of public and non-public data being collated for use within the Epidoros-V2 spatial platform.<sup>13</sup> Data were identified from 3 sources: (i) the Australian Bureau of Statistics, (ii) Health Direct, and (iii) The Australian Institute of Health and Welfare. The Australian Bureau of Statistics 2016 Population and Housing census data<sup>14</sup> were used to identify the number and percentage of people aged 65 and older within each Queensland Statistical Area 1 (the second smallest statistical area within the Australian Statistical Geography Standard) classified as remote or very remote. The locations of hospitals within Queensland were derived from Health Direct's 2019 National Health Service Directory.<sup>15</sup> The number of beds per hospital was identified via the Australian Institute of Health and Welfare's My Hospital Data.<sup>16</sup> As a bed range is provided (for example, 0–50 beds, 50–100 beds), the upper bound of each range was used, except for hospitals with greater than 500 beds, where the number 500 was used.

Distinct measures used across the PNAC domains were as follows (with specific measure in brackets): percentage [percentage of people aged 65 and older within each statistical area], number [number of people aged 65 and older within each statistical area], availability [travel time to the closest hospital], and capacity [number of beds within the closest hospital]. All spatial analyses in line with this study were conducted via Esri's ArcMap 10.4.1 (Esri, Redlands, California), while descriptive statistics

**Figure 1** Travel Time From Priority Rural Area to Closest Hospital.



**Figure 2** Priority Areas and Travel Time to the Closest Hospital.



and outliers established via IBM’s SPSS (IBM Corp, Armonk, New York).

In total 36 priority rural regions were identified. The mean travel time to the closest hospital was 26.96 minutes (standard deviation = 44.98 minutes), minimum travel time 0.65 minutes, and maximum travel time 214.86 minutes. In terms of capacity, the closest hospital to all areas had between 0 and 50 beds. In relation to travel time, 3 regions were extreme outliers, with a travel time above the upper outer fence of a produced box plot (see Figure 1 below). In terms of informing a rural response to COVID-19, while the PNAC application identified priority regions for consideration, the 3 rural outliers

based on travel time to the nearest hospital may be initial candidates for unique service offerings as they face exceptionally long travel times to the closest hospital. Two maps in Figure 2 clarify (i) priority areas, and (ii) travel time to the closest hospital.

### Future Considerations and Concluding Remarks

The PNAC approach can be applied to support rural health service planners’ identification of priority rural areas which require targeted health support to address

incidences of COVID-19 amongst people aged 65 and older. Dependent on the extent of data available, measures used across each PNAC domain can be amended. For example, in relation to the Capacity domain, a ratio of people per health service could be used (see Ref. (7) where this measure has been used). While, in relation to Availability, the number of services available within a specified radius (as done in my earlier work<sup>17</sup>), or the geographic location where services are clustered (as done by Chandak et al<sup>6</sup>) may be best.

Rural locations have distinct demographic and health service capacity issues requiring tailored approaches to service delivery. It is expected that future applications of the PNAC method could be of particular benefit to rural service planning in light of COVID-19, as the approach considers both the demographic distribution (for example age) and service capacity issues unique to rural areas. However, it is important that COVID-19 service planning decisions within regions identified via the PNAC method are culturally and contextually appropriate, and they are informed by stakeholders with lived-experience working within identified rural areas.

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