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### Impact of an outreach program for patients with COVID-19 in an integrated healthcare delivery system

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## Impact of an outreach program for patients with COVID-19 in an integrated healthcare delivery system

Running title: Outreach program during COVID-19 pandemic

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## ABSTRACT

**Objectives:** In the first year of the coronavirus disease 2019 (COVID-19) pandemic, health systems implemented programs to manage outpatients with COVID-19. The goal was to expedite patients' referral to acute care and prevent overcrowding of medical centers. We sought to evaluate the impact of such a program, the COVID-19 Home Care Team (CHCT) program.

**Design:** Retrospective cohort

Setting: Kaiser Permanente Northern California

Participants: Adult members before COVID-19 vaccine availability (2/1/2020-

1/31/2021) with positive SARS-CoV-2 tests

**Intervention:** Virtual program to track and treat patients with COVID-19 "CHCT program."

**Outcomes:** Outcomes were 1) COVID-19-related emergency department visit, 2)

COVID-19-related hospitalization, 3) inpatient mortality or 30-day hospice referral.

Measures We estimated the average effect comparing patients who were and were not

treated by CHCT. We estimated propensity scores using an ensemble super learner

(random forest, XGBoost, Generalized Additive Model and Multivariate Adaptive

Regression Splines) and augmented inverse probability weighting.

Results: There were 98,585 patients with COVID-19. The majority were followed by

CHCT (n=80,067, 81.2%). Patients followed by CHCT were older (mean age 43.9 vs

41.6 years, P<0.001) and more comorbid with COPS2 score  $\geq$ 65 (1.7% vs 1.1%,

P<0.001). Unadjusted analyses showed more COVID-19-related emergency

department visits (9.5% vs 8.5%, P<0.001) and hospitalizations (3.9% vs 3.2%,

P<0.001) in patients followed by CHCT but lower inpatient death or 30-day hospice referral (0.3% vs 0.5%, P<0.001). After weighting, there were higher rates of COVID-19-related emergency department visits (estimated intervention effect -0.8%, 95% CI - 1.4%, -0.3%) and hospitalization (-0.5%, 95% CI -0.9%, -0.1%) but lower inpatient mortality or 30-day hospice referral (-0.5%, 95% CI -0.7%, -0.3%) in patients followed by CHCT.

Conclusions: Despite CHCT following older patients with higher comorbidity burden, there appeared to be a protective effect. Patients followed by CHCT were more likely to present to acute care and less likely to die inpatient.

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## Strengths and limitations of this study

- This is a retrospective cohort study that evaluated a real-world program "the COVID-19 Home Care Team program" in an integrated health system. The program reached a large population of diverse patients. While the program was not randomized, we capitalized on a natural control group when demand for the program (case incidence) went above capacity.
- We ensured that patients included in the control group would have been eligible for the intervention had it been available and carefully defined the time period to be prior to vaccine availability.
- We used robust methods to conduct the analysis (propensity scores with ensemble super learner and augmented inverse probability weighting).
- We found a protective effect of the program despite the program following older and more comorbid patients. This program is scalable, successfully offloaded burden from primary care physicians and could be done in other populations for future waves of the pandemic or other respiratory pandemics.
- We were unable to assess patient or provider satisfaction or provider burnout.

#### INTRODUCTION

The coronavirus disease 2019 (COVID-19) pandemic posed many operational challenges for health systems. During each pandemic wave, bed demand exceeded supply, causing strain within the system to accommodate the influx of patients.<sup>1</sup> Units had to adapt to treat patients with acute respiratory failure outside of the intensive care unit, non-urgent outpatient procedures were delayed, and providers were needed to work additional shifts.<sup>2,3</sup> Several studies have documented higher inpatient mortality during inpatient surge periods.<sup>4-7</sup>

Considerable attention has been given to outcomes of patients hospitalized with COVID-19.<sup>6,8-12</sup> However, limited attention has been given to outpatient care of patients with COVID-19, including managing increased volume of secure messages and clinic visits and developing guidelines for triage to the acute care setting. Integrated health systems have the unique capability of managing patients across inpatient and outpatient settings, providing opportunities to intervene prior to their reaching the acute care setting and expediting their arrival to the acute care setting when necessary. Providing care upstream can reduce emergency department overcrowding by managing patients at home or by outpatient-only touchpoints. Additionally, referring patients early to acute care centers that have capacity, even if physically located further away, prevents overcrowding and actually expedites care. Also, having a handle on the number of referrals made to the emergency department allows providers to call in more staff to assist in triaging and managing patients in a timely way.

In the first year of the pandemic, Kaiser Permanente Northern California (KPNC) repurposed resources to accommodate the increased demands on the healthcare

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system to support primary care physicians, manage patients with COVID-19 in the outpatient setting as much as possible and then expedite their referral to an acute care center that was not overcrowded. The novel intervention was COVID-19 Home Care Team (CHCT), which provided a coordinated system to track and treat outpatients who developed COVID-19. We sought to evaluate the impact of the CHCT program on risk of hospitalization and death.

#### **METHODS**

This is a retrospective cohort study. The work was approved, and informed consent was waived by the Kaiser Permanente Northern California Institutional Review Board (#1634347).

#### Setting

Under a mutual exclusivity agreement, 9,500 physicians of The Permanente Medical Group care for >4.4 million Kaiser Foundation Health Plan members at 21 hospital facilities owned by Kaiser Foundation Hospitals.

#### **Study Population**

To establish our base population, we identified all records of members who were ≥18 years old as of 2/1/2020 who had a positive SARS-CoV-2 polymerase chain reaction test ordered between 2/1/2020-1/31/2021. Prior to 3/13/2020, SARS-CoV-2 tests were performed by the Centers for Disease Control and Prevention and state/county health departments, but the results were uploaded into our electronic

health record system and available in KPNC databases. If a patient had multiple positive tests, we examined characteristics and outcomes of the first positive test. The study end date was chosen because it was prior to widespread dissemination of vaccinations for SARS-CoV-2. We excluded patients who were not eligible for CHCT service, such as those who had first positive test during or after a COVID-19-related emergency department visit or hospitalization. We also excluded patients who were hospitalized within 48 hours of their positive test, because contact with the CHCT team took up to 48 hours to initiate.

#### Variable Extraction

We examined patients' electronic health records for demographic and clinical variables, including the following data elements: self-reported race and ethnicity (in order to show the delivery of CHCT across a population),<sup>13</sup> individual comorbid conditions based on diagnosis codes, and neighborhood deprivation index, a composite index ranging from –5 to 5 with more positive values reflecting lower socioeconomic status.<sup>14</sup> We also captured 2 composite indices that are assigned to adults in the KPNC system: a longitudinal comorbidity score (COmorbidity Point Score, version 2 [COPS2]) and an outpatient physiology-based severity of illness score (abbreviated Laboratory-based Acute Physiology Score [abLAPS]). Each month, all adults with a KPNC medical record number are assigned COPS2, which is based on diagnoses accrued in the preceding 12 months with higher scores associated with increasing mortality risk.<sup>15</sup> They are also assigned a monthly abLAPS score, which is based on 14 laboratory tests obtained in the preceding month; higher scores are associated with increased

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physiologic derangement.<sup>13,16</sup> These variables are more fully described in published studies.<sup>15,17-19</sup>

#### Exposure

Prior to the pandemic, KPNC had several existing population health programs for non-pregnant adults with chronic conditions (e.g., diabetes, 20-22 cancer screening, 23,24 cardiovascular disease<sup>25</sup>). In addition, a variety of follow-up processes existed to support primary care providers, such as combinations of in-person and automated outreach for management of hypertensive patients.<sup>26,27</sup> At the start of the pandemic, KPNC leadership utilized this population health management infrastructure to develop a novel outpatient population health program, CHCT, with the goal of increasing frontline primary care provider support by re-purposing non-physician staff, including nurses and nurse practitioners, as well as physicians from departments outside Adult and Family Medicine. After the State of California issued a Shelter in Place order in March 2020. non-emergent surgeries, procedures (e.g., routine cervical cancer screening, colonoscopies), and routine specialty follow-up appointments were deprioritized. KPNC was thus able to repurpose ~450 non-AFM physicians from over 20 specialties as well as non-physician staff to assist in assessing and caring for COVID-19 patients using standardized protocols which provided recommendations for when to triage patients to a higher level of care. All CHCT staff underwent formal training by the medical director (RD). CHCT provided individualized follow-up of patients with early COVID-19 infection including education, assessment, and, if indicated, explicit directions for how to access emergency department care.

Starting in March 2020, patients were electronically enrolled in CHCT when they developed a positive SARS-CoV-2 test. Program staff attempted to reach patients by phone soon after they were informed of their positive test result (usually within 24 hours after a positive result, including weekends). The electronic medical record integrated population care platform allowed CHCT to document outreach attempts so that multiple outreach attempts could be made. Patients active on KPNC's web portal were sent an automated personal message with links to information about available resources and advice on managing their symptoms. Based on standardized protocols developed by front line physicians, patients were escalated to video visits, in-person outpatient visits or acute care (emergency department). During COVID-19 surges, the number of patients with positive tests exceeded CHCT bandwidth, which provided a natural control group of patients who were not followed by CHCT that could be used to compare outcomes. Those who were attempted to be reached at least once were included in the intervention group.

#### Outcome

The primary outcome was COVID-19-related acute care utilization. We examined 1) COVID-19-related emergency department visit or 2) COVID-19-related hospitalization within 2 weeks of positive test. Attribution of hospitalizations to COVID-19 was based on *International Classification of Diseases, Version 10* codes, timing of test orders and non-elective status using a previously published algorithm.<sup>19,28</sup> As a secondary outcome, we also examined the composite outcome of inpatient mortality or hospice referral within 30 days after a positive test as we and others have done in the past.<sup>6,13</sup>

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#### **Statistical Analysis**

We report mean with standard deviation or median with interquartile range for continuous variables. We report number with percent for categorical variables. We compare univariate values with T tests, Wilcoxon rank sum tests or Chi Squared tests, as appropriate.

For each outcome, we estimated the average treatment effect of the CHCT program, comparing patients who were enrolled in CHCT to those who were not using an augmented inverse probability weighting (AIPW)<sup>29</sup> estimator. In AIPW, models are developed for the propensity of treatment and outcome probability as a function of covariates, utilizing all available data. AIPW has the appealing property that only one of the models needs to be specified correctly, known as "doubly robust." We implemented the approach using the AIPW R package<sup>31</sup> that employs the ensemble machine learning approach Super Learner (random forest, XGBoost, Generalized Additive Model and Multivariate Adaptive Regression Splines) to estimate the probability models. The following variables were used in both the propensity score and outcome probability models: age, sex, neighborhood deprivation index, abLAPS, COPS2, obesity, diabetes, hypertension and month of the pandemic. These three comorbidities were chosen because they are highly prevalent in patients with COVID-19 and impact mortality.<sup>30</sup> Month of the pandemic was included because outcomes of patients have improved over time.<sup>6,11,31</sup> We reported the standardized between-group differences in covariates before and after inverse probability weighting based on the propensity score. We also reported the AIPW adjusted outcome prevalence depending on whether patients were followed

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by CHCT or not and the estimated intervention effect with 95% confidence interval, which is the average treatment effect.<sup>32</sup> Threshold for significance was <0.05.

#### RESULTS

There were 98,585 patients with positive tests, the majority of whom were followed by CHCT (n=80,067, 81.2%). Very few patients (n=18) died in the 4 weeks after a positive test without hospitalization, and there was no difference in the percentage who died depending on whether they were followed by CHCT (n=16, 5.1%) or not (n=2, 4.9%, P=0.40). Patients followed by CHCT were older (mean age 43.9 compared to 41.6 years, P<0.001) and more comorbid with COPS2 score  $\geq$ 65 (1.7% vs 1.1%, P<0.001, **Table 1**). They were less likely to be male (45.0% vs 49.3%, P<0.001). Patients studied were diverse with 14.0% Asian, 43.7% Hispanic and 5.8% Black. They were also more likely to have diabetes (11.4% vs 5.7%, P<0.001), obesity (7.6% vs 4.8%, P<0.001) and hypertension (13.7% vs 8.4%, P<0.001). The majority of patients (n=69,150, 70.1%) had positive tests during the third wave of the pandemic (10/15/20-1/31/21).

The overall raw outcome rates were the following: 9.3% had COVID-19-related emergency department visits, 3.8% had COVID-19-related hospitalizations and 0.4% had inpatient death or 30-day hospice referral (**Table 1**). There were more COVID-19-related emergency department visits (9.5% vs 8.5%, P<0.001) and hospitalizations (3.9% vs 3.2%, P<0.001) in patients followed by CHCT but lower inpatient death or 30-day hospice referral (0.3% vs 0.5%, P<0.001).

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**Table 2** shows the standardized differences in characteristics between patients followed by CHCT and not followed by CHCT. After inverse probability weighting, there were no significant differences in standardized differences of characteristics, as expected. After AIPW, there appeared to be a protective effect from the program. There were higher rates of COVID-19-related emergency department visits (estimated intervention effect -0.8%, 95% CI -1.4%, -0.3%) and hospitalizations (-0.5%, 95% CI -0.9%, -0.1%, **Table 3**) and lower inpatient mortality or 30-day hospice referral (-0.5%, 95% CI -0.7%, -0.3%).

The volume of COVID-19-related outpatient visits increased dramatically during the 3 waves of the pandemic. The majority was conducted by CHCT providers (dark portion of stacked bar chart, **Figure 1**), demonstrating an offloading of outpatient work 01.0 to CHCT providers.

#### DISCUSSION

Using population-level data from an integrated health system and robust methods (AIPW), we found a protective effect of an outpatient management program for COVID-19 which was implemented very early in the pandemic to manage patients at home and expedite their referral to acute care when needed. Patients followed by CHCT were older and had higher comorbidity burden, which argues against the possibility of cherry picking. We interpret the results to mean that patients followed by CHCT were more likely to be referred to acute care because of proactive outpatient outreach and monitoring, which likely explains the lower inpatient mortality or 30-day referral to hospice. We showed that there were drastic increases in the volume of

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COVID-19-related outpatient visits during the 3 surge periods and that much of the outpatient COVID-19-related visits were conducted by CHCT providers, demonstrating an offloading of outpatient burden by the program. We believe these findings are important to disseminate as other health systems struggle to manage entire populations of patients through the waves of the current pandemic. The program is scalable and generalizable, as the program itself is delivered completely virtually.

Strain is a term that refers to the time when a clinical care team's ability to provide high quality care is exceeded due to high occupancy, acuity, or turnover. During the pandemic, the Centers for Disease Control and Prevention estimated that if the critical care bed capacity reached 75% nationwide, 12,000 (95% CI=8,623–17,294) excess deaths would occur nationally 2 weeks later.<sup>7</sup> Preventing emergency room overcrowding due to unnecessary visits during periods of high transmission was critical to prevent strain-related deaths.<sup>7,34-36</sup> Programs such as CHCT attempted to streamline the management of patients with COVID-19 in the outpatient setting and to facilitate appropriate emergency room care when patients demonstrate the need for acute care.

It is important to evaluate real-world programs, such as CHCT, that can impact outcomes for a population of patients. Patients with early COVID-19 had tremendous need for both outpatient and inpatient care. One analysis reported patients required 5.6-9 visits in the 30 days after the diagnosis depending on whether they ultimately were hospitalized.<sup>37</sup> Given the rapidity of the onset of the pandemic, this demand for healthcare resources strained our healthcare system. We showed that the majority of COVID-19-related outpatient visits were managed by CHCT providers, repurposed from specialties who had capacity at certain points during the pandemic. Additionally, the

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majority of CHCT encounters were via telephone, which required fewer resources than video or in-person visits. In this study, we demonstrated the management of large demand upswings due to deployment and redeployment of resources which had a positive impact on patient care.

Other programs like CHCT have been implemented and described in the literature. The direction of our results (favoring program benefit) is consistent with the other 3 programs described herein, but the details of the program and outcomes measured differed. First, the Cleveland Clinic Home Monitoring Program included telephone outreach to 3,975 patients after a positive SARS-CoV-2 test to assess patients' symptoms and escalate their care.<sup>38</sup> They performed a matched propensity score analysis and found lower odds of 30-day and 90-day outpatient visits and hospitalization, but not emergency department visits. In our study, which evaluated a program implemented at scale, we report more acute care hospitalization and improved clinical outcomes. Second, the University of Pennsylvania COVID Watch program was a texting intervention whereby 3,488 patients received twice daily texts to inquire about symptoms.<sup>39</sup> They performed a propensity score analysis and found a 64% relative reduction in death for enrolled patients. They found patients were reporting to the emergency department sooner and had more frequent telemedicine encounters. Our program was implemented on a larger scale but found similar reduction in inpatient mortality. Third, the Home Monitoring Program at Providence health system delivered pulse oximeters and thermometers to the home and administered surveys in English/Spanish to monitor symptoms over time.<sup>40</sup> The authors performed propensity scores with inverse probability of treatment weighting. Of 4,358 participants, the

program was associated with more outpatient and emergency department encounters and resulted in high enrollee satisfaction. This study did not report hospitalization rates or rates of clinical outcomes, such as death.<sup>40</sup>

There are several limitations to the current analysis. We did not measure physician or patient satisfaction/experience related to the program. We also did not measure provider burnout created or alleviated by the program. In an effort to reach as many people as possible, the intervention was not randomized but we used the natural control group formed when program demand exceeded bandwidth.

There are several key takeaways and advantages to our study. We showed that patients followed by CHCT were older, more comorbid and from diverse racial backgrounds. While being older and more comorbid could have disfavored the program, we found higher rates of acute care utilization even after adjusting for confounding and lower inpatient mortality or 30-day hospice referral. Additionally, we examined patients who had had at least 1 outreach by CHCT into the intervention group to make it harder to find a statistically significant difference between the groups. We ensured that patients included in the analysis were eligible for CHCT services, i.e., they were not hospitalized at the time of receiving a positive test or within 48 hours because it took CHCT 48 hours to initiate contact. We appropriately examined the period of the pandemic prior to widespread vaccination; including the post-vaccination period would complicate the interpretation of the result given that unvaccinated patients are more likely to be hospitalized for COVID-19 and may be less likely to engage with the program.<sup>33,41</sup> We also demonstrated that the number of patients who died in the 4 weeks after a positive test were few and not different between whether they were followed by CHCT or not.

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We performed a robust analysis using AIPW and adjustment for confounding, including month of the pandemic,<sup>35</sup> and capitalized on the natural control group that occurred when program demand exceeded bandwidth.

In conclusion, we evaluated the KPNC CHCT program, which was developed and implemented early in the pandemic and at scale to manage rises in outpatient care needs related to COVID-19 surges. Despite CHCT following older patients with higher comorbidity burden, there appeared to be a protective effect with higher likelihood of presenting to acute care but lower likelihood of inpatient mortality. We found the program was successful in offloading outpatient clinical care onto repurposed providers during the early part of the pandemic. This type of program is scalable for future waves of the COVID-19 pandemic or future pandemics.

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## AUTHOR'S CONTRIBUTIONS

LCM is the corresponding author and guarantor for the study. LCM, BLL, GJE, RD, CL, and VXL conceptualized the project. LCM, GJE, VXL, BLL, CL contributed to the data curation, formal analysis, and methodology. LCM wrote the first draft of the manuscript, with subsequent editing done by listed co-authors; following this, LCM prepared the final version which is submitted here.

## DATA SHARING STATEMENT

## DATASETS NOT PUBLICLY AVAILABLE

The datasets generated and/or analyzed during the current study are not publicly available due to their being the property of Kaiser Foundation Health Plan, Inc., but are available to interested collaborators in the context of a formal collaboration approved by the Kaiser Permanente Northern California Institutional Review Board for the Protection of Human Subjects.

## Figure 1: Offloading of outpatient COVID-19 related visits to CHCT providers

### Figure 1 Legend

The figure shows the number of COVID-19-related outpatient encounters per week

depending on whether providers were part of COVID-19 Home Care Team (dark grey)

or not (light grey). The bars shown are stacked.

to occurrence only

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## Table 1: Characteristics of patients with COVID-19+ who were and were not followed by COVID-19 Home Care Team

	All patients	Patients followed by CHCT	Patients not followed by CHCT	P valu
	n=98,585	n=80,067	n=18,518	
Age in years, mean (SD)	43.47 (15.54)	43.90 (15.65)	41.61 (14.91)	<0.00
Sex, male (%)	45,183 (45.8%)	36,052 (45.0%)	9,131 (49.3%)	<0.00
Race, n (%)				
Asian	13,796 (14.0%)	10,890 (13.6%)	2,906 (15.7%)	<0.00
Black	5,721 (5.8%)	4,541 (5.7%)	1,180 (6.4%)	
Hispanic	43,082 (43.7%)	35,111 (43.9%)	7,971 (43.0%)	
White	27,646 (28.0%)	22,958 (28.7%)	4,688 (25.3%)	
Other/unknown race <sup>a</sup>	8,340 (8.5%)	6,567 (8.2%)	1,773 (9.6%)	
NDI (median, Q1-Q3) <sup>b</sup>	-0.07	-0.08	-0.03	<0.00
	[-0.61, 0.64]	[-0.63, 0.64]	[-0.53, 0.62]	
COPS2 (median, Q1-Q3) °	10.0	10.0	10.0	<0.00
	[10.0, 10.0]	[10.0, 10.0]	[10.0, 10.0]	~0.00
$COPS2 \ge 65$ , n (%) Comorbidities	l555 (1.6%)	1,350 (1.7%)	205 (1.1%)	<0.00
	10,176 (10,20/)	0 100 (11 40/)	1 054 (5 70/)	~0.00
Diabetes, n (%)	10,176 (10.3%)	9,122 (11.4%)	1,054 (5.7%)	<0.00
Obesity, n (%)	6,988 (7.1%)	6,105 (7.6%)	883 (4.8%)	< 0.00
Hypertension, n (%)	12,505 (12.7%)	10,948 (13.7%)	1,557 (8.4%)	< 0.00
Chronic pulmonary disease, n (%)	7,119 (7.2%)	6,142 (7.7%)	977 (5.3%)	<0.0
Congestive heart failure, n (%)	764 (0.8%)	665 (0.8%)	99 (0.5%)	<0.0
Cancer, n (%)	1,363 (1.4%)	1,173 (1.5%)	190 (1.0%)	<0.0
abLAPS (median, Q1-Q3) d	0.0 [0.0, 0.0]	0.0 [0.0, 0.0]	0.0 [0.0, 0.0]	<0.00
abLAPS ≥ 4, n (%)	3,219 (3.3%)	2,753 (3.4%)	466 (2.5%)	<0.00
Wave, n (%)				<0.00
1 (2/1/20-5/31/20)	2,476 (2.5%)	1,978 (2.5%)	498 (2.7%)	<0.00
2 (6/1/20-10/14/20)	26,959 (27.3%)	21,580 (27.0%)	5,379 (29.1%)	
3 (10/15/20-1/31/21)	69,150 (70.1%)	56,509 (70.6%)	12,641 (68.3%)	
COVID-19-related emergency department visit, n (%)	9,165 (9.3%)	7,587 (9.5%)	1,578 (8.5%)	<0.00
COVID-19-related hospitalization, n (%)	3,703 (3.8%)	3,116 (3.9%)	587 (3.2%)	<0.0
Inpatient death or 30-day hospice referral, n (%)	365 (0.4%)	265 (0.3%)	100 (0.5%)	<0.00

Abbreviations: CHCT=Coronavirus disease 19 Home Care Team

## FOOTNOTE

a Other race includes: American Indian, Alaska Native, Asian Pacific, Native Hawaiian, Pacific Islander, multiracial.

b Neighborhood deprivation index ranges between -5 to +5 with more positive values indicating lower status. See text of Messer et al. (2006) for additional detail.

c The COmorbidity Point Score, version 2 (COPS2) described in Escobar et al. (2013) is a score assigned every month to all adults with a Kaiser Permanente Northern California medical record number. Range is from 0 to 1010; higher scores indicate worse mortality risk. The univariate relationship between the COPS2 and 1-year mortality is as follows: 0-39, 0.3%; 40-64, 5.3%; 65+, 17.2%.

d The Abbreviated Laboratory-based Acute Physiology Score (abLAPS) is a monthly score employing 14 laboratory tests based on the LAPS score described in Escobar et al. (2008). Range is from 0 to 256; higher scores indicate increasing physiologic abnormalities in the preceding month. The univariate relationship between the abLAPS and 30-day mortality is as follows: 0-4, 0.06%; 4-9, 0.18%; 10+, 1.32%.

## Table 2: Selected characteristics of patients who were and were not followed by COVID-19 Home Care Team before and after weighting

	Before Weighting			After Weighting		
-	Patients followed by CHCT n=80,067	Patients not followed by CHCT n=18,518	Standardized Difference	Patients followed by CHCT	Patients not followed by CHCT	Standardized Difference, %
Age, median (IQR)	43 [31, 55]	40 [30, 52]	0.15	42 [31, 55]	42 [31, 55]	0.01
Sex, male (%)	36,052 (45.0%)	9,131 (49.3%)	-0.09	44,889.3 (45.7)	42,202.6 (45.9)	0.00
NDI, median (IQR) <sup>1</sup>	0.08 [-0.63, 0.64]	0.03 [-0.53, 0.62]	-0.04	-0.07 [-0.61, 0.64]	-0.08 [-0.63, 0.63]	0.01
AbLAPS, median (IQR)	0.00 [0.00, 0.00]	0.00 [0.00, 0.00]	0.05	0.00 [0.00, 0.00]	0.00 [0.00, 0.00]	0.02
COPS2, median (IQR)	10.00 [10.00, 10.00]	10.00 [10.00, 10.00]	0.10	10.00 [10.00, 10.00]	10.00 [10.00, 10.00]	0.04
Obesity, n (%)	6,105 (7.6%)	883 (4.8%)	0.13	6991.7 (7.1)	6,693.1 (7.3)	-0.01
Diabetes, n (%)	9,122 (11.4%)	1,054 (5.7%)	0.25	10,172.2 (10.4)	9,822.9 (10.7)	-0.01
Hypertension, n (%)	10,948 (13.7%)	1,557 (8.4%)	0.19	12,520.3 (12.8)	12,463.6 (13.6)	-0.02
March 2020, n (%)	388 (0%)	112 (1%) 🔽 🗸	-0.02	498.3 (0.5)	549.7 (0.6)	-0.01
April 2020, n (%)	719 (1%)	199 (1%)	-0.02	915.4 (0.9)	903.9 (1.0)	-0.01
May 2020, n (%)	871 (1%)	187 (1%)	0.01	1,061.3 (1.1)	1,011.3 (1.1)	0.00
June 2020, n (%)	3,687 (5%)	827 (4%)	0.01	4,538.3 (4.6)	4,310.7 (4.7)	0.00
July 2020, n (%)	8,316 (10%)	2,277 (12%)	-0.06	10,575.5 (10.8)	10,142.7 (11.0)	-0.01
August 2020, n (%)	5,087 (6%)	1,290 (7%)	-0.02	6,348.4 (6.5)	5,979.4 (6.5)	0.00
September 2020, n (%)	3,118 (4%)	703 (4%)	0.01	3,803.9 (3.9)	3,628.0 (3.9)	0.00
October 2020, n (%)	3,611 (5%)	676 (4%)	0.05	4,293.8 (4.4)	4,243.5 (4.6)	-0.01
November 2020, n (%)	10,888 (14%)	2,416 (13%)	0.02	13,227.8 (13.5)	12,639.9 (13.8)	-0.01
December 2020, n (%)	26,592 (33%)	6,059 (33%)	0.01	32,458.2 (33.1)	29,917.7 (32.6)	0.01
January 2021, n (%)	16,790 (21%)	3,772 (20%)	0.01	20,416.6 (20.8)	18,567.2 (20.2)	0.01

Abbreviations: NDI=neighborhood deprivation index, COPS2=Comorbidity Point Score, Version 2, abLAPS=Abbreviated Laboratorybased Acute Physiology Score

## Table 3: Estimated intervention effect of COVID-19 Home Care Team

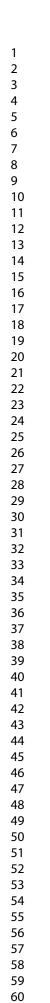
	Adjusted out	come prevalence	Estimated intervention effect (95% Confidence Interval)	
Outcome	СНСТ	No CHCT		
COVID-19-related emergency department visit	9.3%	10.1%	-0.8% (-1.4%,-0.3%)	
COVID-19-related hospitalization	3.8%	4.3%	-0.5% (-0.9%,-0.1%)	
Inpatient death or 30-day hospice referral	0.3%	0.8%	-0.5% (-0.7%,-0.3%)	
The intervention effect being negative indicates a pro-	tective effect of the inte	ervention.		

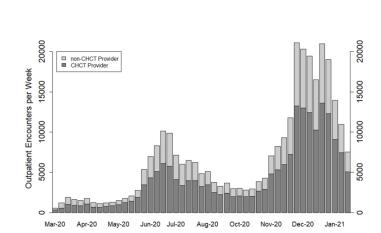
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The figure shows the number of COVID-19-related outpatient encounters per week depending on whether providers were part of COVID-19 Home Care Team (dark grey) or not (light grey). The bars shown are stacked.

338x190mm (300 x 300 DPI)

# **BMJ Open**

#### Evaluation of an outreach program for patients with COVID-19 in an integrated healthcare delivery system: a retrospective cohort study

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## Evaluation of an outreach program for patients with COVID-19 in an integrated healthcare delivery system: a retrospective cohort study

Running title: *Outreach program during COVID-19 pandemic* 

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## ABSTRACT

**Objectives:** In the first year of the coronavirus disease 2019 (COVID-19) pandemic, health systems implemented programs to manage outpatients with COVID-19. The goal was to expedite patients' referral to acute care and prevent overcrowding of medical centers. We sought to evaluate the impact of such a program, the COVID-19 Home Care Team (CHCT) program.

**Design:** Retrospective cohort

Setting: Kaiser Permanente Northern California

Participants: Adult members before COVID-19 vaccine availability (2/1/2020-

1/31/2021) with positive SARS-CoV-2 tests

Intervention: Virtual program to track and treat patients with COVID-19 "CHCT program."

**Outcomes:** Outcomes were 1) COVID-19-related emergency department visit, 2)

COVID-19-related hospitalization, 3) inpatient mortality or 30-day hospice referral.

Measures We estimated the average effect comparing patients who were and were not

treated by CHCT. We estimated propensity scores using an ensemble super learner

(random forest, XGBoost, Generalized Additive Model and Multivariate Adaptive

Regression Splines) and augmented inverse probability weighting.

Results: There were 98,585 patients with COVID-19. The majority were followed by

CHCT (n=80,067, 81.2%). Patients followed by CHCT were older (mean age 43.9 vs

41.6 years, P<0.001) and more comorbid with COPS2 score  $\geq$ 65 (1.7% vs 1.1%,

P<0.001). Unadjusted analyses showed more COVID-19-related emergency

department visits (9.5% vs 8.5%, P<0.001) and hospitalizations (3.9% vs 3.2%,

P<0.001) in patients followed by CHCT but lower inpatient death or 30-day hospice referral (0.3% vs 0.5%, P<0.001). After weighting, there were higher rates of COVID-19-related emergency department visits (estimated intervention effect -0.8%, 95% CI - 1.4%, -0.3%) and hospitalization (-0.5%, 95% CI -0.9%, -0.1%) but lower inpatient mortality or 30-day hospice referral (-0.5%, 95% CI -0.7%, -0.3%) in patients followed by CHCT.

Conclusions: Despite CHCT following older patients with higher comorbidity burden, there appeared to be a protective effect. Patients followed by CHCT were more likely to present to acute care and less likely to die inpatient.

#### 

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## STRENGTHS AND LIMITATIONS OF THIS STUDY

- While the program was not randomized, a natural control group was utilized, when demand for the program (number of cases) went above capacity of the program.
- We ensured that patients included in the control group would have been eligible for the intervention had it been available and carefully defined the time period to be prior to vaccine availability.
- Robust methods were used to conduct the analysis (propensity scores with ensemble super learner and augmented inverse probability weighting).

### INTRODUCTION

The coronavirus disease 2019 (COVID-19) pandemic posed many operational challenges for health systems. During each pandemic wave, bed demand exceeded supply, causing strain within the system to accommodate the influx of patients.[1] Units had to adapt to treat patients with acute respiratory failure outside of the intensive care unit, non-urgent outpatient procedures were delayed, and providers were needed to work additional shifts.[2, 3] Several studies have documented higher inpatient mortality during inpatient surge periods.[4-7]

Considerable attention has been given to outcomes of patients hospitalized with COVID-19.[6, 8-12] However, limited attention has been given to outpatient care of patients with COVID-19, including managing increased volume of secure messages and clinic visits and developing guidelines for triage to the acute care setting. Integrated health systems have the unique capability of managing patients across inpatient and outpatient settings, providing opportunities to intervene prior to their reaching the acute care setting and expediting their arrival to the acute care setting when necessary. Providing care upstream can reduce emergency department overcrowding by managing patients at home or by outpatient-only touchpoints. Additionally, referring patients early to acute care centers that have capacity, even if physically located further away, prevents overcrowding and actually expedites care. Also, having a handle on the number of referrals made to the emergency department allows providers to call in more staff to assist in triaging and managing patients in a timely way.

In the first year of the pandemic, Kaiser Permanente Northern California (KPNC) repurposed resources to accommodate the increased demands on the healthcare

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system to support primary care physicians, manage patients with COVID-19 in the outpatient setting as much as possible and then expedite their referral to an acute care center that was not overcrowded. The novel intervention was COVID-19 Home Care Team (CHCT), which provided a coordinated system to track and treat outpatients who developed COVID-19. We sought to evaluate the impact of the CHCT program on risk of hospitalization and death.

### **METHODS**

This is a retrospective cohort study. The work was approved, and informed consent was waived by the Kaiser Permanente Northern California Institutional Review Board (#1634347). A STROBE checklist is presented in the Supplemental Methods.

#### Setting

Kaiser Permanente Northern California is an integrated healthcare delivery system that cares for 30% of the population in northern California. Under a mutual exclusivity agreement, 9,500 physicians of The Permanente Medical Group provide integrated healthcare for >4.4 million Kaiser Foundation Health Plan members at 21 hospitals owned by Kaiser Foundation Hospitals and 242 medical office buildings.

#### **Study Population**

To establish our base population, we identified all records of members who were ≥18 years old as of 2/1/2020 who had a positive SARS-CoV-2 polymerase chain reaction test ordered between 2/1/2020-1/31/2021. Prior to 3/13/2020, SARS-CoV-2

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tests were performed by the Centers for Disease Control and Prevention and state/county health departments, but the results were uploaded into our electronic health record system and available in KPNC databases. If a patient had multiple positive tests, we examined characteristics and outcomes of the first positive test. The study end date was chosen because it was prior to widespread dissemination of vaccinations for SARS-CoV-2. We excluded patients who were not eligible for CHCT service, such as those who had first positive test during or after a COVID-19-related emergency department visit or hospitalization. We also excluded patients who were hospitalized within 48 hours of their positive test, because contact with the CHCT team took up to 48 hours to initiate.

### Variable Extraction

We examined patients' electronic health records for demographic and clinical variables, including the following data elements: self-reported race and ethnicity (in order to show the delivery of CHCT across a population),[13] individual comorbid conditions based on diagnosis codes, and neighborhood deprivation index, a composite index ranging from –5 to 5 with more positive values reflecting lower socioeconomic status.[14] We also captured 2 composite indices that are assigned to adults in the KPNC system: a longitudinal comorbidity score (COmorbidity Point Score, version 2 [COPS2]) and an outpatient physiology-based severity of illness score (abbreviated Laboratory-based Acute Physiology Score [abLAPS]). Each month, all adults with a KPNC medical record number are assigned COPS2, which is based on diagnoses accrued in the preceding 12 months with higher scores associated with increasing

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mortality risk.[15] They are also assigned a monthly abLAPS score, which is based on 14 laboratory tests obtained in the preceding month; higher scores are associated with increased physiologic derangement.[13, 16] These variables are more fully described in published studies.[15, 17-19]

### Exposure

Prior to the pandemic, KPNC had several existing population health programs for non-pregnant adults with chronic conditions (e.g., diabetes, [20-22] cancer screening, [23, 24] cardiovascular disease [25]). In addition, a variety of follow-up processes existed to support primary care providers, such as combinations of in-person and automated outreach for management of hypertensive patients. [26, 27] At the start of the pandemic, KPNC leadership utilized this population health management infrastructure to develop a novel outpatient population health program, CHCT, with the goal of increasing frontline primary care provider support by re-purposing non-physician staff, including nurses and nurse practitioners, as well as physicians from departments outside Adult and Family Medicine. After the State of California issued a Shelter in Place order in March 2020, non-emergent surgeries, procedures (e.g., routine cervical cancer screening, colonoscopies), and routine specialty follow-up appointments were deprioritized. KPNC was thus able to repurpose ~450 non-AFM physicians from over 20 specialties as well as non-physician staff to assist in assessing and caring for COVID-19 patients using standardized protocols which provided recommendations for when to triage patients to a higher level of care. All CHCT staff underwent formal training by the medical director (RD). CHCT provided individualized follow-up of patients with early

COVID-19 infection including education, assessment, and, if indicated, explicit directions for how to access emergency department care.

CHCT was designed based on Kaiser Permanente Northern California's 20+ years of doing population care. The framework for the program was "Right patients, right clinicians, right tools, right oversight." Starting in March 2020, patients were electronically enrolled in CHCT when they developed a positive SARS-CoV-2 test. Program staff attempted to reach patients by phone soon after they were informed of their positive test result (usually within 24 hours after a positive result, including weekends). The population care platform that was embedded in the electronic medical record allowed CHCT to document outreach attempts so that multiple outreach attempts could be made. The platform allowed staff to easily record outreach attempts, customize follow-up intervals via electronic reminders, and track multiple contact attempts per day. As most members were KP org active for secure messaging, self-care instructions and care resource information were echoed by program staff to eligible patients. Patients received in the mail a kit of supplies, including a pulse oximeter, to prevent them from needing to enter the public domain while contagious. Patients active on KPNC's web portal were sent an automated personal message with links to information about available resources and advice on managing their symptoms. Examples of patients who were prioritized for outreach and follow up had a history of organ transplant, supplemental oxygen at home and active treatment for cancer. Patients at moderate risk were those aged >60, uncontrolled diabetes, cardiac or emergency department within 30 days. Patients with no high risk features were offered an e-visit first. Based on standardized protocols developed by front line physicians, patients were escalated to

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video visits, in-person outpatient visits or acute care (emergency department). During COVID-19 surges, the number of patients with positive tests exceeded CHCT bandwidth, which provided a natural control group of patients who were not followed by CHCT that could be used to compare outcomes. Those who were attempted to be reached at least once were included in the intervention group. Additional information about the program is available in the Supplemental Methods.

#### Outcome

The primary outcome was COVID-19-related acute care utilization. We examined 1) COVID-19-related emergency department visit or 2) COVID-19-related hospitalization within 2 weeks of positive test. Attribution of hospitalizations to COVID-19 was based on *International Classification of Diseases, Version 10* codes, timing of test orders and non-elective status using a previously published algorithm.[19]·[28] As a secondary outcome, we also examined the composite outcome of inpatient mortality or hospice referral within 30 days after a positive test as we and others have done in the past.[6, 13]

#### **Statistical Analysis**

We report mean with standard deviation or median with interquartile range for continuous variables. We report number with percent for categorical variables. We compare univariate values with T tests, Wilcoxon rank sum tests or Chi Squared tests, as appropriate.

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For each outcome, we estimated the average treatment effect of the CHCT program, comparing patients who were enrolled in CHCT to those who were not using an augmented inverse probability weighting (AIPW)[29] estimator. In AIPW, models are developed for the propensity of treatment and outcome probability as a function of covariates, utilizing all available data. AIPW has the appealing property that only one of the models needs to be specified correctly, known as "doubly robust." We implemented the approach using the AIPW R package[30] that employs the ensemble machine learning approach Super Learner (random forest, XGBoost, Generalized Additive Model and Multivariate Adaptive Regression Splines) to estimate the probability models. The following variables were used in both the propensity score and outcome probability models: age, sex, neighborhood deprivation index, abLAPS, COPS2, obesity, diabetes, hypertension and month of the pandemic. These three comorbidities were chosen because they are highly prevalent in patients with COVID-19 and impact mortality.[31] Month of the pandemic was included because outcomes of patients have improved over time.[6, 11, 32] We reported the standardized between-group differences in covariates before and after inverse probability weighting based on the propensity score. We also reported the AIPW adjusted outcome prevalence depending on whether patients were followed by CHCT or not and the estimated intervention effect with 95% confidence interval, which is the average treatment effect. [30] Threshold for significance was <0.05.

#### Patient involvement

Patients were not involved in the design or conduct of the study.

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## RESULTS

There were 98,585 patients with positive tests, the majority of whom were followed by CHCT (n=80.067, 81.2%). Very few patients (n=18) died in the 4 weeks after a positive test without hospitalization, and there was no difference in the percentage who died depending on whether they were followed by CHCT (n=16, 5.1%) or not (n=2, 4.9%, P=0.40). Patients followed by CHCT were older (mean age 43.9 compared to 41.6 years, P<0.001) and more comorbid with COPS2 score  $\geq$ 65 (1.7% vs 1.1%, P<0.001, **Table 1**). They were less likely to be male (45.0% vs 49.3%, P<0.001). Patients studied were diverse with 14.0% Asian, 43.7% Hispanic and 5.8% Black. They were also more likely to have diabetes (11.4% vs 5.7%, P<0.001), obesity (7.6% vs 4.8%, P<0.001) and hypertension (13.7% vs 8.4%, P<0.001). The majority of patients (n=69,150, 70.1%) had positive tests during the third wave of the pandemic (10/15/20-1/31/21). The time from positive test result to first contact with CHCT staff was median 1 day (IQR 0, 4). In the 30 days after positive test, the median time until first ambulatory encounter was 1 day (IQR 1,4) for patients followed by CHCT and 3 days (1,7) for patients not followed by CHCT. The median time between positive test until presentation to acute care (emergency department) was 4 days (IQR 2,7) for those followed by CHCT and 4 days (IQR 1,8) for those not followed by CHCT.

The overall raw outcome rates were the following: 9.3% had COVID-19-related emergency department visits, 3.8% had COVID-19-related hospitalizations and 0.4% had inpatient death or 30-day hospice referral (**Table 1**). There were more COVID-19-related emergency department visits (9.5% vs 8.5%, P<0.001) and hospitalizations

(3.9% vs 3.2%, P<0.001) in patients followed by CHCT but lower inpatient death or 30day hospice referral (0.3% vs 0.5%, P<0.001).

**Table 2** shows the standardized differences in characteristics between patients followed by CHCT and not followed by CHCT. After inverse probability weighting, there were no significant differences in standardized differences of characteristics, as expected. After AIPW, there appeared to be a protective effect from the program. There were higher rates of COVID-19-related emergency department visits (estimated intervention effect -0.8%, 95% CI -1.4%, -0.3%) and hospitalizations (-0.5%, 95% CI -0.9%, -0.1%, **Table 3**) and lower inpatient mortality or 30-day hospice referral (-0.5%, 95% CI -0.7%, -0.3%).

The volume of COVID-19-related outpatient visits increased dramatically during the 3 waves of the pandemic. The majority was conducted by CHCT providers (dark portion of stacked bar chart, **Figure 1**), demonstrating an offloading of outpatient work to CHCT providers. In the 30 days after a positive test, patients followed by CHCT had the following encounter types (71% telephone only, 6% video only, 23% both), and patients not followed by CHCT had the following encounter types (70% telephone only, 16% video only, 12% both). In the 30 days after a positive test, patients followed by CHCT had the following encounter types (70% telephone only, 16% video only, 12% both). In the 30 days after a positive test, patients followed by CHCT had encounters with the following clinicians (58% MD only, 10% RN only, 32% both), and patients not followed by CHCT had encounters with the following clinicians (96% MD only, 1% RN only, 3% both).

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## DISCUSSION

Using population-level data from an integrated health system and robust methods (AIPW), we found a protective effect of an outpatient management program for COVID-19 which was implemented very early in the pandemic to manage patients at home and expedite their referral to acute care when needed. Patients followed by CHCT were older and had higher comorbidity burden, which argues against the possibility of cherry picking. We interpret the results to mean that patients followed by CHCT were more likely to be referred to acute care because of proactive outpatient outreach and monitoring, which likely explains the lower inpatient mortality or 30-day referral to hospice. We showed that there were drastic increases in the volume of COVID-19-related outpatient visits during the 3 surge periods and that much of the outpatient COVID-19-related visits were conducted by CHCT providers, demonstrating an offloading of outpatient burden by the program. We believe these findings are important to disseminate as other health systems struggle to manage entire populations of patients through the waves of the current pandemic. The program is scalable and generalizable, as the program itself is delivered completely virtually.

Strain is a term that refers to the time when a clinical care team's ability to provide high quality care is exceeded due to high occupancy, acuity, or turnover. During the pandemic, the Centers for Disease Control and Prevention estimated that if the critical care bed capacity reached 75% nationwide, 12,000 (95% CI=8,623–17,294) excess deaths would occur nationally 2 weeks later.[7] Preventing emergency room overcrowding due to unnecessary visits during periods of high transmission was critical to prevent strain-related deaths.[7, 33-35] Programs such as CHCT attempted to

streamline the management of patients with COVID-19 in the outpatient setting and to facilitate appropriate emergency room care when patients demonstrate the need for acute care.

It is important to evaluate real-world programs, such as CHCT, that can impact outcomes for a population of patients. Patients with early COVID-19 had tremendous need for both outpatient and inpatient care. One analysis reported patients required 5.6-9 visits in the 30 days after the diagnosis depending on whether they ultimately were hospitalized.[36] Given the rapidity of the onset of the pandemic, this demand for healthcare resources strained our healthcare system. We showed that the majority of COVID-19-related outpatient visits were managed by CHCT providers, repurposed from specialties who had capacity at certain points during the pandemic. Additionally, the majority of CHCT encounters were via telephone, which required fewer resources than video or in-person visits. In this study, we demonstrated the management of large demand upswings due to deployment and redeployment of resources which had a positive impact on patient care.

Other programs like CHCT have been implemented and described in the literature. The direction of our results (favoring program benefit) is consistent with the other 3 programs described herein, but the details of the program and outcomes measured differed. First, the Cleveland Clinic Home Monitoring Program included telephone outreach to 3,975 patients after a positive SARS-CoV-2 test to assess patients' symptoms and escalate their care.[37] They performed a matched propensity score analysis and found lower odds of 30-day and 90-day outpatient visits and hospitalization, but not emergency department visits. In our study, which evaluated a

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program implemented at scale, we report more acute care hospitalization and improved clinical outcomes. Second, the University of Pennsylvania COVID Watch program was a texting intervention whereby 3,488 patients received twice daily texts to inquire about symptoms.[38] They performed a propensity score analysis and found a 64% relative reduction in death for enrolled patients. They found patients were reporting to the emergency department sooner and had more frequent telemedicine encounters. Our program was implemented on a larger scale but found similar reduction in inpatient mortality. Third, the Home Monitoring Program at Providence health system delivered pulse oximeters and thermometers to the home and administered surveys in English/Spanish to monitor symptoms over time.[39] The authors performed propensity scores with inverse probability of treatment weighting. Of 4,358 participants, the program was associated with more outpatient and emergency department encounters and resulted in high enrollee satisfaction. This study did not report hospitalization rates or rates of clinical outcomes, such as death.[39]

There are several limitations to the current analysis. We did not measure physician or patient satisfaction/experience related to the program. We also did not measure provider burnout created or alleviated by the program. In an effort to reach as many people as possible, the intervention was not randomized but we used the natural control group formed when program demand exceeded bandwidth.

There are several key takeaways and advantages to our study. We showed that patients followed by CHCT were older, more comorbid and from diverse racial backgrounds. While being older and more comorbid could have disfavored the program, we found higher rates of acute care utilization even after adjusting for confounding and

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lower inpatient mortality or 30-day hospice referral. Additionally, we examined patients who had had at least 1 outreach by CHCT into the intervention group to make it harder to find a statistically significant difference between the groups. We ensured that patients included in the analysis were eligible for CHCT services, i.e., they were not hospitalized at the time of receiving a positive test or within 48 hours because it took CHCT 48 hours to initiate contact. We appropriately examined the period of the pandemic prior to widespread vaccination; including the post-vaccination period would complicate the interpretation of the result given that unvaccinated patients are more likely to be hospitalized for COVID-19 and may be less likely to engage with the program.[40, 41] We also demonstrated that the number of patients who died in the 4 weeks after a positive test were few and not different between whether they were followed by CHCT or not. We performed a robust analysis using AIPW and adjustment for confounding, including month of the pandemic,[34] and capitalized on the natural control group that occurred when program demand exceeded bandwidth.

In conclusion, we evaluated the KPNC CHCT program, which was developed and implemented early in the pandemic and at scale to manage rises in outpatient care needs related to COVID-19 surges. Despite CHCT following older patients with higher comorbidity burden, there appeared to be a protective effect with higher likelihood of presenting to acute care but lower likelihood of inpatient mortality. We found the program was successful in offloading outpatient clinical care onto repurposed providers during the early part of the pandemic. This type of program is scalable for future waves of the COVID-19 pandemic or future pandemics.

The figure shows the number of COVID-19-related outpatient encounters per week depending on whether providers were part of COVID-19 Home Care Team (dark grey) or not (light grey). The bars shown are stacked.

## Figure 1: Offloading of outpatient COVID-19 related visits to CHCT providers

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# AUTHOR CONTRIBUTIONS

LCM is the corresponding author and guarantor for the study. LCM, BLL, GJE, YFC, RD, CL, and VXL conceptualized the project. LCM, GJE, VXL, BLL, CL, YFC contributed to the data curation, formal analysis, and methodology. KAD provided administrative support and contributed to generating tables and figure. LCM wrote the first draft of the manuscript, with subsequent editing done by listed co-authors; following this, LCM and KAD prepared the final version which is submitted here.

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# DATA SHARING STATEMENT

# DATASETS NOT PUBLICLY AVAILABLE

The datasets generated and/or analyzed during the current study are not publicly available due to their being the property of Kaiser Foundation Health Plan, Inc., but are available to interested collaborators in the context of a formal collaboration approved by the Kaiser Permanente Northern California Institutional Review Board for the Protection of Human Subjects.

## ETHICAL APPROVAL STATEMENT

The work was approved, and informed consent was waived by the Kaiser Permanente

Northern California Institutional Review Board (#1634347).

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# Table 1: Characteristics of patients with COVID-19+ who were and were not followed by COVID-19 Home Care Team

	All patients	Patients followed by CHCT	Patients not followed by CHCT	P va
	n=98,585	n=80,067	n=18,518	
Age in years, mean (SD)	43.47 (15.54)	43.90 (15.65)	41.61 (14.91)	<0.0
Sex, male (%)	45,183 (45.8%)	36,052 (45.0%)	9,131 (49.3%)	<0.0
Race, n (%)				
Asian	13,796 (14.0%)	10,890 (13.6%)	2,906 (15.7%)	<0.0
Black	5,721 (5.8%)	4,541 (5.7%)	1,180 (6.4%)	
Hispanic	43,082 (43.7%)	35,111 (43.9%)	7,971 (43.0%)	
White	27,646 (28.0%)	22,958 (28.7%)	4,688 (25.3%)	
Other/unknown race <sup>a</sup>	8,340 (8.5%)	6,567 (8.2%)	1,773 (9.6%)	
NDI (median, Q1-Q3) <sup>b</sup>	-0.07	-0.08	-0.03	<0.0
	[-0.61, 0.64]	[-0.63, 0.64]	[-0.53, 0.62]	
COPS2 (median, Q1-Q3) °	10.0	10.0	10.0	<0.0
	[10.0, 10.0]	[10.0, 10.0]	[10.0, 10.0]	-0.0
$COPS2 \ge 65, n (\%)$	l555 (1.6%)	1,350 (1.7%)	205 (1.1%)	<0.0
Comorbidities	10,170 (10,00/)	0 400 (44 40/)		-0.0
Diabetes, n (%)	10,176 (10.3%)	9,122 (11.4%)	1,054 (5.7%)	<0.0
Obesity, n (%)	6,988 (7.1%)	6,105 (7.6%)	883 (4.8%)	<0.0
Hypertension, n (%)	12,505 (12.7%)	10,948 (13.7%)	1,557 (8.4%)	<0.0
Chronic pulmonary disease, n (%)	7,119 (7.2%)	6,142 (7.7%)	977 (5.3%)	<0.0
Congestive heart failure, n (%)	764 (0.8%)	665 (0.8%)	99 (0.5%)	<0.0
Cancer, n (%)	1,363 (1.4%)	1,173 (1.5%)	190 (1.0%)	<0.0
abLAPS (median, Q1-Q3) <sup>d</sup>	0.0 [0.0, 0.0]	0.0 [0.0, 0.0]	0.0 [0.0, 0.0]	<0.0
abLAPS ≥ 4, n (%)	3,219 (3.3%)	2,753 (3.4%)	466 (2.5%)	<0.0
Wave, n (%)				<0.0
1 (2/1/20-5/31/20)	2,476 (2.5%)	1,978 (2.5%)	498 (2.7%)	<0.0
2 (6/1/20-10/14/20)	26,959 (27.3%)	21,580 (27.0%)	5,379 (29.1%)	
3 (10/15/20-1/31/21)	69,150 (70.1%)	56,509 (70.6%)	12,641 (68.3%)	
COVID-19-related emergency department visit, n (%)	9,165 (9.3%)	7,587 (9.5%)	1,578 (8.5%)	<0.0
COVID-19-related hospitalization, n (%)	3,703 (3.8%)	3,116 (3.9%)	587 (3.2%)	<0.0
Inpatient death or 30-day hospice referral, n (%)	365 (0.4%)	265 (0.3%)	100 (0.5%)	<0.0

Abbreviations: CHCT=Coronavirus disease 19 Home Care Team

## FOOTNOTE

a Other race includes: American Indian, Alaska Native, Asian Pacific, Native Hawaiian, Pacific Islander, multiracial.

b Neighborhood deprivation index ranges between -5 to +5 with more positive values indicating lower status. See text of Messer et al. (2006) for additional detail.

c The COmorbidity Point Score, version 2 (COPS2) described in Escobar et al. (2013) is a score assigned every month to all adults with a Kaiser Permanente Northern California medical record number. Range is from 0 to 1010; higher scores indicate worse mortality risk. The univariate relationship between the COPS2 and 1-year mortality is as follows: 0-39, 0.3%; 40-64, 5.3%; 65+, 17.2%.

d The Abbreviated Laboratory-based Acute Physiology Score (abLAPS) is a monthly score employing 14 laboratory tests based on the LAPS score described in Escobar et al. (2008). Range is from 0 to 256; higher scores indicate increasing physiologic abnormalities in the preceding month. The univariate relationship between the abLAPS and 30-day mortality is as follows: 0-4, 0.06%; 4-9, 0.18%; 10+, 1.32%.

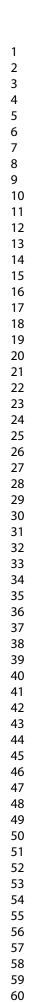
# Table 2: Selected characteristics of patients who were and were not followed by COVID-19 Home Care Team before and after weighting

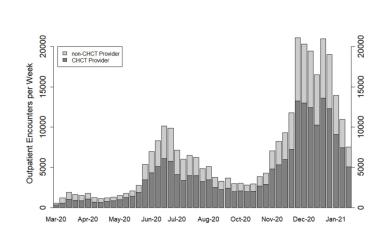
	Be	efore Weighting		A	fter Weighting	
	Patients followed by CHCT n=80,067	Patients not followed by CHCT	Standardized Difference	Patients followed by CHCT	Patients not followed by CHCT	Standardized Difference, %
		n=18,518				
Age, median (IQR)	43 [31, 55]	40 [30, 52]	0.15	42 [31, 55]	42 [31, 55]	0.01
Sex, male (%)	36,052 (45.0%)	9,131 (49.3%)	-0.09	44,889.3 (45.7)	42,202.6 (45.9)	0.00
NDI, median (IQR) <sup>1</sup>	0.08 [-0.63, 0.64]	0.03 [-0.53, 0.62]	-0.04	-0.07 [-0.61, 0.64]	-0.08 [-0.63, 0.63]	0.01
AbLAPS, median (IQR)	0.00 [0.00, 0.00]	0.00 [0.00, 0.00]	0.05	0.00 [0.00, 0.00]	0.00 [0.00, 0.00]	0.02
COPS2, median (IQR)	10.00 [10.00, 10.00]	10.00 [10.00, 10.00]	0.10	10.00 [10.00, 10.00]	10.00 [10.00, 10.00]	0.04
Obesity, n (%)	6,105 (7.6%)	883 (4.8%)	0.13	6991.7 (7.1)	6,693.1 (7.3)	-0.01
Diabetes, n (%)	9,122 (11.4%)	1,054 (5.7%)	0.25	10,172.2 (10.4)	9,822.9 (10.7)	-0.01
Hypertension, n (%)	10,948 (13.7%)	1,557 (8.4%)	0.19	12,520.3 (12.8)	12,463.6 (13.6)	-0.02
March 2020, n (%)	388 (0%)	112 (1%) 🔽 🗸	-0.02	498.3 (0.5)	549.7 (0.6)	-0.01
April 2020, n (%)	719 (1%)	199 (1%)	-0.02	915.4 (0.9)	903.9 (1.0)	-0.01
May 2020, n (%)	871 (1%)	187 (1%)	0.01	1,061.3 (1.1)	1,011.3 (1.1)	0.00
June 2020, n (%)	3,687 (5%)	827 (4%)	0.01	4,538.3 (4.6)	4,310.7 (4.7)	0.00
July 2020, n (%)	8,316 (10%)	2,277 (12%)	-0.06	10,575.5 (10.8)	10,142.7 (11.0)	-0.01
August 2020, n (%)	5,087 (6%)	1,290 (7%)	-0.02	6,348.4 (6.5)	5,979.4 (6.5)	0.00
September 2020, n (%)	3,118 (4%)	703 (4%)	0.01	3,803.9 (3.9)	3,628.0 (3.9)	0.00
October 2020, n (%)	3,611 (5%)	676 (4%)	0.05	4,293.8 (4.4)	4,243.5 (4.6)	-0.01
November 2020, n (%)	10,888 (14%)	2,416 (13%)	0.02	13,227.8 (13.5)	12,639.9 (13.8)	-0.01
December 2020, n (%)	26,592 (33%)	6,059 (33%)	0.01	32,458.2 (33.1)	29,917.7 (32.6)	0.01
January 2021, n (%)	16,790 (21%)	3,772 (20%)	0.01	20,416.6 (20.8)	18,567.2 (20.2)	0.01

Abbreviations: NDI=neighborhood deprivation index, COPS2=Comorbidity Point Score, Version 2, abLAPS=Abbreviated Laboratorybased Acute Physiology Score

## Table 3: Estimated intervention effect of COVID-19 Home Care Team

	Adjusted out	come prevalence	Estimated intervention effec
Outcome	СНСТ	No CHCT	(95% Confidence Interval)
COVID-19-related emergency department visit	9.3%	10.1%	-0.8% (-1.4%, -0.3%)
COVID-19-related hospitalization	3.8%	4.3%	-0.5% (-0.9%, -0.1%)
Inpatient death or 30-day hospice referral	0.3%	0.8%	-0.5% (-0.7%, -0.3%)
The intervention effect being negative indicates a prot		ervention.	





The figure shows the number of COVID-19-related outpatient encounters per week depending on whether providers were part of COVID-19 Home Care Team (dark grey) or not (light grey). The bars shown are stacked.

338x190mm (300 x 300 DPI)

# Evaluation of an outreach program for patients with COVID-19 in an integrated healthcare delivery system: a retrospective cohort study

Myers et al

## Supplemental Methods

This study was approved by KPNC Institutional Review Board (Protocol #1045), who allowed a waiver of informed consent. See STROBE checklist below for details. Outpatient encounters could be telephone, video or in-person clinic visits. We describe patients' clinical characteristics, including self-identified race, which we extracted electronically from the electronic health record. "Other" race included: American Indian, Alaska Native, Asian Pacific, Native Hawaiian, Pacific Islander, multiracial.

# STROBE 2007 (v4) Statement—Checklist of items that should be included in reports of *cohort studies*

Section/Topic	ltem #	Recommendation	Reported on page #
Title and abstract	1	( <i>a</i> ) Indicate the study's design with a commonly used term in the title or the abstract	1
		( <i>b</i> ) Provide in the abstract an informative and balanced summary of what was done and what was found	2, 3
Introduction			
Background/rationale	2	Explain the scientific background and rationale for the investigation being reported	6
Objectives	3	State specific objectives, including any prespecified hypotheses	7
Methods		6	
Study design	4	Present key elements of study design early in the paper	7
Setting	5	Describe the setting, locations, and relevant dates, including periods of recruitment, exposure, follow-up, and data collection	7, 8
Participants	6	( <i>a</i> ) Give the eligibility criteria, and the sources and methods of selection of participants. Describe methods of follow-up	7, 8
		( <i>b</i> ) For matched studies, give matching criteria and number of exposed and unexposed	NA
Variables	7	Clearly define all outcomes, exposures, predictors, potential confounders, and effect modifiers. Give diagnostic criteria, if applicable	8, 9, 10
Data sources/ measurement	8	For each variable of interest, give sources of data and details of methods of assessment (measurement). Describe comparability of	7, 8

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44 45 46 47 48 49		

		assessment methods if there is more than one group	
Bias	9	Describe any efforts to address potential sources of bias	8
Study size	10	Explain how the study size was arrived at	7
Quantitative variables	11	Explain how quantitative variables were handled in the analyses. If applicable, describe which groupings were chosen and why	12
Statistical methods	12	( <i>a</i> ) Describe all statistical methods, including those used to control for confounding	17
	6	( <i>b</i> ) Describe any methods used to examine subgroups and interactions	12
		(c) Explain how missing data were addressed	NA
		( <i>d</i> ) If applicable, explain how loss to follow-up was addressed	NA
		(e) Describe any sensitivity analyses	NA
Results		4.	
Participants	13	(a) Report numbers of individuals at each stage of study—eg numbers potentially eligible, examined for eligibility, confirmed eligible, included in the study, completing follow-up, and analysed	13
		(b) Give reasons for non-participation at each stage	NA
		(c) Consider use of a flow diagram	NA
Descriptive data	14	(a) Give characteristics of study participants (eg demographic, clinical, social) and information on exposures and potential confounders	8
		(b) Indicate number of participants with missing data for each variable of interest	NA
		(c) Summarise follow-up time (eg, average and total amount)	NA

Outcome data	15	Report numbers of outcome events or summary measures over time	13
Main results	16	( <i>a</i> ) Give unadjusted estimates and, if applicable, confounder-adjusted estimates and their precision (eg, 95% confidence interval). Make clear which confounders were adjusted for and why they were included	13
		( <i>b</i> ) Report category boundaries when continuous variables were categorized	NA
O,		(c) If relevant, consider translating estimates of relative risk into absolute risk for a meaningful time period	NA
Other analyses	17	Report other analyses done—eg analyses of subgroups and interactions, and sensitivity analyses	15
Discussion		~	
Key results	18	Summarise key results with reference to study objectives	14
Limitations		4.	
Interpretation	20	Give a cautious overall interpretation of results considering objectives, limitations, multiplicity of analyses, results from similar studies, and other relevant evidence	16
Generalisability	21	Discuss the generalisability (external validity) of the study results	16
Other information		1	
Funding	22	Give the source of funding and the role of the funders for the present study and, if applicable, for the original study on which the present article is based	1