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The Alaska permanent fund dividend increases residential payments for water services†

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Alaska has the lowest rate of access to in-home water services in the United States. At the same time, the state also has the world's oldest Universal Basic Income (UBI) program, and every Alaska resident receives an annual payment through the Alaska Permanent Fund Dividend (PFD) program. In this study, we use a panel dataset of rural Alaska water and sewer utilities in 18 Alaska villages from 2012 to 2016 to explore the impact of the PFD on residential payments. We estimate fixed effects for eight models. Models are developed by grouping villages by low and high variability in payments, enrollment in Alaska Native Claims Settlement Act (ANCSA) regional corporations and Community Development Quota (CDQ) organizations. We find that on average, each utility is missing \$14 710 in customer payments yearly, and have a median residential delinquency rate of 14%. The model with all the villages ($p < 0.01$), ANCSA models ($p < 0.05$), and CDQ models ($p < 0.05$) all show a significant increase in residential payments when the PFD is paid in October. Average residential payments in October are \$3671 to \$10 058 higher than in other months. The increased payments represent 2% to 6% of the total revenue of utilities. We estimate that across rural Alaska the PFD generates between \$734 200 to \$2 011 600 in additional payments for water utilities. These findings suggest that the PFD and other unrestricted cash transfers can play an important role in increasing household water security in rural Alaska and other places with similar problems.

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Environmental significance

The ability of water and wastewater utilities to provide public health and environmental benefits requires generating enough revenue to cover their operations and maintenance (O&M) costs. O&M are typically covered from user revenue, and rising O&M result in higher costs for consumers. Already, across the United States, access to in-home water services is hindered by affordability and the problem is expected to worsen. Universal Basic Income (UBI) may help increase water security by increasing household income. This paper uses the Alaska Permanent Fund Dividend – the longest-running UBI – to explore the impact of UBI on water services. We find that households increase their payments for water services in October, when they receive their PFD.

Introduction

There is fear that Artificial Intelligence (AI) will increase unemployment and decrease incomes.^{1–3} While the impact of AI is still unknown, an increase in unemployment would affect water security. Already, over 10% of American households struggle to pay their water bills,⁴ and rising operating costs are expected to increase the number of households with

unaffordable bills to 35%.⁵ AI has brought attention to Universal Basic Income (UBI),^{6,7} a recurring payment to an entire population with no spending restrictions.⁸ UBIs could improve access to water services since affordability is a function of both price and household disposable income.^{9–11} We explore this relationship by looking at Alaska which has the lowest access to water services in the United States¹² and the world's oldest UBI,¹³ the Alaska Permanent Fund Dividend (PFD).

Started in 1982, the PFD is funded by the investment of oil and gas royalties. Every year, the PFD is paid in October to every Alaska resident. The nominal value of payments has ranged from \$331 in 1984 to \$3284 in 2022.¹⁴ UBIs and other unrestricted cash transfers decrease poverty.^{15–17} Studies in urban areas found that UBIs increase consumption.^{18,19} In a poll of 1023 Americans, 13% of respondents indicated they would use a UBI to pay for utilities.²⁰ In California, households participating in a pilot UBI project spent 10% (\$50 per month) of their UBI on utility bills.²¹ Yet, to our knowledge, no peer-reviewed study has examined the impact of UBIs on payments for

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utilities. This work fills a literature gap by exploring the impact of the PFD on payment for water services in rural Alaska.

Today, only 78% of rural Alaska households have in-home plumbing²² compared to 99.6% nationally,²³ resulting in higher rates of diseases.^{22,24–26} Most villages are off-the-road system, and only accessible by plane or boat, resulting in high transportation costs and costs of living. Village economies are a mix of subsistence and cash generating activities, with a strong informal (unreported trade of services and goods) sector.²⁷ The remote location is due to colonial policies which resulted in the creation of permanent settlements without any planning.^{28,29} Thirty-two villages have no plumbing and must self-haul their drinking water and waste.³⁰ In plumbed villages, access is limited by breakdowns and high operating costs^{31,32} and unaffordable user bills.^{33,34}

This work examines the impact of the PFD on payment for piped water services using a monthly panel dataset of 18 villages in rural Alaska from 2012 to 2016. The PFD is paid in October, and the amount varies yearly. We exploit the dataset's spatial and temporal variation to isolate the PFD's impact on household payments for water services while controlling for the village and time-fixed effects. In our analysis, we also control for villages' enrollment in Community Development Quota (CDQ) groups and with an Alaska Native Claims Settlement Act (ANCSA) corporation.

Methods

ANCSA and CDQs

Alaska Native Claims Settlement Act (ANCSA) corporations are important political and economic entities in Alaska and are often involved in infrastructure development. In 1971, ANCSA conveyed 44 million acres of land to the newly established 12 regional and over 200 village for-profit corporations.³⁵ ANCSA settled long-standing Alaska Native claims settlement issues, allowing the development of the North Slope oil fields. ANCSA corporations distribute dividends to their shareholders, invest in their respective regions, and provide financial assistance to their shareholders.³⁶ In 2021, ANCSA regional corporations had between \$232 million and \$3.9 billion in revenue.³⁷ Nonetheless, the median household income in eight ANCSA regions is lower than the state average, and seven regions have a lower plumbing access (ESI Table 1†).¹² ANCSA corporations publicly release little information about dividends beyond the amount per share.¹⁵

The Western Alaska Community Development Quota (CDQ) program is another important economic actor. Established in 1998, the CDQ program created six non-profit economic development CDQ groups to manage fisheries quota on behalf of eligible villages, previously excluded from the fisheries due to privatization.³⁸ The CDQ groups receive royalties from fishing quotas and must facilitate economic development in member villages.³⁸ While CDQ groups cannot pay out dividends they can and do provide energy subsidies and scholarships and invest in infrastructure development,³⁹ which can increase the disposable income of households. Per the most recent estimate available, the CDQ groups spend between \$1787 and \$80 131 per resident.⁴⁰ As such, there is potentially some direct or indirect effect on water security.

Data

We use a panel constructed from the 2012–2016 accounting records of water and sewer utilities (“water utilities”) and publicly available socioeconomic data. The water utilities are enrolled in a program that helps with administrative tasks, purchases, and maintenance. The utilities must meet specific criteria to join and remain in the program. The names of the program and water utilities are withheld for privacy reasons. We exclude 10 water utilities from the panel. Of these, six are excluded to avoid small samples biasing the results. We define a small sample as fewer than 90 people in the village or a utility having 20 or fewer residential customers. It is likely that in every village every month at least one household (residential customer) is late paying their bill for a variety of reasons. A single missing payment can be significant in small villages (one household is 5% of 20 households) but would be insignificant in larger villages. While there is not a universal definition of what constitutes a small sample and setting thresholds is somewhat arbitrary⁴¹ there is evidence that a sample size of at least 25 is sufficient to identify variability.⁴² Since payments for water services are at the household level, we set a threshold so that all included villages have an average of 25 households at a minimum. We exclude two water utilities that received subsidies of unknown amounts as the subsidies are recorded as residential payments and would bias the results. We also exclude two water utilities with missing socioeconomic data. All dollar values are inflation-adjusted to January 2022. The descriptive statistics are shown in Table 1. Boxplots showing the inter-annual variations are included in the supplementary section (ESI Fig. 1–6†).

Population. The resulting dataset is an unbalanced monthly panel of 18 water utilities from three rural regions in Alaska from January 2012 to December 2016 ($T = 60$ months) with 1020 observations. The descriptive statistics are shown in Table 1 and emphasize the differences between the villages served by the water utilities. While there are considerable spatial variations, most of the socio-demographic variables are constant across time. The villages have between 93 to 876 people and 23 to 197 households, and household size varies between 2.9 to 5.7.¹²

Employment and income. The most accurate village-level wage and employment data is from the Alaska Department of Labor (ADOL). The dataset only includes earned income (wages), which excludes transfer payments and the PFD. ADOL reports annual aggregated wages (at the village level) and quarterly employment data. We decompose the quarterly employment data into monthly by dividing it by three. We decompose annual wages to the monthly level by dividing the wages by the sum of the quarterly employees. We then multiply the result by the number of people employed in a given quarter and divide that by three to obtain the total monthly wages in the village. We divide this number by the number of households to calculate the average monthly household wage.

Again, there are significant differences between villages.⁴³ Employment levels vary between 64% and 86%, unemployment ranges from 14% to 36%, and labor participation rates are



Table 1 Descriptive statistics of water rates, revenues, socio-economic and environmental variables for 18 villages (during 2012–2016) described by the models.

| | Min | Max | Mean | Median | Stand. dev. |
|-------------------------------|------------|-------------|-----------|-----------|-------------|
| Water | | | | | |
| Residential rate (\$/month) | \$69.78 | \$310.09 | \$153.32 | \$142.59 | \$51.29 |
| Customers | 29 | 136 | 74.33 | 71 | 32.39 |
| Residential collection rate | 0.00% | 544.65% | 89.53% | 83.67% | 39.91% |
| Residential payment(\$/month) | \$0 | \$59 321.52 | \$10 028 | \$7309.13 | \$7783.55 |
| Commercial revenue (\$/month) | -\$836.63 | \$12 8300 | \$2257.59 | \$1075.43 | \$5605.22 |
| School revenue (\$/month) | -\$3421.97 | \$26 115.72 | \$4452.99 | \$3951.72 | \$3783.93 |
| Electricity | | | | | |
| Price (\$/kW h) | \$0.53 | \$1.13 | \$0.74 | \$0.74 | \$0.09 |
| PCE (\$/kW h) | \$0.27 | \$0.84 | \$0.48 | \$0.48 | \$0.08 |
| Household costs (\$/month) | \$48.59 | \$896.98 | \$177.1 | \$157.60 | \$71.80 |
| Demographics | | | | | |
| Population | 93 | 876 | 421.57 | 383 | 233.68 |
| Households | 23 | 197 | 95.67 | 98 | 45 |
| Household size | 2.85 | 5.69 | 4.36 | 4.41 | 0.68 |
| Employment | | | | | |
| Labor force participation | 66.13% | 113.36% | 88.20% | 89.05% | 10.42% |
| Employment rate | 63.85% | 85.54% | 76.05% | 75.90% | 5.07% |
| Unemployment rate | 14.45% | 36.14% | 23.94% | 24.09% | 5.07% |
| Income | | | | | |
| Household wage (\$/house) | \$694.02 | \$6242.42 | \$2844.77 | \$2764.84 | \$920.56 |
| Wage per capita | \$191.05 | \$1668.13 | \$674.71 | \$644.70 | \$219.67 |
| PFD/person | \$1057 | \$2410 | \$1631 | \$1427 | \$578 |
| PFD/village | \$102 308 | \$1 771 264 | \$676 618 | \$585 550 | \$442 223 |
| Environmental | | | | | |
| Temperature (°C) | −35.57 | 25.90 | −0.71 | −0.89 | 10.91 |

between 66% and 113%. It is unclear why some rates are above 100%, and this is likely due to data collection issues in rural Alaska.¹⁵ Monthly household wages range from \$694 to \$6242, with a median of \$2845. The annual PFD payments range from \$1057 to \$2410 per person (Fig. 1).¹⁴ The villages receive between \$102 308 and \$1 771 264 in total PFD payments, depending on their population.

Public utilities. The characteristics of the water utilities vary by village and year. All the water utilities charge residential customers a combined rate for water and sewer. Typically, residential and school customers are charged a flat rate, while commercial customers have a metered account. The monthly water and sewer residential rates range from \$70 to \$310, with a median rate of \$134. Water rates remain constant throughout the study and account for 2% to 33% of the average household wage, for an average burden of 6%.

Payments to the water utilities vary monthly and yearly. The utilities record payments in the month they are received, regardless of whether customers pay current, or past bills, or pre-pay. Six utilities recorded \$0 in residential customer payments in one or more months for a total of 10 observations which are included in the analysis to avoid bias in the data.

**Fig. 1** Annual PFD payment during the period of study.

The residential collection rate (amount paid over amount billed) ranges from 0% to 545%, and monthly residential payments range from \$0 to \$59 321 with a mean of \$10 028. Schools and commercial entities are also customers of the water utilities. Monthly school payments range from −\$3422, to \$26 116 and commercial payments range from −\$837 to \$128 300.

The dataset includes residential electricity costs to account for changes in the costs of living in the villages. The electricity data is obtained from the Alaska Energy Authority (AEA) Power Cost Equalization (PCE) database.⁴⁴ The PCE is a State of Alaska program that subsidizes the first 500 kW h consumed by rural households every month.⁴⁴ Subsidized rates range from \$0.27/kW h to \$0.85/kW h with a median of \$0.48/kW h, about four times higher than the national average.⁴⁵ The average monthly residential electricity costs are computed using the following equation:

$$E_c = \frac{Q_{PCE}P_{PCE} + Q_U P_U}{\text{households}} \quad (1)$$

Electricity bills are lagged by a month, with households paying for the previous month's consumption. P_{PCE} is the subsidized price of electricity (\$/kW h), Q_{PCE} is the amount of subsidized electricity (kW h) consumed by all households in the previous month, P_U (\$/kW h) is the unsubsidized price of electricity and Q_U (kW h) is the unsubsidized amount consumed in the previous month. We divide by the number of households to calculate the average per-household cost (E_c). The household electricity costs range from \$49 to \$897, with an average cost of \$177. These costs account for 1.4% to 21% of household wages, with an average burden of 6%.

Weather. Environmental factors, such as temperature, can also impact payments. In cold temperatures, pipes may freeze, halting service delivery. Many utilities do not charge customers during freeze-ups. To our knowledge, freeze-ups are not tracked. Instead, we include temperature data to account for environmental differences. We calculate the monthly temperature data by averaging the daily records for that month.⁴⁶ The temperature ranges from −36 °C to 26 °C with an average temperature of −1 °C, just below freezing.

Empirical approach

The underlying assumption of this study is that several residential customers struggle to pay their water bills. We start by determining whether the annual and monthly billing amounts exceed payments:

$$\text{nonpayment}_{ict} = \text{billing}_{ict} - \text{paid}_{ict} \quad (2)$$

where i denotes each village, c is the customer group (residential, commercial or school) and t is each time period. Billing is the total amount billed to the customer group in that time period in that village and paid is the total amount paid by the customers. A positive nonpayment value indicates that some residential customers are delinquent. We then calculate the nonpayment rate (delinquency rate) using the following formula:

$$\text{Nonpayment rate}_{it} = \left(\frac{\text{nonpayment}_{it}}{\text{billing}_{it}} \times 100 \right) \quad (3)$$

We study the impact of PFD on residential payments for water and sewer bills using a fixed effects model. Since the PFD is paid in October, we expect that month's residential payments to be higher. A basic model of the fixed effects estimates is:

$$y_{it} = \beta_1 x_{it} + a_i + \mu_{it} \quad (4)$$

Hence, the fixed effects models the impact of the PFD on residential payments within each village and at each period. The variable a is the fixed effect of each village – a set of unobserved characteristics within each village that may mediate the impact of independent variables like the PFD. To isolate the impact of the PFD, we must control for the variability within each village (individual heterogeneity). The fixed effect transformation controls for individual heterogeneity by time-demeaning the data through differencing where the average of the model over time is subtracted from each period:

$$y_{it} - \bar{y}_i = \beta_1 (x_{it} - \bar{x}_i) + a_i - a_i + \mu_{it} - \bar{\mu}_i \quad (5)$$

The transformation results in the unobserved characteristics (a) being zeroed out and disappearing from the model. Differencing also zeroes out time-invariant variables such as location and ANCSA corporation.⁴¹ Therefore, the changes in residential payments (dependent variable) must be due to the independent variables remaining in the model after differencing, which vary over time.⁴⁷

We adapt the methodology of Watson *et al.* (2020)⁴⁸ and estimate the coefficients for the following empirical model:

$$\text{Pay}_{it} = \beta_0 + \beta_1 M_{it} + \beta_2 Y_{it} + \beta_3 P_{it} + \beta_4 E_{c,it} + \beta_5 S_{it} + \beta_6 C_{it} + \beta_7 H_{it} + \beta_8 HW_{it} + \beta_9 T_{it} + a_i + \mu_{it} \quad (6)$$

As before, i denotes the villages, and t is the month and year (time period) the observation occurred in. Each utility in the dataset is located in a different village. Pay is the dependent variable and is the total residential payments a utility receives in a time period. Month (M) and year (Y) time effects are included to control for the timing of the PFD and other unexpected events.⁴¹ The residential rate (P) controls for changes in the price of water services which could impact residential demand and payments. The household cost of electricity (E_c) is the average household electricity bill and is a proxy for living costs. The model includes payments for water services for schools (S) and commercial entities (C) as indicators of differences in billing and administration within and between utilities. H is the number of households in the village, since water services are provided per household, and HW is the average household wage. Finally, the model includes monthly average temperature (T) data to account for environmental factors, such as a prolonged cold spell that could decrease household income or cause systems to freeze.

The time-fixed effects estimates are calculated about a base month and year. When October is the base month, the



coefficients of the other months are the difference in payments received compared to October, so we can quantify the additional payments received. The fixed effect estimates are generated using STATA version 17 (ref. 49) and the user-built xtsc⁵⁰ and xtt3 (ref. 51) packages. The Hausman test confirms that fixed effects are appropriate ($p < 0.01$). We use Driscoll-Kraay standard errors since the Breusch-Pagan LM test of independence and modified Wald statistics test are significant ($p < 0.01$), indicating the presence of cross-sectional dependence and heteroskedasticity.⁵²

The key assumption underlying this empirical approach is that the PFD is the only regular but non-monthly payment to occur in October, which to our knowledge is the case. Nonetheless, we want to account for the possibility that there may be other factors, such as ANCSA dividends, or fuel subsidies in CDQ villages, occurring in October. To do this, we estimate fixed effects for eight models. In Model 1, estimates are generated for the entire panel dataset. To explore whether certain villages with higher variability in residential payments in October bias the fixed effects estimates in Model 1, we group the villages by low (Model 2) and high variation (Model 3). To categorize villages by variation in residential payments, we calculate average October z-score for each village and set the low and high threshold as the median z-score (1.56) across all villages.

We also split the villages by ANCSA corporation and enrollment in a CDQ group to further explore how unknowns may impact the October fixed effects estimates. In Models 4–6 we group villages by ANCSA corporation. Each corporation is independent, and they issue dividends and other assistance at different times. If an ANCSA corporation issues a payment we are unaware of in October, these models should capture that. In Model 4, the villages' ANCSA corporation issues its dividend in April, while in Model 5 the other ANCSA corporation's dividends are issued in February, June and/or December, depending on the year. In Model 6, we aggregate four villages which are enrolled with three different ANCSA corporations to obtain a panel large enough to generate fixed effect estimates. The corporations in this model pay their dividends in March, September, or December. We use a similar approach to explore the possible impact of CDQ groups. Fixed effects estimates are generated in Model 7 for the eleven villages not part of CDQ groups, and in Model 8 for the seven CDQ villages. We aggregate the CDQ villages in a single model as there is insufficient data to generate estimates for each CDQ group.

Lastly, we calculate the share of the PFD captured by water utilities by again, using an approach similar to Watson *et al.* (2020)⁴⁸ by estimating fixed effects for the following:

$$\text{Pay}_{it} = \beta_0 + \beta_1 \text{PFDTot}_{it} + \beta_2 \text{PFDTot}_{it}^2 + a_i + \mu_{it} \quad (7)$$

where Pay is the residential payments, and PFDTot is the total PFD disbursed in a village. To calculate PFDTot, we multiply the individual PFD payment by the population of the village. Notably, in some villages up to 45% of households are disconnected from the water utility.³³ To account for this, we generate another set of fixed effects estimates for eqn (7), using the total PFD amount received by 55% of the population.

Results and discussion

PFD and residential payments

It is clear from the estimated equations (Table 2) that the residential payments in October, when the PFD is issued, are significantly higher than in other months (Fig. 2). This finding holds for the aggregated model (model 1) as well as when villages are disaggregated by the size of the October variance from the average month (models 2 and 3), by when the ANCSA corporate dividend is paid (models 4 to 6) and whether the village is part of a CDQ group (models 7 and 8). These are very robust findings in that the October payments are significantly higher than in any other month for all the eight models considered here, with most p -values 0.01.

The increase in residential payments differs across villages and months. The average difference in payments between October and other months varies between \$3671 (Model 2) and \$10 058 (Model 6). Taking the average across all the models, we find that October residential payments are \$6961 higher (per village). Since the median annual revenue of the utilities in the panel is \$168 847, the additional residential payments in October account for 2% to 6% of total annual revenue. To extrapolate these results to the all rural water utilities across Alaska, we multiply the range of estimated excess October residential water payments by 200 (~number of villages) to suggest that the PFD generates between \$734 200 to \$2 011 600 in additional payments every year.

Our findings complement the existing studies on cash transfers and household consumption. They are consistent with Kueng (2015),¹⁹ which found that the PFD increases urban consumption. One difference is that unlike Kueng (2018),¹⁸ we found that the impact of the PFD on water and sewer utilities is restricted to October and does not extend to November and December. Other studies have also found that household consumption increases with cash transfers^{53–55} and households in a UBI pilot program in California spent 10% of the UBI on utility services.²¹ Lower-income households spend a higher proportion of their cash transfer on non-durable goods such as food and on paying off debt^{56–59} which is consistent with our results.

Payments by schools and commercial entities

Recall that we included payments for water services from schools (S) and commercial entities (C) as indicators of differences in billing and administration within and between utilities. We theorized that billing practices impact customer payments and would vary by model. A \$1 increase in school payments is associated with a statistically significant increase in residential payments between \$0.13 (Model 3 and 4, $p < 0.05$ and $p < 0.01$) and \$0.60 (Model 5, $p < 0.01$). School payments are not significant in Models 6 and 8. Commercial payments also positively affect residential payments, however, generally with smaller effects. A \$1 increase in commercial payments is only associated with an increase in residential payments between \$0.04 (Model 1, $p < 0.05$) and \$0.12 (Model 5, $p < 0.1$) in Models 1, 3, 5, and 6 and is not statistically significant in the others. This may be explained by commercial rates being structured differently than rates for residents and schools.



Table 2 Fixed effect estimations with October as base month. Only the significant parameters are shown. See ESI Table 2 for the full table with standard errors. Statistical significance is indicated by superscript asterisks^a

| What impacts residential payments? | Variation | | | ANCSA-corporation dividend | | | CDQ | |
|------------------------------------|------------------|-------------|---------------|----------------------------|--------------------|----------------------|-------------|---------------|
| | (1) All villages | (2) Low | (3) High | (4) April | (5) Feb, June, Dec | (6) March & Sep, Dec | (7) Non-CDQ | (8) CDQ |
| Month (base = Oct) | | | | | | | | |
| January | −7210.29*** | −2503.48** | −12 056.36*** | −5221.93*** | −8832.34*** | −14 799.47*** | −5651.47*** | −11 691.02*** |
| February | −5403.81*** | −2474.21** | −8330.14*** | −5472.74*** | −6706.68*** | −9248.26*** | −4449.82*** | −8425.25*** |
| March | −4645.08*** | −2532.94** | −7371.76*** | −3718.82*** | −7537.74*** | −8447.74** | −4975.17*** | −6350.37** |
| April | −6291.85*** | −3889.64*** | −8994.70*** | −4444.29*** | −7161.39*** | −12 077.63*** | −5185.84*** | −9467.80*** |
| May | −7456.35*** | −4627.87*** | −10 786.18*** | −6352.41*** | −6708.74*** | −12 516.61*** | −4990.62*** | −12 305.77*** |
| June | −6768.12*** | −4143.52*** | −9750.12*** | −6452.40*** | −4688.27*** | −10 178.69*** | −4297.89*** | −11 158.17*** |
| July | −6096.36*** | −4248.98*** | −7557.75*** | −5166.73*** | −4400.36** | −6233.08*** | −3753.04*** | −8107.10*** |
| August | −6540.13*** | −4405.28*** | −8007.79*** | −5451.65*** | −4898.85*** | −7298.68*** | −4212.18*** | −8245.85*** |
| September | −6477.64*** | −4703.00*** | −8007.68*** | −5423.59*** | −6351.39*** | −7062.01** | −4575.36*** | −8089.37*** |
| November | −6742.70*** | −2811.45* | −10 439.02*** | −4589.13*** | −8471.72*** | −11 398.28*** | −5562.15*** | −9427.55*** |
| December | −7033.54*** | −4037.08*** | −10 057.66*** | −5158.61*** | −9582.53*** | −11 378.53*** | −5936.59*** | −9904.33*** |
| Year (base = 2012) | | | | | | | | |
| 2013 | 1128.10** | 1124.33* | 1558.42*** | 633.91 | 3757.74*** | −519.69 | 1812.95*** | 632.77 |
| 2014 | 1778.43*** | 2440.50*** | 1557.99*** | 300.13 | 5585.84*** | 522.74 | 2970.83*** | 982.35 |
| 2015 | 755.19 | 886.97 | 994.74* | 152.08 | 2385.32** | −163.20 | 1050.76* | 1461.10* |
| 2016 | 692.51* | 988.90 | 970.86* | −366.13 | 1566.12 | −383.17 | 1050.82** | 1124.63 |
| Public utilities | | | | | | | | |
| Residential (P) | 29.08** | 28.19 | 35.00*** | 11.55 | 38.46 | 47.75** | 28.74** | 42.98** |
| School pay (S) | 0.30*** | 0.52*** | 0.13** | 0.13*** | 0.60*** | 0.09 | 0.48*** | 0.07 |
| Commercial (C) | 0.04** | 0.02 | 0.07* | −0.03 | 0.12* | 0.07* | 0.05 | 0.05 |
| Electr. cost (E _c) | 3.85 | 0.49 | 3.11 | 19.49*** | 0.09 | −8.35 | −1.17 | 10.36** |
| SocioEcon | | | | | | | | |
| Households (H) | 12.27 | −5.14 | −40.99 | 22.86 | −81.25** | −61.57 | −65.74** | −75.97 |
| Household wage (HW) | −0.01 | 0.84** | −3.12*** | −2.96* | 1.16*** | −5.68*** | 0.56* | −6.50*** |
| Temperature (T) | 18.18 | 82.20* | −53.61 | 111.27** | −107.16 | −193.43 | −37.20 | 31.63 |
| Prob > F | 0.000 | 0.000 | 0.000 | 0.000 | 0.000 | 0.000 | 0.000 | 0.000 |
| Within-R ² | 0.2006 | 0.2029 | 0.3056 | 0.2731 | 0.3431 | 0.2491 | 0.2499 | 0.2941 |
| Observations | 1020 | 510 | 510 | 444 | 336 | 240 | 636 | 384 |
| Villages | 18 | 9 | 9 | 8 | 6 | 4 | 11 | 7 |

^a ***($p < 0.01$), **($p < 0.05$), *($p < 0.1$).

Missing payments and delinquency rates

Since customer payments account for almost the entirety of the utilities' revenue, missing payments hinder the ability of utilities to perform maintenance and repairs.³³ We estimate that the 18 utilities are missing \$264 770 in customer payments every year. This works out to \$14 710 per utility, or approximately one month of revenue (\$14 070) and an overall delinquency rate of 7%. Residential bills account for 97% of the unpaid amount. Since the PFD is associated with an average increase in payments of \$6961, we infer that without the PFD the average unpaid amount would be \$21 670. Hence, the PFD reduces missing payments by just over 30%.

The value of the unpaid bills and the delinquency rates vary across time and by customer group (Fig. 3 and ESI Table 3†). The 2012–2016 delinquency rate is 11% (median of 14%) for residential customers, 0% for commercial entities and 1% for schools, which indicates that over time, commercial and school bills are paid in full but not residential ones. As expected, residential delinquency rates are lowest in October ($p < 0.01$), when the residential delinquency rate averages −36% (overpayment), compared to 15% in the other months. Notably, October is the only month when residential delinquency rates are negative. Residential delinquency rates peak November through January and in May (average of 19%). Cash-generating activities vary



Monthly Variation in Residential Payments



Fig. 2 We normalize the monthly revenue from residential payments by dividing each month's average residential revenue by the average residential revenue in October for each model (a)–(h). This results in October always being 1. To convey the differences in residential payments between the models we include the total value of residential payments in October in yellow.



Fig. 3 Monthly variation in unpaid bills for water services by type of customer. The yellow dots are the month of October.

seasonally,²⁷ which may explain why residential payments vary seasonally. The commercial delinquency rate in October is like the rate in January, April and June ($p > 0.1$) and it is slightly lower

than in the other months ($p < 0.1$). While we have no explanation for these variations, we hypothesize that they are linked to the businesses cash flow. The schools' October delinquency rate is not significantly different than in other months ($p > 0.1$).

Residential water rates

The water utilities must balance setting a price their customers are able to pay and generating enough revenue to cover their operating costs. This can be an impossible balancing act. Like other small utilities in the countries, these utilities have a small customer base over which to distribute their high costs.^{60–62} There are few alternatives to rate revenue. The State of Alaska only sporadically provides funding for operations and maintenance costs and the Indian Health Service is the only federal agency permitted to fund operations costs, but Congress has never appropriated the funding.³³

Nonetheless, residential water rates only have a moderate effect on residential payments in our models. Water rates are not significant in Models 2, 4, and 5. In the other models, a \$1 increase in residential rates is associated with an increase in village payments of \$29–\$48 ($p < 0.5$ to $p < 0.01$). To better investigate the impact of residential rates, we estimate the random effects for the models and report the results in ESI Table 4.† Random effects examine drivers of differences between villages.⁴¹ The random effect estimates for



residential rates are significant in all the models, indicating that their moderate effect in the fixed effects models may be a statistical anomaly due to the rates remaining relatively constant across the years (ESI Fig. 4†).

Cost of electricity

Electricity costs are included as a proxy for costs of living as there is no other village-level data for this. In our models, electricity is only significant in models 4 and 8. A \$1 increase in household electricity expenditure is associated with an approximate increase of \$19 in residential payments in Models 4 ($p < 0.01$) and \$10 in Model 8 ($p < 0.5$). Electricity costs may be a poor indicator of changes in costs of living, or they may not change significantly in the period of study. In the random effects model, the electricity costs account for differences in residential payments in three models ($p < 0.05$).

Water and electricity burden

We hypothesize that the effect of the PFD is linked to the affordability of water bills and costs of living. In rural Alaska, costs are high. In 2020, the average cost of electricity was \$0.60/kW h,⁶³ compared to a national average of 0.14/kW h.⁴⁵ In our panel, households spend between 5% to 50% of their wages just on water, sewer, and electricity, for an average burden of 12%, almost twice the national average for household expenditure on all utilities.⁶⁴ In this context, it is unsurprising that even the higher-income villages experience an increase in payments with cash transfers, highlighting the extent of the affordability issue across rural Alaska. These problems are expected to worsen as the operating costs of utilities are increasing.^{32,65}

Household income and PFD

The PFD represents a substantial proportion of household income. From 2012 to 2016, the PFD accounted for 11.6% of the

average annual household income of the 18 villages in the panel. During that same period, the PFD accounted for 4.3% of the average annual household income in Alaska (Fig. 4). The difference is likely because although the villages have different economies, all have high costs of living and few formal employment opportunities.^{27,66,67} In addition, cash-generating activities often conflict with subsistence, of which it is hard to overstate the importance.^{28,68–70}

Wages

The impact of wages (earned income) varies by model. Household wages are not statistically significant in the combined model and have a contrasting impact in the other models. In Models 3, 5, and 7, a \$1 increase in household wage results in an increase in residential payments between \$0.6 (Model 7, $p < 0.1$) and \$1.2 (Model 5, $p < 0.01$). In models 3, 4, 6, and 8, a \$1 increase in household wages is associated with a decrease in residential payments between \$3.1 (Model 3, $p < 0.01$) and \$6.5 (model 8, $p < 0.01$). Again, we refer to the random effects estimate (ESI Table 3†) to better understand the role of household wages. Household wages are only significant in four models and again have opposite signs.

The negative effect between household wages and residential payments in some models is somewhat surprising. These results may be due to using wage data, which excludes unearned income (e.g. PFD or social security), or to the assumptions used to decompose the wages to the monthly level. Another reason may be seasonality, although we would expect the time-fixed effects to remove time trends. Nonetheless, we notice that household wages are over \$1200 higher in the summer (ESI Table 5†), and there is no significant increase in residential payments. Summer is also an important time for subsistence. We hypothesize that households prioritize subsistence activities, which are intrinsic both to Indigenous ways of life^{28,69,70} and food security.⁶⁸ Engaging in subsistence



Fig. 4 PFD as a percentage of annual household income in Alaska and in the dataset (18 villages).



Table 3 Estimates of fixed effect of the share of PFD paid to water utilities. Statistical significance is indicated by superscript asterisks^a

| What share of each PFD dollar is paid to water utilities? | % of population connected to the water utility | |
|---|--|------------------------|
| | 100% | 55% |
| PFD _{Tot} | 0.014*** | 0.026*** |
| PFD _{Tot} ² | -4.63×10^{-9} | -1.53×10^{-8} |
| Prob > F | 0.000 | 0.000 |
| Within-R ² | 0.1486 | 0.1486 |

^a ***($p < 0.01$), **($p < 0.05$), *($p < 0.1$).

requires purchasing equipment and fuel^{71,72} and costs vary by village. For example, CDQ villages (Model 8) are coastal villages with a tradition of traveling onto the ocean to fishing,^{73,74} which may entail higher costs compared to inland villages.

Share of the PFD

We estimate that residential customers spend 1.4–2.6% of their PFD on water services ($p < 0.01$, Table 3). The average total PFD amount received in each village is \$676 618, with \$9472 (1.4% of PFD payments, Table 3) of that going to the water utility. If only 55% of the population is connected, they would receive \$372 140 in PFD payments. We find that on average a water utility would receive \$9676 (2.6% of PFD, Table 3 payments). Both these figures are slightly higher than the previously estimated average increase in payments of \$6961 but are within the previously estimated range of \$3671–\$10 058. The disparity is likely due to the different approaches used.

Informal economy

These findings are particularly relevant to communities with a significant informal economy and limited formal employment opportunities. A benefit of UBIs like the PFD is that they reach everyone, including workers in the informal economy.⁷⁵ Broadly defined, an informal economy is the unreported trade of goods and services that may or may not involve cash exchanges.^{27,76} The informal economy accounts for a large but unquantified proportion of the rural Alaska economy,^{27,77,78} 30% of the global GDP⁷⁹ and 8% of the American GDP.⁷⁶ The ubiquity of the informal economy around the world makes our findings relevant to communities outside of Alaska.

Limitations

It is important to consider the limitations of this study. The statistical analysis does not establish a causal relationship between the PFD and increased residential payments. This could be a spurious relationship resulting from their interaction with unaccounted variables.⁸⁰ Another limitation stems from the lack of household level data. We disaggregate village-level data which may obscure differences in household behavior. Given these data limitations, we cannot

control for the impact of outlier households on our results. Taken together, these limitations may explain the inconsistent impact of wages and electricity costs on residential payments. The impact of household wages may be underestimated as the data does not include unearned income and transfer payments can account for a significant proportion of household income.¹⁵ Finally, we lack data on the number of households who are customers of the water utility. In addition, the data does not allow us to distinguish between back payments, payments for current bills and pre-payments. Given this, we are unable to estimate some useful metrics, such as how many households are delinquent. These results would also be strengthened by a study of other unrestricted cash transfers, like ANCSA dividends and tax refunds.

Conclusion

Our findings suggest that the PFD plays an important role in household water security and helping utilities cover their costs in rural Alaska. We find that, on average, each water utility is missing \$14 710 annually in payments from all customers. The median residential delinquency rate is 14%. Residential payments increase significantly in October, when the PFD is received by households. October residential payments are on average \$6961 higher than in other months. In the panel, the increased payments represent between 2% and 6% of the total revenue of utilities. We estimate that annually the PFD results in \$734 200 to \$2 011 600 in additional payments across rural Alaska and decreases missing revenue by almost 40%.

The PFD appears to play an important role in both ensuring household access to water services and helping utilities recoup their costs. These findings are contextualized within the discussion of affordability. We find that on average the households in the panel dataset spend 12% of their income on water and electricity, which is almost twice the national average for household expenditure on all utilities. In some villages, we estimate that water and electricity accounts for half of the average household income. This situation is exacerbated by a limited number of cash-generating opportunities. This issue of affordability is not unique to rural Alaska. Across the country, an increasing number of households struggle to pay their utility bills.

In addition to the growing affordability crisis, there are increasing fears of mass unemployment due to AI. Based on the results of this study, we find that the PFD and other UBIs and unrestricted cash transfers may be effective tools to enhance household access to basic services, increasing water security.

Author contributions

Barbara Johnson: conceptualization, methodology, software, data curation, visualization, writing–original draft preparation, writing–review and editing. Dr Allen Molina: data curation, software, visualization, writing–original draft preparation. Dr Mark Herrmann: software, writing–original draft preparation, writing–review and editing. Dr Srijan Aggarwal: methodology,



data curation, visualization, supervision, writing-original draft preparation, writing-review and editing.

Conflicts of interest

There are no known conflicts to declare.

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