Decoding the Divide: Analyzing Disparities in Broadband Plans Offered by Major US ISPs

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ABSTRACT
Digital equity in Internet access is often measured along three axes: availability, affordability, and adoption. Most prior work focuses on availability; the other two aspects have received less attention. In this paper, we study broadband affordability in the US by focusing on the nature of broadband plans offered by major ISPs. To this end, we develop a broadband plan querying tool (BQT) that obtains broadband plans (upload/download speed and price) offered by seven major wireline US ISPs for any street address in the US. We then use this tool to curate a dataset, querying broadband plans for over 837 k street addresses in thirty cities for these ISPs. We use a plan’s carriage value, defined as the Mbps of a user’s traffic that an ISP carries for one dollar, to compare plans. Our analysis provides us with the following new insights: (1) ISP plans vary inter-city. Specifically, up to 60% of the census block groups in a city can receive low carriage value plans from an ISP; (2) ISP plans intra-city are spatially clustered, and the carriage value can vary as much as 600% within a city; (3) Cable-based ISPs offer up to 30% higher carriage value to users when they are competing with fiber-based ISPs in a block group compared to when they are operating alone or in conjunction with a DSL-based ISP; and (4) Fiber deployments, which have better carriage values, are associated with higher average income block groups. While we hope our tool, dataset, and analysis in their current form are helpful for policymakers at different levels (city, county, state), they are only a small step toward quantifying digital inequity. We conclude with recommendations to further advance our understanding of broadband affordability.

CCS CONCEPTS
• Networks → Network measurement. Public Internet; • Social and professional topics → Government technology policy; Broadband access; Governmental regulations;

KEYWORDS
Public Internet, Broadband Access, Broadband Pricing.

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1 INTRODUCTION
The National Digital Inclusion Alliance (NDIA) in the US defines digital equity as “a condition in which all individuals and communities have the information technology capacity needed for full participation in our society, democracy, and economy” [48]. As modern life has moved increasingly online, high-quality Internet access has become a key component of digital equity. The Covid-19 pandemic, and the post-pandemic “new normal” of remote interaction, have drastically changed the need for home Internet access; work-from-home, online/remote schooling, telemedicine, and other networked applications have become increasingly indispensable. As a result, individuals without home access to highly reliable, high-speed broadband are severely disadvantaged [41].

Policymakers cannot take effective corrective actions, such as offering subsidies [38], regulating rates [7], and funding access infrastructure [49], without understanding the true characteristics of digital inequity. Digital equity, especially in the context of Internet access, is often measured along three axes: availability, affordability, and adoption [54]. Many past efforts [36, 51, 52], including ones in our research community, have focused on measuring availability. Researchers have disaggregated availability into coverage and quality. Here, coverage answers whether broadband access is available in a geographical region, while quality answers questions related to access type (e.g., cable, fiber, DSL), and upload/download speed. Researchers and policymakers use publicly-available datasets, such as the FCC’s Form 477 [28], Measuring Broadband America (MBA) [39], and Measurement Lab (M-Lab) speed test [46], as well as proprietary ones, such as Ookla’s speed test [50], to characterize Internet connectivity. More recently, as part of the Broadband Equity, Access, and Deployment (BEAD) program, the US Congress directed the FCC to develop an accurate map of fixed broadband availability across the US. Though it is still a work in progress, when completed, the FCC National Broadband Map [17] will provide information regarding broadband availability (i.e., provider, access type, maximum upload/download speed) at the granularity of street addresses.

Whereas the existing datasets in the US broadband sector, including the most recent FCC National Broadband Map, measure availability, affordability has received less attention. To answer any
question related to broadband affordability, extracting the “cost of broadband connectivity”, i.e., the nature of the “deal” a user is getting, at fine-grained geographical granularity, is important. Using cost data, one can answer policy questions such as (1) what pricing policies do ISPs employ to users in different regions (i.e., neighborhoods, cities, states)?; (2) where, within a region, are different types of deals offered by ISPs?; (3) how does the (lack of) competition among ISPs affect broadband prices in a region?; and (4) how do socioeconomic and demographic factors correlate with broadband prices?

Most previous studies have either focused on manually querying ISP websites [42, 47] or self-reporting from ISPs [40], and, at best, they scratch the surface of questions (1) and (2). A more recent study by a team of investigative journalists curated broadband availability and cost data at street-level granularity for four major ISPs across 43 cities.\(^1\) However, among other limitations, this study did not analyze the broadband plans for major cable-based ISPs (e.g., Cox), and thus, it could not fully answer questions (3) and (4).

Our goal is to curate a new dataset that enables a better understanding of broadband affordability in the US, addressing the limitations of prior related efforts. To this end, we present the design and implementation of a new broadband plan querying tool (BQT). BQT takes a street-level address as input and returns the available broadband plans offered by major ISPs at that address. Here the plans entail the maximum upload speeds, download speeds, and corresponding prices in US dollars; typically, multiple plans are available to each residential address. BQT automates mimicking the behavior of a real user interacting with an ISP’s website to query available broadband plans for a given street address. It addresses various challenges to offer a high hit rate, i.e., the number of street addresses it can successfully query for an ISP and the number of major ISPs it can query.

We use BQT to curate our broadband plans dataset while ensuring our data collection effort does not overwhelm ISP websites. Specifically, we collect and analyze plan data in thirty US cities with diverse populations, population density, and median income. We identify seven major ISPs that reach 89% of the total census blocks in the US [17]. For each (ISP, city) pair, we sample a subset of residential addresses extracted from a dataset provided by Zillow [3]. We feed these addresses to BQT to curate the desired broadband affordability dataset.

We use this dataset to answer multiple policy questions about broadband affordability in the US. Specifically, we use the metric \textit{carriage value} to characterize broadband plans.\(^2\) This metric quantifies the amount of user Internet traffic (in megabits) that an ISP can carry per second, per dollar spent on a monthly broadband plan. For example, the carriage value for a broadband plan with a download speed of 100 Mbps at $50/month is 2 Mbps/$. Intuitively, the higher the carriage value, the better the deal the user receives for their broadband subscription, and vice versa. We use this metric to study the quality of “deals” ISPs offer within and between different cities. From an end user’s perspective, we explore how this metric varies across different ISPs active in a region, how the nature of the deal correlates with various demographic and socioeconomic factors, and the state of competition among ISPs locally. By using this metric, our paper and its findings can contribute directly to the ongoing discussion currently active in the US on broadband pricing, ensuring consistency and relevance.

In summary, our work offers three major contributions:

- **Broadband plan querying tool (Section 3).** We present the design and implementation of a broadband plan querying tool that reliably queries the websites of seven major ISPs, mimicking a real user, to extract the available broadband plans for a given street address.

- **Broadband plans dataset (Section 4).** We present our methodology to curate a broadband plans dataset by querying 837 k unique addresses (1.2 M plans) across 30 cities (18 k census block groups) and seven major ISPs in the US. Our emphasis is metropolitan/urban areas across the US. However, our work can be expanded to include small towns and rural areas.

- **Characterization of broadband plans (Section 5).** We conduct a multi-dimensional analysis to study the intra- and inter-city distribution of broadband plans (i.e., carriage value) for each ISP and how these plans are affected by competition among ISPs and various demographic and socioeconomic factors. Our analysis offers the following key insights: (1) ISP plans vary by city, i.e., the fraction of census block groups that receive high (and low) carriage value plans are variable across cities.\(^3\) (2) ISP plans within a city are spatially clustered, and the carriage value can vary as much as 600% within a city. (3) Cable-based ISPs deliver up to 30% greater carriage value to users when in competition with fiber-based ISPs within a block group, as opposed to when they operate independently or alongside a DSL-based ISP. (4) Block groups with higher average income tend to be associated with higher fiber deployments, which offer superior carriage values. However, racial composition and population density, when considered independently of average income, do not correlate with differences in fiber deployment.

We view this work as an important step towards understanding broadband affordability in the US at scale. We note that broadband affordability is multifaceted, with numerous factors to consider. While our analysis provides valuable insight, it only scratches the surface of what policymakers must address when assessing broadband affordability. The evaluation of broadband affordability in a specific region or for a particular population may require consideration of additional factors beyond the scope of this paper. To enable other researchers and policymakers to advance our understanding of this critical topic, we will make our tool and a privacy-preserving version of our dataset publicly available. We conclude this study with recommendations for different stakeholders to further improve the understanding of broadband affordability.

**Ethical concerns.** Please refer to Section 4.2 for a discussion of how we address ethical concerns regarding our data-collection tool and methodology.

## 2 BACKGROUND & MOTIVATION

**Broadband providers in the US.** Thousands of US ISPs offer broadband connectivity, reaching approximately a hundred million

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\(^{1}\)Our team provided technical assistance for this investigative reporting.

\(^{2}\)A paper recently proposed this metric in the legal literature [47] that the White House referred to in announcing a new Executive Order [8] citing a call to arms to address the lack of competition among broadband service providers in the US.

\(^{3}\)Xfinity emerges as an exception as its plans are invariant across the specific cities we study in this work.
residences. Most of these ISPs operate locally and have a fairly small footprint [27, 33, 34]. This paper considers seven major ISPs, each serving at least one million residences. Together they reach 89% of the total census block groups in the US. We can divide these ISPs into two broad categories: DSL/fiber-based and cable-based providers. Our work, like others [6], confirms that these ISPs either operate as a monopoly or duopoly, i.e., at max, only two major ISPs compete with each other in a census block group. Also, ISPs of the same type do not compete with each other: DSL/fiber-based ISPs do not compete with each other, and cable-based ISPs do not compete [6]. Moreover, in major cities, cable-based ISPs dominate in terms of coverage, i.e., they serve almost all the block groups [17]. In contrast, DSL/fiber providers serve a smaller fraction of block groups. Finally, in part because fiber deployments are relatively new and more expensive to deploy, DSL is often (though not always) offered in more block groups than fiber. Given these trends, cable-based ISPs operate in three distinct modes: cable monopoly, cable-DSL duopoly, and cable-fiber duopoly.

Existing broadband availability datasets. The FCC recently launched a street address-level map of broadband availability [17]. This is an improvement over the previous iteration, based on provider input through Form 477 [28], which offered this information in census block-level granularity. This new map reports the maximum upload and download speeds and the access technology (e.g., fiber, cable) at street-level granularity and relies on self-reporting from ISPs. Previous efforts curated similar data by manually [26] or automatically [44] querying ISP web interfaces, also referred to as a broadband availability tool (BAT). Such third-party efforts enable auditing self-reported data from different ISPs [5, 32, 44].

These datasets improve our understanding of broadband availability, both in terms of coverage and quality. However, without any pricing information, it is not possible to characterize broadband affordability.

Existing broadband plan datasets. Prior efforts have typically curated broadband plan datasets by manually querying ISP BATs. For example, the California Community Foundation and Digital Equity Los Angeles queried Spectrum’s website to curate a list of broadband plans for 165 street addresses in Los Angeles County (California) [42]. One study [47] manually compiled a dataset of 126 street addresses across seven states to obtain available plan information. While these studies highlight the disparity in broadband plans, small-scale datasets are, at best, suggestive of broader and more general trends.

More recently, an online investigative platform, The Markup [20], extended the BAT client [44] approach to automate the extraction of broadband plans for four major ISPs in 43 US cities. Their study [15], which is the most closely related prior work to ours, finds significant variability in the download speed offered by major ISPs at different price points. For instance, the authors found that, for $55/month, AT&T offers 1000 times greater maximum download speed to some addresses in the same city; this phenomenon is referred to as “tier-flattening” [2]. The Markup’s study also finds that some major ISPs, such as AT&T and CenturyLink, provide lower speeds to more vulnerable populations, e.g., low-income and high-minority communities, than others. Based on this analysis, the authors highlight the importance of analyzing the cost of Internet service and download speed instead of download speed in isolation. A limitation of the Markup’s study, however, is that it does not include cable-based ISPs, which serve most of the US population [14]. Consequently, their dataset is not suited to explore the dynamics between cable and DSL/fiber providers nor to the study of how competition between the two changes the nature of broadband plans in a region. In addition to that, as discussed in Section 3.2, extending BAT clients to collect data for all major ISPs is non-trivial.

Our approach. In this work, we address the key gaps of previous efforts by curating a comprehensive broadband plan dataset in terms of location and type of ISPs. First, we develop BQT to obtain plan information across 837k street addresses for three major cable providers and four major DSL/fiber providers. Our dataset provides insight into the ISP plan structure in 30 cities around the US. Using this dataset, we can characterize how ISP plans change between cities, within a city, and in the presence of another ISP.

3 THE BROADBAND-PLAN QUERYING TOOL

Our goal is to develop a robust measurement tool that can accurately report the broadband plans offered by major ISPs for a given set of street-level addresses at scale. Rather than relying on user surveys [42] or self-reporting [40] from ISPs, we focus on directly querying ISP BATs. Minimizing disruption to end users using BAT is an important priority while developing this tool. In essence, for a given list of input addresses, we want this tool to achieve a high hit rate, i.e., successfully extract broadband plans for as many input street addresses as possible, promptly, yet without disrupting the normal service offered by the ISP to end users.

3.1 Challenges

In theory, obtaining broadband plan information from an internet service provider’s BAT should be straightforward. However, in practice, it is often complicated due to the quality of street address datasets. Most street address datasets are crowdsourced [30, 35], which can result in incomplete, incorrect, or ambiguous information. As a result, the querying process is a dynamic, multi-step process, where the information displayed at each step is based on the internal logic and state of each BAT, as well as the input provided by the user in the previous step. For instance, after the user enters a street address, the next web page may either show available broadband plans, indicate an incorrect input address, or inform the user that they are already a subscriber at that address. Additionally, ensuring that the tool can query all major ISP’s is challenging because different ISP’s use different formats and interfaces, such as drop-down menus or click buttons, to present this information and allow users to respond.

To illustrate, Figure 1 shows different steps that our tool needs to follow to extract the broadband plans. Here we use AT&T as an example, but we confirm that all other ISP BATs also follow these steps. In the first case, as illustrated in Figure 1a, AT&T could not identify the input street address. When faced with this scenario, the expected response for the end user is to access the drop-down

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3We categorize DSL and fiber providers together as, if an ISP offers a DSL-based service, it typically also offers a fiber-based service, and vice versa.

5Note for privacy reasons, we have blurred the specific street address in this example.
menu that the BAT provides and then select an address from the offered address set. As a next step, AT&T could indicate that an active customer already exists in this specific street address. In this scenario, the BAT offers three distinct choices, as shown in Figure 1b. If a user is already an AT&T subscriber residing in that address, the first two options give them the ability to change their plan or add a new plan. This would prompt the BAT to render an authentication form to ensure the user is an active subscriber. The third option applies to a new customer who is interested in viewing the set of AT&T plans at that address. This step does not require any authentication. Finally, a particular address could be a multi-dwelling unit, i.e., with an apartment/unit number that was not input during the initial stage. For that scenario, as demonstrated in Figure 1c, the BAT provides an option to select one of the possible apartments/units at that address.

3.2 Strawman: Extend Existing BAT Client
A potential solution to obtain broadband plan information is to enhance the BAT client approach proposed in previous research [44]. This approach was designed to query the binary availability of broadband service (i.e., service/no service) for a specific street address. For every ISP, a BAT client was designed, which involved reverse-engineering each ISP BAT by observing how it uses different RESTful APIs to extract the desired information, such as broadband availability. For example, the BAT client can observe that when a browser sends a request with a street address, it receives a response with an ID, and subsequent requests in the next step use this ID and, in some cases, a session cookie from the previous step. The BAT client then uses the Python requests library to directly send a series of requests to the ISP’s RESTful APIs. Directly querying the APIs is scalable; thousands of street addresses can be handled in parallel. In 2020, the authors in [44] used this approach to query approximately 35 million street addresses. Their data analysis revealed the limitations of the information provided by the FCC’s Form 477 [28], reinforcing the need for such information to be made available at street-level granularity as previously suggested by other research [37, 51].

Limitations. Since the BAT client approach has been successfully used to query millions of street addresses for all major ISPs, extending it to extract offered broadband plans seems like a natural choice. However, we observed that the proposed approach has several limitations that make it difficult to adapt to satisfy our goals. Specifically, since the publication of the previous work [44], ISPs have safeguarded their RESTful APIs from such direct querying. For example, some ISPs have now started using dynamic cookies that append unique server-side parameters to each user session. Some BATs have started blocking queries from an IP address that uses the same cookie across multiple API requests. Dynamically generating a new cookie for each API request is non-trivial and is not supported by the original BAT client.

3.3 BQT Approach
To decouple the querying process from ISP safeguarding strategies, our approach avoids directly querying their RESTful APIs. Instead, we use a popular web automation tool, Selenium, to mimic different end-user interactions for extracting the desired broadband plan information.

As a first step, we manually inspect the workflow for different ISP BATs. Each BAT employs a specific template to display the information for each step in the workflow. As part of this manual bootstrapping step, we enumerate all possible templates and identify unique patterns in their HTML content using regular expressions to help detect them at runtime.

The second challenge is to identify how to mimic a user’s behavior using Selenium to advance successfully to the next step. This step is critical for ensuring a high hit rate for BQT. Specifically, we handle different templates as follows.

Incorrect address. As mentioned earlier, street addresses are noisy due to inherent ambiguity between different identifiers. For example, for the same street address, some databases might use “Ave” instead of Avenue and “CT” or “Ct” instead of Court. Whenever there is a mismatch between the input street address and the one in the ISP’s database, it shows an “incorrect address” web page and often provides a list of one or more street addresses as suggestions. Given the prevalence of this occurrence, addressing it is critical to ensure a high hit rate for BQT. We address this issue by storing the list of suggested street addresses for offline analysis. We then apply string-matching over each suggested address in this list to find the one that best matches the input street address. As a sanity check,
we ensure that the selected street addresses have the same zip code as our initially queried address. We then query the ISP’s BAT to extract the broadband plan information.

Multi-dwelling units. For addresses where a specific street address has multiple dwelling units (e.g., two or more apartments), the ISP BAT typically shows a “multi-dwelling unit” web page and suggests more refined street addresses (e.g., specific apartment numbers). Similar to previous work [44], we replace the input street address with a randomly selected address from this list. We then use this new address to query the ISP’s BAT to extract the broadband plan information.

Existing customers. If the resident of an input street address is already a subscriber, the ISP BAT displays an “existing customer” web page and offers two options. The first option directs the user to their account, while the second allows a new user to query offered plans. Given our interest in extracting the available broadband plans, we select the second option.

To avoid failures, we must ensure that all the Document Object Model elements for a step are successfully downloaded before applying any user action. The download times can vary across different templates and ISPs. For example, the step that displays available broadband plans after inputting the street address takes less than 30 seconds for AT&T but 60 seconds for Spectrum. Thus, we measure the download times for all possible templates and pause for this period (i.e., max observed download time) before applying the user action.

Microbenchmarks. The two crucial performance metrics of BQT are hit rate and query resolution time. The hit rate informs the fraction of total queried addresses for which we are able to obtain a response from a particular ISP BAT successfully. As shown in Figure 2, our hit rate for all ISPs exceeds 80%; we achieve the highest hit rate of 96% for Cox and the lowest for Spectrum (82%). Such high hit rates across all ISPs ensure that BQT is able to extract plan information for the majority of the addresses. Our investigation into the instances where BQT encounters failures reveals that the primary cause is the denial of connectivity by the IP proxy service. Furthermore, some ISPs classify certain requests as originating from data centers due to IP addresses, resulting in service denial and subsequent failures. If we re-run the addresses that previously failed, there is an increase in the BQT’s hit rate for each ISP. The query resolution time for a given street address is the amount of time it takes BQT to obtain a response from an ISP BAT. Figure 3 presents the distribution of query resolution time for each ISP. The median time for Frontier query resolution is lowest, at 27 seconds, while it is highest, at 100 seconds, for Spectrum, despite no significant difference in the number of intermediate webpages rendered. Given that this latency can be significant, we describe the methodology we adopt to make BQT more scalable in Section 4.1.

Limitations. BQT has been specifically designed to work with the BATs offered by seven major ISPs. However, any changes made to the interfaces of these BATs by the ISPs, such as the addition of new drop-down forms, will require BQT to be updated. To ensure that BQT continues to function properly over time, we must monitor the BATs for all the supported ISPs and upgrade BQT as necessary to accommodate any changes. In the future, we plan to make BQT more modular, which will help minimize the effort required to adapt it to these changes.

4 BROADBAND PLAN DATASET CURATION

In this section, we describe the dataset we aggregate through BQT. We first describe our methodology to query a subset of street addresses and ISPs to curate the broadband affordability dataset. We then describe how we selected the ISPs, cities, and street addresses for data collection (Section 4.1). Next, we discuss how we addressed different ethical concerns regarding our data-collection methodology (Section 4.2). Finally, we discuss the limitations of our dataset (Section 4.3).

4.1 Data Collection Methodology

In the US, seven ISPs serve approximately 90 million street addresses (87% of the total US census blocks) [17]. Through our data usage agreement with Zillow [35], we have access to about 104 million “residential” US street addresses. Note that while this database does not represent every US address (it is comprised of addresses that had a transaction during a specific period), it encompasses a very large subset. Further, compared to alternative address datasets, such as the National Address Database (NAD) [30] offered by the US Department of Transportation, the Zillow dataset offers more complete coverage and is less noisy. Specifically, it includes nearly
We focus on cities that are served by at least two of the seven ISPs we selected. 30 major cities in 27 states (see Figure 4) are well represented in the Zillow dataset. After applying this filter, Table 2 in the appendix, these cities represent a broad spectrum of diverse socioeconomic attributes (e.g., average income) that are considered in our work to ensure that we capture any trends that diminish returns. Our initial exploration of the collected data revealed that broadband plans are spatially clustered, so plans for street addresses in the same neighborhood (i.e., a census block group) are similar. Additionally, our primary focus for this study is metropolitan/urban areas around the US. Given the coverage of Zillow’s address database, we can extend the scope of the study to micropolitan and rural areas in future work.

With this context in mind, we now describe our selection methodology for the ISPs, cities, and street addresses for our study.

**ISP selection.** We focus on fixed, terrestrial broadband providers that offer queriable BATs and serve at least a million street addresses in Zillow’s dataset. After applying this filter, there are seven major ISPs: AT&T, Verizon, CenturyLink, Frontier, (Comcast) Xfinity, (Charter) Spectrum, and Cox. Among them, Xfinity, Spectrum, and Cox are cable-based providers, and AT&T, Verizon, CenturyLink, and Frontier offer DSL and fiber-based plans. Previous work reported that Comcast Xfinity’s offerings are invariant to location [15]. Our analysis using the data collected using BQT from six major US cities confirmed these observations, and so we omit collecting data for this provider.

**City selection.** With a goal of wide geographic distribution, we examined cities with a range of population densities as well as diverse socioeconomic attributes (e.g., average income) that are well represented in the Zillow dataset. After applying this filter, we selected 30 major cities in 27 states (see Figure 4). As shown in Table 2 in the appendix, these cities represent a broad spectrum of demographic and socioeconomic attributes. For example, the range of population densities varies from 1 k to 42 k per sq. mile [1], and the median yearly household income varies from $31 k to $101 k. We focus on cities that are served by at least two of the seven ISPs considered in our work to ensure that we capture any trends that emerge as a result of competition between ISPs in a region.

**Street address selection.** Each city in the US is divided into census blocks, which are aggregated into census block groups. The US Census Bureau defines a census block group (CBG) as representing approximately 600–3000 people that are considered to be homogeneous in terms of their demographic and socioeconomic characteristics. For the cities considered in this work, Zillow’s database includes addresses in all the census block groups for each city, ensuring comprehensiveness. However, as querying every address in a city would impose a significant load on the ISPs’ infrastructure, we opt for a sampling strategy. To ensure that our sampling strategy mimics the socioeconomic composition of the city, we uniformly sample street addresses at the census block group level. Specifically, for each (ISP, city) pair, we identify the set of block groups covered by the ISP in a city. We randomly sample 10% of street addresses for each such block group. If we are unable to obtain the BQT data for any of those addresses, we continue sampling until we have a successful sample of 10% of street addresses in each CBG.

**Scaling data collection.** To gather the needed samples for our study, BQT needs to query 837 k street addresses, the total number of addresses resulting from sampling 10% of every census block group. We run multiple instances of BQT in parallel to scale the data collection. We use Docker containers to run these instances concurrently on a single local data-collection server. We can theoretically use as many containers as street addresses for different ISPs to expedite data collection. However, such an approach will overwhelm ISP BATs and degrade the user experience for actual customers.

Though we cannot directly measure the experience for real users, we conducted an experiment where we measured ISP response time for 1, 50, 100, and 200 Docker instances. We hypothesize that if running multiple Dockers is affecting user experience, we should expect a statistically significant difference in ISP response time for different settings. However, we observed that the response time for any ISP did not change as we increased the number of Docker instances. Based on this experiment, we are confident that using up to 200 Docker instances does not overwhelm ISP servers enough to disrupt the user experience. Nevertheless, we scale back and utilize 50–100 distinct containers for our data collection. Note that our choice of 200 instances is based on the intuition that an ISP should not get overwhelmed by such a small number. By no means is it an upper bound on how many Docker containers we can run in parallel.

To ensure that all our queries do not originate from a single non-residential IP address, we utilize a pool of residential IP addresses provided by Bright Initiative, the non-profit branch of Bright Data [25] (formerly known as Luminati). This organization offers free access to data scraping tools for nonprofits and academic organizations. Previous efforts [15, 44] have also used this service.

We conduct our data collection campaign from December 2022 to February 2023.

**Public release.** We will make a version of this dataset publicly available to empower other researchers and policymakers to improve our communal understanding of broadband affordability in the US. Due to the proprietary nature of Zillow’s data, we cannot include specific street addresses in our dataset. Instead, for each queried street address, we will only reveal its block group identifier along with the corresponding ISP plans. Considering the limited variability in broadband plans within a block group (see
We now discuss a few limitations of our dataset and how to address them. Our dataset includes addresses from every census block group in the 30 cities examined in this study. It also shows which of the seven major ISPs are active in each city and hence in our dataset. Overall, our dataset covers 837 k distinct street addresses, representing 18 k block groups (around 7.5% of total block groups in the US). None of the thirty cities are served by more than two major ISPs. This trend indicates the presence of monopolies and duopolies in these cities [6].

Table 2 in the appendix summarizes the number of street addresses and block groups we cover for each of the thirty cities. It also shows which of the seven major ISPs are active in each city and hence in our dataset. Overall, our dataset covers 837 k distinct street addresses, representing 18 k block groups (around 7.5% of total block groups in the US). None of the thirty cities are served by more than two major ISPs. This trend indicates the presence of monopolies and duopolies in these cities [6].

Table 1 summarizes the available broadband plans from the seven major ISPs. The dashed line separates DSL/fiber-based providers from cable-based ones.

<table>
<thead>
<tr>
<th>ISP</th>
<th>Unique Plans</th>
<th>Download (Mbps)</th>
<th>Upload (Mbps)</th>
<th>Monthly Price ($)</th>
<th>cv</th>
</tr>
</thead>
<tbody>
<tr>
<td>AT&amp;T</td>
<td>11</td>
<td>0.768–1000</td>
<td>0.768–1000</td>
<td>55–80</td>
<td>0.01–12.5</td>
</tr>
<tr>
<td>Verizon</td>
<td>4</td>
<td>3.1–1000</td>
<td>1–1000</td>
<td>50–100</td>
<td>0.4–11.1</td>
</tr>
<tr>
<td>CenturyLink</td>
<td>8</td>
<td>1.5–940</td>
<td>0.5–940</td>
<td>50–65</td>
<td>0.03–14.5</td>
</tr>
<tr>
<td>Frontier</td>
<td>2</td>
<td>0.2–2000</td>
<td>0.2–2000</td>
<td>50–100</td>
<td>0.0004–20.0</td>
</tr>
<tr>
<td>Spectrum</td>
<td>5</td>
<td>36–750</td>
<td>5–35</td>
<td>20–57</td>
<td>17.5–24.3</td>
</tr>
<tr>
<td>Cox</td>
<td>6</td>
<td>100–1000</td>
<td>5–35</td>
<td>20–120</td>
<td>10.0–28.6</td>
</tr>
<tr>
<td>Xfinity</td>
<td>3</td>
<td>25–1200</td>
<td>5–35</td>
<td>20–80</td>
<td>3.8–15.0</td>
</tr>
</tbody>
</table>

Table 1: Overview of broadband plans offered by the seven major ISPs. The dashed line separates DSL/fiber-based providers from cable-based ones.

4.2 Ethical Considerations

We query ISP plans at the street address level and do not collect or analyze Personally Identifiable Information (PII). Our work does not involve human subjects research, and the private dataset provided by Zillow under the data use agreement does not reveal any individual’s identity. Furthermore, the data gathered from the website does not include any PII. We do not have the means to identify residents, the selected broadband subscription tiers, or the actual performance received at any address. Our methodology involves obtaining ISP plan information from their websites, which is consistent with legal requirements and research community norms [11, 12, 19].

4.3 Limitations

We now discuss a few limitations of our dataset and how to address them in the future.

Staleness issues. Our dataset provides a single snapshot of broadband plans, which may change over time as ISPs update their infrastructure and pricing structures. We observe that many ISPs are actively deploying new fiber, and we expect their offered plans to change in the near future. Also, ISPs occasionally offer discounts (i.e., higher carriage value (cv) plans), especially in areas where they compete with other major ISPs. Our dataset does not discriminate between normal and discounted offers and, thus, might not best reflect the most recent carriage for a subset of street addresses.

Limited coverage. Although our dataset includes addresses from every census block group in the 30 cities examined in this study, it represents only about 7.5% of all block groups in the US. We currently use Zillow’s data, which is biased toward high-density urban areas. We need a better representation of street addresses in semi-urban and rural areas. Though curating such datasets is challenging, recent efforts from the FCC to develop broadband availability maps at street address granularity demonstrate such an approach’s feasibility. In future work, we will complement Zillow’s dataset as needed with other sources, such as the NAD, to cover other areas where Zillow’s data alone lacks sufficient representation.

Veracity of reported plans. There is no system or database to confirm the accuracy of the download speed and price data provided by the ISPs when querying a street address. However, as mentioned in [44], it is not in the interest of ISPs to report false or misleading information to potential customers, including poor performance or low-valued plans. We note that the total cost incurred by subscribers for ISP services often exceeds the initially advertised prices. This includes subscriber-specific discounts, undisclosed fees, taxes, and additional charges [23]. However, our focus in this study is the initial advertised price offered by ISPs; the analysis of the final amount paid by subscribers is beyond the scope of the current work.

5 BROADBAND PLAN CHARACTERIZATION

In this section, we will demonstrate how our broadband affordability dataset provides the means for various stakeholders to address crucial policy questions that previously were difficult to answer. To do so, we will first present an overview of the BQT dataset. We will then answer the following critical questions: ❶ Do the broadband plans, characterized by their carriage value, change by city for different ISPs? ❷ Does the carriage value change within a city? If yes, which neighborhoods (identified by their census block groups) receive good and bad deals (high and low carriage values)? ❸ Does competition among ISPs impact the carriage value offered to the end users? If yes, is there a trend in which neighborhoods experience competition? ❹ Is the quality of available deals correlated with demographic and socioeconomic factors? If yes, which population groups receive better or worse deals from the ISPs?

5.1 Dataset and Metrics

Dataset overview. Table 2 in the appendix summarizes the number of street addresses and block groups we cover for each of the thirty cities. It also shows which of the seven major ISPs are active in each city and hence in our dataset. Overall, our dataset covers 837 k distinct street addresses, representing 18 k block groups (around 7.5% of total block groups in the US). None of the thirty cities are served by more than two major ISPs. This trend indicates the presence of monopolies and duopolies in these cities [6].

Figure 5: Distribution of coefficient of variation of carriage values in a block group for each ISP.
Calculating carriage values. We use the carriage value to characterize a broadband plan offered by an ISP, and we curate this metric for all input street addresses. Since the entropy of available download speeds is greater than the upload speeds, we focus on download speed to calculate carriage value. While not shown, we verified that our results are consistent if we use upload speed to determine carriage value.

Each ISP offers a fixed number of plans across all cities. For example, AT&T offers 11 different plans across the 14 cities it serves in our study. However, an ISP only offers a subset of these plans at any given street address. For example, for a specific street address in New Orleans, AT&T offers three different plans: (1000 Mbps, $80/month), (500 Mbps, $65/month), and (300 Mbps, $55/month), which translates to carriage values of 12.5, 7.7, and 5.5, respectively. To represent the value provided by an ISP through a set of plans to a street address, for every address, we consider the best carriage value (\(c_v\)) to the above address. We note that the \(c_v\) metric has inherent limitations due to the nature of broadband pricing. Since speed tends to vary more than price—e.g., at the address mentioned above, 1.5x cost gets 3.3x more bandwidth—the highest carriage value (\(c_v\)) plan available for an address is also the highest-speed plan. Users may not require the highest speed available or want to pay for it, so \(c_v\) is not necessarily a reliable proxy for the subjective value of a plan to its customers. For this and other reasons, policy decisions should not optimize around \(c_v\) alone.

In some of the analysis that follows, we compare block groups by carriage value. The \(c_v\) of a block group provided by an ISP is computed as the median of the maximum carriage values of the plans sampled from the addresses in that block group. Using an aggregate metric at block group granularity simplifies spatial analysis, and ensures that our analysis is not biased by block groups with more street addresses in the Zillow dataset. However, it also hides variability within block groups. To characterize this variability, Figure 5 shows a distribution of the coefficient of variation (CoV), i.e., the ratio of the standard deviation to mean, for the \(c_y\) available per ISP for every block group in our data set. Most ISPs show low CoV across all block groups, meaning the aggregate \(c_y\) metric hides little information. However, there is a long tail for AT&T and CenturyLink, which sometimes offer both DSL (very low \(c_y\)) and fiber (very high \(c_y\)) plans within the same block group. We checked the robustness of our per-block-group findings by performing an analysis where block group \(c_y\) was computed as the median of the minimum \(c_y\) of the sampled plans; our conclusions (e.g., Section 5.2, Figure 8) were consistent regardless of this choice.

Comparing plans. To compare an ISP’s plans across different cities or the plans of two competing ISPs within a city, we need to quantify the differences in the plans. To this end, we represent the available plans from an ISP in a city using a plans vector of 30 dimensions, each representing a discrete carriage value. We then quantify the differences using the L1 norm between the two vectors. The weight for each dimension is determined by the fraction of block groups in the city that receive that specific carriage value, and the \(\ell_1\) operator is used to discretize the carriage values. For example, Cox offers a carriage value of around 10.5 and 11.3 in 35% and 12% of block groups in New Orleans, 12% and 6% of block groups in Oklahoma City, and 4% and 21% of block groups in Wichita. The L1 norm between New Orleans and Oklahoma City plans is 1.78 (different). Between New Orleans and Wichita, is 1.57 (different), and between Oklahoma City and Wichita is 0.36 (relatively similar).

Note that the maximum carriage value we observed across all ISPs and cities is 28.6 (Table 1).
We observe that DSL/fiber-based provider plans are less diverse in Wichita and Oklahoma City.

To answer the question of why DSL/fiber plans have higher carriage values, we analyze the distribution of plans at block group granularity. We only visualize one provider from each DSL/fiber (AT&T) and cable (Cox) category for brevity. To simplify the exposition, Figure 6 shows the distribution of carriage value for only five cities (out of 14 and 6, respectively) for each ISP.

For AT&T, we observe two sets of peaks in broadband plans. The higher carriage value peak is attributable to fiber-based plans and the lower to DSL-based plans. The fraction of block groups that receive fiber plans differs in each of the cities. For example, in New Orleans, 32% of block groups receive fiber-based access, which is significantly smaller than the 54% and 57% of block groups in Wichita and Oklahoma City.

For Cox, we observe six different peaks, and the distribution of the carriage values across block groups varies significantly by city. For example, Cox offers CV of about 28 Mbps/$ to 7% of block groups in New Orleans. In contrast, Cox offers similar plans to 21% and 18% of block groups in Oklahoma City and Wichita, respectively. On the other hand, 44%, 46%, and 50% of block groups in Wichita, New Orleans, and Oklahoma City receive CV of 14.6 Mbps/$.

To illustrate how this trend generalizes for other cities and ISPs, Figure 7 shows the distribution of L1 norm, i.e., the difference in available plans between all pairs of served cities for each ISP. A low L1 norm indicates similarities in broadband plans and vice versa. We observe that DSL/fiber-based provider plans are less diverse across different cities than cable-based providers, with AT&T (most similar) and Spectrum (most diverse) at the extremes. This result demonstrates that some ISPs alter their plans between cities while others maintain consistent offerings throughout their service areas.

5.2 Inter-City Broadband Plans

To understand why AT&T and Cox offer disparate plans to users within a city, we zoom in on Cox and AT&T in New Orleans, individually and as a pair (see Figure 8c). Comparing Figures 6a and 8b, we observe that Cox offers better coverage and higher CV values in most block groups. Given its lower proliferation of high CV fiber plans, if we look at the plans only from AT&T in this city, which was the case in one of the previous studies [15], we might get an impression that the nature of broadband plans is problematic for all New Orleans residents. Specifically, the broadband plans are sparse and highly variable (DSL vs. fiber), and most residents get the “worst” deal, i.e., low CV values. However, the competing cable-based provider is the dominant ISP in the city, and its plans are not as extreme nor sparse. Figure 8c shows that if we consider the AT&T and Cox plans together, i.e., when we report the highest CV value from either of the two providers, the best CV value is similar to that of the dominant cable-based ISP, i.e., Cox in this case. We make similar observations for other cities as well. In our dataset, we do not find a case where the DSL/fiber-based providers offer better coverage or higher average CV values than the cable-based providers.

Spatial clustering. We visually observe that broadband plans are clustered, i.e., the likelihood that two contiguous block groups have similar available plans is high. To validate this visual understanding, we compute the spatial autocorrelation metric using Moran’s I method [21] to characterize the extent of correlation in CV values among nearby block groups. This metric has been widely used in previous studies [43, 55] to understand the spatial distribution of a variable of interest (i.e., CV value) within a geographic region (i.e., city). A positive value of Moran’s I statistic means that similar CV values tend to be found near each other, while a negative value means dissimilar values are found near each other, with zero indicating a complete lack of association of CV values with locations.

We computed the Moran’s I statistic for all (ISP, city) pairs to measure the spatial autocorrelation of broadband plans. The results show that, with the exception of Xfinity, the median value ranges between 0.3–0.5, indicating a high level of spatial clustering in broadband plans across ISPs within a city. Given that AT&T is a
DSL/fiber-based provider, such clustering of its carriage value can be attributed to its fiber infrastructure deployment around the city. Table 3 in the appendix reports the median value across all cities for each ISP.

Our results show that both DSL/fiber and cable ISPs offer similar \( cv \) plans to neighboring census block groups within a city. Similar to the case for AT&T, the spatial clustering of plans for DSL/fiber providers is related to the nature of access technology. Neighborhoods with fiber deployments receive better carriage value and vice versa. However, since cable-based ISPs use the same technology across the city, spatial clustering in their plans is intriguing. In the next section, we explore whether this behavior is attributable to competition among ISPs.

### 5.4 Impact of Competition

To answer \( \Theta \) (does competition among ISPs impact the carriage value offered to the end users? If yes, is there a trend in which neighborhoods experience competition?), we explore whether the cable-based ISP’s plans change when they operate as a monopoly vs. when they compete as a duopoly. We did not analyze DSL/fiber-based providers alone from the perspective of operating as both a monopoly and a duopoly because we did not observe this pattern in any of the thirty cities. We employ a statistical test to discern whether competition (or lack thereof) leads to a change in cable providers’ carriage value. For every city with competition between cable and DSL/fiber providers, we run two one-tailed 2-sample Kolmogorov–Smirnov (KS) tests [29].

Our null hypothesis \( (H_0) \) is that there is no difference in the \( cv \) offered by a cable provider in locations where they operate as a cable monopoly compared to locations where they operate as a cable-DSL duopoly or cable-fiber duopoly. To test this hypothesis, we run one test for each of the following alternate hypotheses \( (H_i) \).

In the first one-tailed test, we propose \( H_1 \), which states that the \( cv \) provided by the cable provider is greater for block groups in duopoly locations than those in cable monopoly locations. In the second test, we reverse the hypothesis from the previous test and propose \( H_2 \), which states that cable providers provide better \( cv \) for block groups in cable monopoly locations than those in each duopoly category. By conducting two tests per category, we can detect either scenario and provide robust statistical evidence of the impact of competition on cable offerings for different types of DSL/fiber-based offerings.

If we achieve a p-value of less than 0.05, we reject the null hypothesis \( (H_0) \) for the corresponding test and record the corresponding KS test statistic, denoted by the \( D \) value. We conduct this analysis for all combinations of cable and DSL/fiber providers in other cities. In the remainder of the section, we use New Orleans as a case study to explain our findings.

**Cable-DSL Duopoly:** In the first test, our \( H_1 \) is Cox’s \( cv \) in cable monopoly block groups is lower than the cable-DSL duopoly block groups. Conversely, our \( H_2 \) is Cox’s \( cv \) in cable monopoly block groups is higher than cable-DSL duopoly block groups. Figure 9 shows that Cox’s offered \( cv \) in the DSL duopoly block groups is similar to its \( cv \) in monopoly block groups. This is further confirmed through the K-S test, where we fail to reject \( H_0 \), which signifies there is no statistical difference in Cox’s \( cv \) distribution in block groups where it serves alone and block groups where it competes with AT&T’s DSL offerings. The median \( cv \) for both cases is 11.38 Mbps/$. We observe the same trend for other pairs of Cable-DSL duopolies within cities in our dataset.

**Cable-Fiber Duopoly:** We posit a similar hypothesis for cable-fiber duopolies. Figure 9 shows the difference in Cox’s \( cv \) distributions between these block group types, which is further reinforced by the K-S test where we reject \( H_0 \) with statistical significance in favor of \( H_1 \). Contrarily, \( H_2 \) cannot be accepted as the p-values exceed 0.05. This result points towards Cox increasing the \( cv \) provided through its plans by lowering the price for the same download speed in block groups where it faces competition from AT&T’s higher \( cv \) fiber offerings. The median \( cv \) from Cox in such addresses is 14.63 Mbps/$, 30% more than the monopoly and DSL block groups’ median \( cv \). For the remaining combinations of cable and DSL/fiber providers in other cities, we capture the same trend, indicating differential pricing structures from cable providers in the presence of high \( cv \) competition.

Our analysis in this section has demonstrated that cable providers tend to improve the carriage value offered through their plans in locations where fiber-based plans are present. This places fiber plans in a critical position because they tend to yield better broadband deals.

### 5.5 Influencing Socioeconomic Factors

In the prior sections, we established that low \( cv \) is associated with DSL plans. In this section, we investigate whether there is a trend in which sociodemographic groups predominantly receive DSL plans and, therefore, worse \( cv \). This analysis will enable us to answer \( \Theta \) (is the quality of available deals affected by demographic and socioeconomic factors? If yes, which population groups receive better or worse deals from the ISPs?). To do so, we compute the percentages of block groups within every city that receive DSL or fiber plans disaggregated by the block group level median household income.
We merge our dataset with the ACS data to obtain the median household income of every census block group. Concretely, we adopt a methodology similar to [15, 36] to group each city’s census block group-level income into two distinct categories: low (below the city’s median household income) and high (exceeding the city’s median household income). For each income group class within a city, we calculate the percentage of block groups that have access to fiber-based plans. Subsequently, we compute the percentage difference in fiber deployment between the high and low-income groups of the block group.

Figure 10 presents the breakdown of the percentage of block groups that receive AT&T’s DSL and fiber plans in the two income categories of block groups in New Orleans. 41% of the low-income census block groups receive AT&T’s fiber plans while 57% of the high-income block groups exceed 10% in seven cities, while in four cities, it is below 10%. No difference is observed in Austin, TX; however, in Wichita, KS, and Atlanta, GA, a higher proportion of low-income census block groups receive fiber from AT&T compared to high-income groups. Figure 11 shows that CenturyLink and Verizon exhibit a comparable pattern, where a larger proportion of high-income block groups across cities receive fiber compared to lower-income groups. Frontier emerges as an outlier in this analysis.9 Given that the lack of fiber also leads to lower cv from cable providers, internet users in block groups that lack fiber connectivity tend to get more bad deals overall compared to others. We conducted a similar analysis for the demographic attributes of race and population density. The results for these variables did not produce comparable trends.

6 RELATED WORK

In [6], the authors analyzed FCC Form 477 data and reported that close to 50 million people in the US live in locations served by a single ISP, i.e. in an ISP monopoly. While not considering the price/cost associated with internet access, several studies have sought to understand how internet quality itself varies between different locations and demographic variables. The Census Bureau produces an annual list of US cities with the lowest Internet connectivity using data from the American Community Survey (ACS) One Year estimates [24]. However, this estimate does not take into account the cost of access. The work conducted in [44] demonstrated that the FCC National Broadband Report significantly overestimates coverage and examined the digital divide in terms of the lack of coverage in rural and marginalized communities. Similar inaccuracies of the FCC map were found for mobile networks in [45]. In [4], the authors analyzed the relationship between income and download speed at the geographic granularity of US zip codes. The work utilized income data, grouped into five income bins, obtained from 2017 tax returns filed with the Internal Revenue Service. The study demonstrated a positive correlation between zip code income and download speed. The authors of [36] analyzed publicly available data from Ookla [50], a popular speed test vendor, and found significant differences in key internet quality metrics between communities with different income levels. In [51], the authors utilized M-Lab [46] speed test data in California and found higher internet quality in urban and high-income areas. Several studies have also examined how the cost of electricity varies across locations and demographic variables. The authors in [53] discovered that minority groups in various US cities pay a disproportionate amount for electricity compared to other communities. Similar findings are reported in [22].

7 CONCLUSION

In this work, we explore broadband affordability in the US. Specifically, we analyze the nature of broadband plans offered by seven
major ISPs across thirty different US cities. To aid this effort, we developed BQT, a new scalable tool that extracts broadband plans offered by the seven major US ISPs at any street address. We use this tool to curate a dataset that reports broadband plans offered to 837 k street addresses, spanning 18 k census block groups in the thirty cities. To the best of our knowledge, this is the largest such broadband plan pricing dataset in existence. Our analysis sheds light on pricing strategies adopted by different ISPs, which have previously been opaque. Our results highlight the importance of competition, and specifically on how fiber deployments benefit end users. It also identifies the population groups reaping the benefits of competition and fiber deployments. We believe this effort is a step towards improving public understanding of US broadband affordability. We will make our tool and dataset publicly available to facilitate further research.

Drawing from our experiences, we recommend that regulators and policymakers take the following actions: (1) The FCC should consolidate the broadband availability maps [17] and urban rate survey [40] to ensure that the public has access to both availability and pricing information at the street address level. Based on our findings, it is evident that the speed offered by an ISP is a crucial factor to take into account. However, there is significant variability in the prices at which these speeds are offered to customers. Moreover, it is essential to assess the complete, actual costs incurred by subscribers for these ISP services. Previous work [23] has documented that ISPs frequently include extra fees and charges in their pricing structure. If such comprehensive information is collected by the FCC and subsequently made public, these pricing strategies can be better studied, decreasing the lack of transparency that currently exists within the ISP service provider sector. (2) Beyond the availability and cost of access, actual performance data about fixed broadband service is critical for fully characterizing digital inequality, yet it remains elusive. While the FCC currently operates the Measuring Broadband America (MBA) [39] project, its coverage is not pervasive (less than 3k households around the country). If ISPs are mandated to provide information about actual performance experienced by their subscribers, we can complement the research presented in our work to understand not just what service ISPs promise to deliver, but what service they actually do deliver. (3) Even if the FCC provides such data, third-party audits are essential to verify the accuracy of self-reported information from ISPs. However, existing US street address datasets are private, sparse, and noisy, posing a challenge to such third-party efforts. Therefore, local governments (e.g., county) should put more effort into improving the quality and availability of street address datasets in their areas. This will enable and encourage additional research within this field, consequently leading to a more comprehensive understanding of facets related to ISP service provisioning. Finally, (4) policymakers should consider subsidizing fiber deployment efforts [10] or enforcing rate regulations [7], even in urban areas, to help improve the carriage value for broadband plans in low-income block groups that can be ignored or deprioritized by major ISPs. This would improve competition and carriage value, as our study has demonstrated that fiber deployments play a critical role in providing subscribers with the option of high carriage value plans from different types of ISPs.

ACKNOWLEDGMENTS
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REFERENCES
Analyzing Disparities in Broadband Plans Offered by Major US ISPs

A. APPENDICES

Appendices are supporting material that has not been peer-reviewed.

A.1 Xfinity cv

Figure 12 shows the cv of Xfinity the six cities where we collected its plan information. As mentioned previously, our data indicates Xfinity does not implement a differential pricing structure, unlike the six other ISPs considered in our work.

<table>
<thead>
<tr>
<th>City</th>
<th>cv</th>
</tr>
</thead>
<tbody>
<tr>
<td>Atlanta</td>
<td>0.0</td>
</tr>
<tr>
<td>Baltimore</td>
<td>0.0</td>
</tr>
<tr>
<td>Boston</td>
<td>0.0</td>
</tr>
<tr>
<td>Chicago</td>
<td>0.0</td>
</tr>
<tr>
<td>Fort Wayne</td>
<td>0.0</td>
</tr>
<tr>
<td>Philadelphia</td>
<td>0.0</td>
</tr>
</tbody>
</table>

Figure 12: Distribution of broadband plans for Xfinity.
<table>
<thead>
<tr>
<th>Block Groups</th>
<th>Street Addresses (k)</th>
<th>Population Density (k)</th>
<th>Median Income (k)</th>
<th>Major ISPs</th>
</tr>
</thead>
<tbody>
<tr>
<td>Albuquerque, NM</td>
<td>387</td>
<td>14</td>
<td>1.8</td>
<td>53</td>
</tr>
<tr>
<td>Atlanta, GA</td>
<td>389</td>
<td>12</td>
<td>1.2</td>
<td>65</td>
</tr>
<tr>
<td>Austin, TX</td>
<td>487</td>
<td>25</td>
<td>1.7</td>
<td>74</td>
</tr>
<tr>
<td>Baltimore, MD</td>
<td>1188</td>
<td>42</td>
<td>1.7</td>
<td>81</td>
</tr>
<tr>
<td>Billings, MT</td>
<td>98</td>
<td>3</td>
<td>1.1</td>
<td>61</td>
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<td>Birmingham, AL</td>
<td>354</td>
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<td>716</td>
<td>47</td>
</tr>
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<td>Boston, MA</td>
<td>373</td>
<td>17</td>
<td>8.4</td>
<td>72</td>
</tr>
<tr>
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<td>472</td>
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<td>2</td>
<td>73</td>
</tr>
<tr>
<td>Chicago, IL</td>
<td>1933</td>
<td>86</td>
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<td>64</td>
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<td>5</td>
<td>1</td>
<td>59</td>
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<tr>
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<td>5</td>
<td>1.5</td>
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<td>0.9</td>
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<td>New Orleans, LA</td>
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<td>2.9</td>
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<td>50</td>
</tr>
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<td>28</td>
<td>1.7</td>
<td>62</td>
</tr>
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<td>981</td>
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<td>8</td>
<td>46</td>
</tr>
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<td>32</td>
<td>1.9</td>
<td>64</td>
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<td>2</td>
<td>79</td>
</tr>
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<td>Seattle, WA</td>
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<td>28</td>
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<td>101</td>
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<td>536</td>
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<tr>
<td>Virginia Beach City, VA</td>
<td>112</td>
<td>4</td>
<td>1.8</td>
<td>80</td>
</tr>
<tr>
<td>Wichita, KS</td>
<td>304</td>
<td>13</td>
<td>1.3</td>
<td>50</td>
</tr>
<tr>
<td><strong>Total</strong></td>
<td><strong>18k</strong></td>
<td><strong>837</strong></td>
<td></td>
<td></td>
</tr>
</tbody>
</table>

Table 2: Dataset coverage. The major ISPs are listed in the following order: (1) ATT, (2) Verizon, (3) CenturyLink, (4) Frontier, (5) Spectrum, (6) Cox, and (7) Xfinity. Note that Xfinity also provides service in Albuquerque, but we did not include this service in our study.

<table>
<thead>
<tr>
<th>Individual ISPs</th>
<th>1</th>
<th>2</th>
<th>3</th>
<th>4</th>
<th>5</th>
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Table 3: Statistical evidence for spatial clustering. We report the median of Moran I statistics across all cities.