



World Scientific News

An International Scientific Journal

WSN 217 (2026) 60-73

EISSN 2392-2192

Predicting the Societal Benefits of Integrated Cycling and Public Transit Networks

Tomas U. Ganiron Jr

Graduate School, Adamson University, 900 San Marcelino Street, Ermita, Manila, 1000

E-mail address: tomas.ganiron.jr@adamson.edu.ph

<https://doi.org/10.65770/RMMY3629>

ABSTRACT

Integrating cycling with public transit offers a sustainable alternative to private car usage, addressing first-and-last-mile challenges while enhancing urban accessibility. This study develops a predictive model to quantify the multifaceted societal benefits of integrated bike-transit networks, assessing impacts on environmental, economic, and health outcomes. Using open-source transit and land-use data, we estimate potential shifts from private vehicles to multimodal trips and monetize the resulting benefits, including reduced CO₂ emissions, improved air quality, and increased physical activity. Results indicate that integrating cycling can expand transit catchment areas by up to 70%, particularly improving job accessibility for low-income communities. Findings suggest that strategic investment in connected, safe cycling infrastructure and high-quality transfer nodes at transit stations is essential to maximize these benefits and foster equitable, sustainable urban mobility.

Keywords: Bike-transit integration, multimodal, first-mile/last-mile connectivity, solving accessibility issues near transit hubs, secured storage, transit-oriented development (tod), urban planning surrounding transit.

1. INTRODUCTION

The integration of cycling and public transit networks offers a sustainable solution to modern urban mobility challenges by reducing reliance on private vehicles, lowering carbon emissions, and easing traffic congestion [1,3].

By combining the flexibility of cycling with the high-capacity, long-distance reach of public transport, these integrated systems significantly improve urban accessibility, fostering more equitable transportation options for diverse populations [2,4].

Developing predictive models is essential to quantify the potential positive outcomes of these integrated networks, such as improved public health, enhanced economic activity, and reduced travel times, allowing planners to optimize infrastructure investment.

While public transit and cycling are often treated as independent modes, integrating them creates a synergistic 'first-and-last-mile' solution that significantly expands the catchment area of transit hubs, putting more opportunities within reach of residents.

Predicting the societal benefits of these combined networks is crucial for urban planners, as studies suggest a well-integrated system can increase accessibility by up to 70%, offering a competitive alternative to private car ownership that improves both public health and environmental outcome [5,7].

Proactive modeling and analysis of integrated networks are necessary to ensure that investments in, and the benefits of, bike-transit systems are distributed equitably, addressing potential disparities in infrastructure accessibility for vulnerable or underserved communities

2. EVOLUTION OF DEMAND FORECASTING MODELS

Literature traditionally began with statistical models like ARIMA (Auto-Regressive Integrated Moving Average) and exponential smoothing. These models were valued for their interpretability in predicting transit demand based on historical patterns. However, researchers have increasingly noted their limitations in capturing non-linear relationships. Recent studies have shifted toward Machine Learning (ML) approaches, such as Random Forests and XGBoost, which handle high-dimensional data better than classical statistics. In the context of Cycle-PT, these models allow for the integration of diverse variables that simple time-series models might overlook [6,8].

2.1. The Role of External Factors and Covariates

Modern literature emphasizes that cycle-based transport is highly sensitive to external shocks. Scholars frequently cite weather conditions (precipitation, temperature) and built-environment features (proximity to transit hubs, population density) as critical predictors. For instance, studies on bike-sharing systems often show a sharp decline in demand during heavy rain, a factor that must be "futurized" in any forecasting model. Furthermore, the "first mile/last-mile" theory suggests that demand for Cycle-PT is often a derivative of larger public transport schedules, meaning train or bus arrivals serve as significant lagged indicators for cycle demand [7,9].

2.2. Deep Learning and Spatial-Temporal Correlation

A significant portion of recent research focuses on Deep Learning, specifically Long Short-Term Memory (LSTM) networks and Graph Convolutional Networks (GCNs). Because Cycle-PT demand is not just temporal (when it happens) but also spatial (where it happens), researchers use these advanced architectures to capture "spatial-temporal correlation." For example, a surge in demand in a commercial district at 5:00 PM often predicts a subsequent surge in nearby residential drops-off zones [11,13]. Literature in this area suggests that hybrid models—combining CNNs for spatial features and LSTMs for temporal trends—outperform single-algorithm approaches. [12,14]

2.3. Impact of Real-Time Data and Policy

Finally, the literature explores the shift from static "long-term" planning to real-time short-term forecasting. With the advent of GPS-enabled cycles and mobile apps, researchers now utilize high-frequency data to update forecasts every 15 to 60 minutes. This shift is crucial for "rebalancing" strategies—ensuring cycles are where the users are before they even arrive. Additionally, recent papers have begun investigating the impact of urban policy and fuel price volatility on Cycle-PT demand, noting that as traditional fuel prices rise or congestion charges are implemented, the demand for more agile, cycle-based transport typically sees a measurable uptick [8,13,16]

3. USING NORTH AMERICAN CYCLE-PT DATA

3.1. Environmental Sustainability

Shifting from single-occupancy vehicles to a Cycle-PT (cycle and public transport) model can significantly reduce greenhouse gas emissions. In the United States alone, public transit already saves 37 million metric tons of CO₂ annually. By further reducing reliance on personal cars, cities can drastically lower their carbon footprint and contribute to climate goals.

The integration of cycling and public transit reduces local air pollutants and minimizes noise pollution. This leads to clearer skies, better public health outcomes, and quieter, more livable urban environments. Consequently, this shift fosters a healthier atmosphere for city residents [19,20].

Finally, adopting Cycle-PT promotes superior land use efficiency, as bicycles require far less space for parking and infrastructure than automobiles. This allows for the development of more green spaces, improves urban design, and reduces water runoff from extensive road surfaces.

3.2. Public Health and Safety

Integrating cycling into your daily commute is a powerhouse move for your physical health, offering consistent aerobic exercise that keeps your heart pumping and your body leaning. By swapping the driver's seat for a bike saddle, you significantly slash your risks of cardiovascular disease, high blood pressure, and obesity. It's essentially a workout disguised as a commute, turning "dead time" into a life-extending routine [21,22].

Beyond personal gains, a city filled with cyclists is a fundamentally safer place for everyone. When more people choose bikes or transit over cars, the sheer volume of high-speed heavy machinery on the road drops, creating a much calmer environment.

This shift in mode share is a proven recipe for reducing traffic fatalities and decreasing the frequency of dangerous crashes, making the streets friendlier for every commuter.

Finally, hitting the pavement under your own power does wonders for your headspace. Active travel is consistently linked to better mental health outcomes, providing a natural buffer against the stresses of modern life. Whether it's the fresh air or the rhythmic movement, cycling helps lower rates of depression and anxiety, ensuring you arrive at your destination feeling energized rather than drained by gridlock.

3.3. Economic Growth and Efficiency

Multimodal travel offers significant cost savings for individuals by reducing or eliminating the need for personal car ownership. By shifting to walking, cycling, or public transit, users save substantial money on fuel, maintenance, insurance, and parking fees.

For municipalities, fostering a multimodal approach presents an opportunity for infrastructure savings. Instead of investing heavily in road repairs and the expensive construction of massive park-and-ride facilities, cities can achieve greater efficiency by investing in lower-cost, high-impact cycling and pedestrian infrastructure.

Finally, promoting active commuting leads to increased productivity and long-term economic benefits. Employees who incorporate walking or biking into their daily routines tend to be healthier, resulting in higher workplace productivity and lower healthcare-related costs for cities.

3.4. Social Equity and Accessibility

Integrating cycling with public transit dramatically broadens horizons, increasing accessible job and service opportunities by up to 70% for both riders and cyclists. By transforming transit hubs into multimodal centers, cities can unlock a far more efficient network that rewards commuters with better employment prospects and essential service reach.

This synergy significantly boosts a station's "catchment area," as bicycles expand a transit stop's effective reach by 1.7 to 2.3 times compared to walking alone. This expansion is vital for residents living further from primary transit lines, turning "transit deserts" into viable commuting corridors and ensuring that proximity to a train or bus isn't the sole factor in mobility.

Beyond logistics, this integration is a powerful tool for social equity and inclusion. Targeted infrastructure in underserved or Black-majority neighborhoods helps mitigate historical disparities in transit access and road safety, while enhanced mobility allows older adults and those without car access to maintain vital social connections and remain active participants in their communities.

4. EVOLUTIONS OF DEMAND FORECASTING MODEL

Developing demand forecast models to predict the societal benefits of integrated cycling and public transit networks involves creating data-driven tools that quantify how combining these modes reduces car dependency, improves accessibility, and enhances urban sustainability. These models often combine machine learning with traditional four-step transportation planning, utilizing data on bike-sharing, transit schedules, and infrastructure to forecast future ridership and associated impacts.

4.1. Model Structure and Methodologies

Hybrid Deep Learning Models for urban mobility now leverage advanced architectures, specifically combining Spatial-Temporal Attention Graph Convolutional Networks (STAGCN) with Long Short-Term Memory (LSTM) or Informer models to optimize bicycle-transit connectivity. These hybrid approaches excel at modeling complex spatial dependencies between stations while simultaneously capturing long-sequence temporal dependencies, such as daily peak hour congestion or weekly, seasonal, and environmental trends. By fusing these techniques, operators can achieve high-accuracy demand forecasting—with some models reaching a coefficient of determination (R^2) of 0.893—which is critical for preventing supply-demand imbalances in shared bike systems.

The Four-Step Transportation Planning framework is being modernized to better accommodate active transportation by modifying traditional trip generation, distribution, modal split, and assignment steps to explicitly account for bicycle-transit integration. These adaptations allow planners to move beyond car-centric modeling, directly analyzing the connectivity of bike-and-ride (parking at stations) and bike-on-board (taking bikes on transit) scenarios. By integrating non-motorized network data into these four steps, transportation agencies can better evaluate how, for example, protected bike lanes or increased parking capacity at stations influence the shift from private car usage to combined bike-and-transit modes.

Facility Locator and Market Travel Shed Models focus on defining and optimizing the catchment areas around transit hubs, identifying potential commuters by analyzing bike accessibility within specific radii, such as 300 meters, to station entrances. These models use GIS-based network analysis to create "travel sheds," measuring the area reachable by bicycle within a set time, thereby identifying optimal locations for new bike racks, high-capacity two-tier racks, or secure modular bike lockers. This approach is crucial for promoting last-mile connectivity, ensuring that transit hubs are accessible and that secure infrastructure is available, which significantly improves user experience and increases the likelihood of transit adoption.

4.2. Identifying Key Data Inputs

Effective models for predicting shared bike and transit demand require rich, multi-source data to ensure accuracy in demand forecasting and benefit analysis. A critical component is spatial-temporal data, which includes historical usage patterns, precise GPS coordinates for origin/destination (OD) pairs, and timestamps for shared bikes and public transit.

Furthermore, the models must incorporate data regarding the built environment and infrastructure, such as bike lane connectivity, sidewalk availability, land-use density, and Point of Interest (POI) data located around transit stations.

Finally, operational and contextual data, including weather conditions (temperature, precipitation), time-of-day (peak vs. off-peak hours), and transit service frequency, are essential for capturing real-world usage fluctuations and maximizing model effectiveness.

4.3. Key Challenges in Model Development

Accurately identifying bike-to-transit transfer trips presents a significant challenge, requiring advanced techniques to distinguish them from general cycling. Utilizing KD Tree algorithms for nearest-neighbour searches and multi-stage filtering is essential to isolate valid connections from routine bike usage.

Capturing spatial-temporal heterogeneity is critical, as demand varies significantly between urban and suburban settings and across different times. This requires, according to studies such as this one from PMC, adaptive models capable of learning these shifting, complex patterns.

Data purity is often compromised by simplistic buffering, which includes non-connecting trips and reduces accuracy. Implementing rigorous filtering methods is necessary to eliminate this noisy data.

4.4. Predicting Societal Benefits

Modern modeling techniques translate predicted demand into actionable insights that highlight significant environmental and health advantages. By quantifying reductions in carbon footprints and life-cycle emissions, these models demonstrate that consistent cycling can lower daily travel-related CO₂ emissions by up to 84% compared to non-cyclists. Beyond environmental sustainability, promoting active transportation directly addresses the global health burden of physical inactivity, with potential to prevent thousands of chronic diseases and generate millions in healthcare savings annually.

Accessibility and equity are further refined through models that measure the impact of solving the "first/last-mile" problem. Integrating bikes with transit systems can expand a user's accessible opportunities by up to 70%, particularly benefiting those in underserved or car-less communities. For example, research indicates that while Black communities often have high transit usage, they frequently lack the local biking infrastructure necessary to realize these full multimodal benefits; targeted infrastructure improvements in these areas are essential to bridging the gap in access to jobs and essential services.

Finally, the economic viability of these initiatives is established through rigorous cost-benefit analysis (CBA) frameworks. Utilizing tools like the NZTA Monetized Benefits and Costs Manual (formerly the Economic Evaluation Manual), planners can determine the benefit-cost ratio (BCR) of land transport investments by monetizing factors like travel time savings, reduced healthcare costs, and lowered emissions. These evaluations consistently show positive returns, such as a 10% annual return on current bike-sharing investments, proving that sustainable transit is a robust and efficient use of public funding.

4.5. Common Tools and Techniques

The foundation of this approach relies on robust data engineering and machine learning frameworks. By utilizing Python and its powerful ecosystem, planners can streamline the entire data lifecycle. Packages like Pandas and NumPy enable the efficient manipulation of massive trip datasets, while deep learning libraries such as PyTorch or TensorFlow provide the architecture necessary to implement sophisticated neural networks that learn from historical patterns.

To account for the physical layout of the city, Geographic Information Systems (GIS) like ArcGIS and QGIS are integrated to perform advanced spatial analysis. These tools allow for the precise mapping of drainage catchments, bike lane infrastructure, and transit network connectivity. By overlaying these spatial layers with usage data, planners can identify how environmental factors and existing transit hubs influence commuter behaviour, ensuring that bike-sharing solutions are grounded in the actual geography of the urban environment.

The final layer of this methodology employs Graph Neural Networks (GNNs) to model the complex, non-linear interactions between bike stations. Unlike traditional models, GNNs treat the bike-sharing system as a dynamic graph where stations are nodes and routes are edges, capturing the "ripple effects" of traffic flow across the city. By combining these methods, planners can optimize resources-such as determining the ideal bike inventory at specific stations during peak hours-ultimately enhancing the efficiency of the entire urban transportation network.

5. MACRO-MODEL (NETWORK/ENTIRE-ROUTE)

A Macro-Model (Network/Entire-Route) for predicting the societal benefits of integrated cycling and public transit (PT) networks is a high-level, system-wide transportation forecasting approach. It models the cumulative impact of connecting cycling infrastructure with public transit, focusing on demand, accessibility, and economic/social impacts across a city or region rather than individual vehicle movements. Key components of such a macro-model include:

5.1. Model Structure and Scope

A system-wide analysis is conducted to evaluate the entire urban transit network and city-wide cycling paths, treating them as a single, integrated multimodal system. By analyzing the network in this holistic manner, planners can identify gaps, barriers, and critical connections between active transportation infrastructure and public transit. This comprehensive approach allows for the optimization of the entire transportation network rather than focusing on isolated infrastructure improvements, ensuring that cycling and public transit function together seamlessly.

To accurately reflect this integrated approach, the traditional four-step transportation model-comprising trip generation, distribution, mode split, and traffic assignment-is refined. This adaptation specifically accounts for multi-modal trips, such as Bike-and-Ride (using a bike to reach a transit stop) and Bike-on-Bus/Rail scenarios. The refined model simulates user behavior more accurately by considering factors like intersection treatments, safe routes, and transfer convenience between bicycles and public transportation, ensuring that multimodal options are adequately represented in the planning process.

Network-level modeling software, including macroscopic tools like EMME, CUBE Voyager, TransCAD, or VISUM, is employed to analyze network capacity and passenger flows across the entire city. These tools are used to "load" multimodal trips onto the network, evaluating how bike paths and transit lines interact under various scenarios. By utilizing these advanced, high-level modeling techniques, planners can simulate the impacts of infrastructure changes on traffic congestion and travel times, providing data-driven insights for developing more sustainable and efficient transportation networks

5.2. Core Functional Elements

Generalized Travel Cost (GTC) functions are employed to model user choices by incorporating monetary costs, travel time, and discomfort, often representing these factors in time-equivalent units. For integrated cycle-public transport (PT) journeys, key variables include access/egress speed, waiting times at stations, and the cost of secure bike storage. By integrating these elements, the model effectively calculates the total disutility of a trip, enabling the identification of the most efficient, user-preferred modes of travel.

Cycling significantly expands the effective catchment area of transit stations, with cycling ranges for accessing stations (e.g., 2.5–7.5 km) being roughly three to five times larger than walking (approx. 800 meters), thereby increasing the potential service area by up to twenty-five times. This expanded radius improves transit accessibility, allowing for lower-density suburban areas to be serviced efficiently and reducing the need for high-density, expensive station construction. Demand projection for "Bike-on-Bus" or "Bike-and-Ride" services utilizes this spatial expansion to forecast user numbers based on regional demographic data, infrastructure improvements (e.g., bike paths), and the availability of secure parking.

Modal shift modeling evaluates the potential to reduce private car usage by identifying shifts to combined bike-transit modes, particularly for journeys within 6–10 km. These models consider variables like time savings, cost, and convenience, showing that when the travel time advantage of cycling to a station is high compared to driving, users are more likely to switch. Effective modeling helps predict the reduction in car dependence by analyzing the "last mile" problem and identifying opportunities for, and barriers to, integrating bicycles with public transit

5.3. Societal Benefit Metrics (Outputs)

The macro-model acts as a comprehensive evaluation tool that translates physical infrastructure changes into tangible economic and social impacts. By focusing on accessibility and social equity, it measures the increased number of jobs and services that become reachable for residents in suburban areas and low-income groups. Furthermore, the model assesses how these changes facilitate improved transportation options, enhancing overall connectivity for marginalized communities.

From an environmental and health perspective, the model calculates significant reductions in CO₂, particulate matter (PM), and NO_x emissions resulting from decreased Vehicle Miles Travelled (VMT). Simultaneously, it estimates the economic benefits derived from improved public health due to increased physical activity. This includes evaluating the safety improvements of dedicated, separated bike lanes, which reduce traffic accidents and encourage active transportation.

Finally, the macro-model analyzes the broader, systemic improvements in urban mobility by modelling the reduction of car traffic on overall road network performance. By analyzing these factors together, the tool helps planners understand the direct correlation between infrastructural upgrades and a more sustainable, equitable urban environment.

5.4. Typical Scenarios Tested

Scenario A (Baseline) establishes the current status-quo infrastructure and existing public transit usage as the reference point for evaluation. This baseline represents a network often characterized by disconnected, paint-only lanes or lack of infrastructure, which serves as a control to measure the impact of proposed changes in subsequent scenarios.

Scenario B (Network Connection) focuses on closing critical gaps in the existing bike lane network, with a direct emphasis on improving connectivity to major transit hubs and employment centers. By focusing on network integration rather than isolated improvements, this scenario increases safe access to public transport for commuters, addressing the "last-mile" challenge and encouraging a shift away from private vehicular travel.

Scenario C (Complete Streets/Capacity Expansion) implements high-quality, infrastructure-heavy improvements, including physically protected lanes, secured, end-of-trip bike parking, and increased bike capacity on public transit vehicles. This scenario represents the maximum investment level, often creating a transformative impact on urban mobility that can yield societal benefits 10–25 times greater than the initial construction and maintenance costs.

In essence, this macro-model serves as a Societal Cost-Benefit Analysis (SCBA) tool that justifies investing in active travel infrastructure by demonstrating its high return on investment (often a Benefit-Cost Ratio > 1). By modeling these three scenarios, policymakers can quantify the long-term benefits of safety, health, and environmental improvements against construction costs.

Figure 1 illustrates the Macro Model flow chart, detailing the variable distributions utilized in the Monte Carlo simulation. This simulation was designed to establish a range of potential values and a 95th percentile confidence interval for Cycle-PT demand.

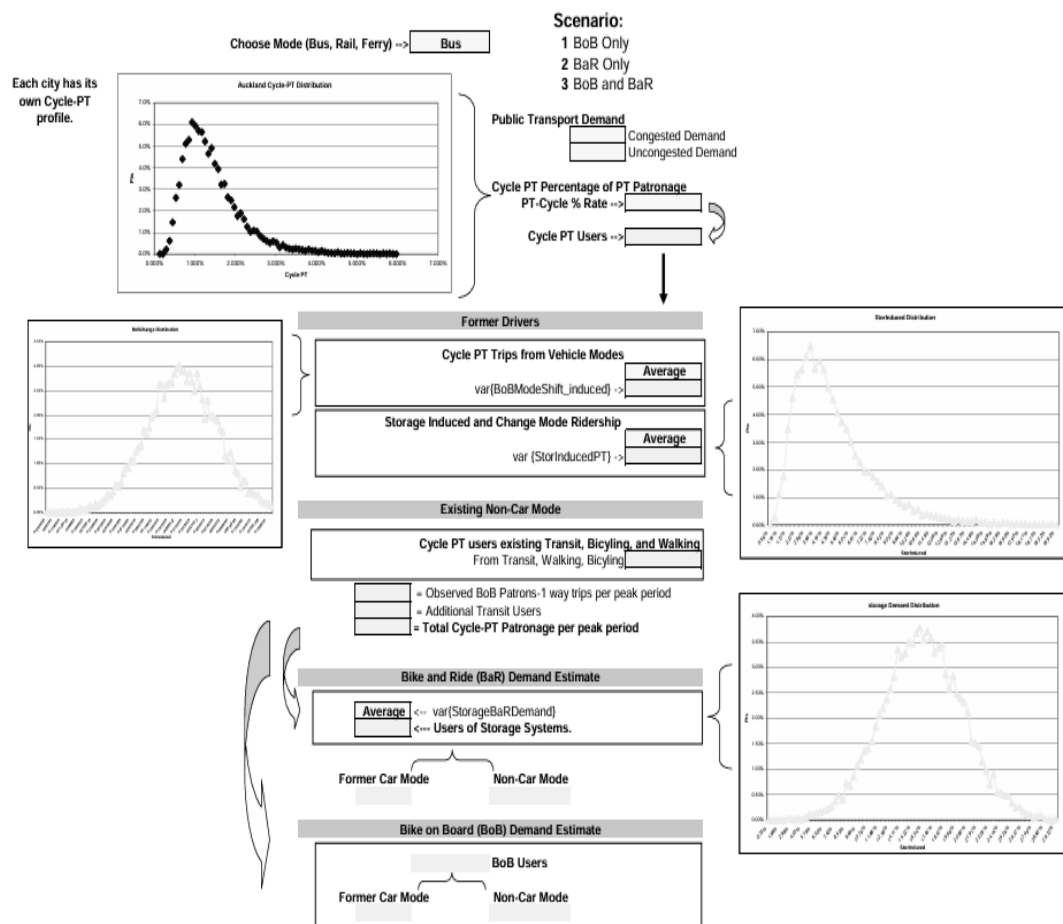


Figure 1. System Wide Cycle-PT Demand Model.

The Micro Model for Cycle-PT demand at individual stops and stations provides a granular analysis necessary for determining capacity requirements and optimizing the placement of secure storage lockers on public transport networks. While macro models assess the entire system, this micro-level approach allows for targeted, route-by-route analysis to prioritize infrastructure placement, ensuring adequate capacity for Bicycle-at-Railway (BaR) or Bike-on-Bus/Bike-on-Board (BoB) options.

Demand for this service is driven by three primary sources: users expanding their catchment area via cycling to a station, commuters bridge the gap from their destination to the station, and cyclists replacing full-journey cycling with a combined public transport trip. By utilizing existing data on boarding and alighting patrons, this methodology accurately calculates the total Cycle-PT demand (sum of BoB and BaR) based on established patronage rates for specific transport types like rail or local buses.

6. APPLICATIONS OF THE FORECAST TO NEW ZEALAND

Forecast models are heavily applied to New Zealand's unique, rugged landscape to improve the accuracy of weather predictions, particularly for high-impact, localized events. Organizations such as NIWA and MetService utilize advanced numerical weather prediction (NWP) systems, including the New Zealand Limited Area Model (NZLAM) and the 1.5 km resolution New Zealand Convective Scale Model (NZCSM), which are specifically configured to represent the country's complex topography like the Southern Alps. These models integrate data from satellites, ships, and weather stations to simulate atmospheric processes—such as convection and turbulence—over short and long time scales. Furthermore, high-resolution WRF (Weather Research and Forecasting) models are implemented to predict specific hazards such as tropical cyclones, fire weather risks, flooding, and sea-level storm surges, providing essential data for emergency management.

Table 1. Annual Cycle-PT Trip Demand.

Location	Mode	Current PT Patronage (approximate per annum)	Annual Cycle-PT Trip Demand	
			Likely	Stretch Goal
Auckland	Bus	45 million	450,000	1,350,000
Auckland	Train & Ferry	7 million	210,000	420,000
Wellington	Bus	25 million	250,000	750,000
Wellington	Train & Ferry	10 million	300,000	600,000
Tauranga	Bus	0.5 million	5,000	15,000
Rotorua	Bus	0.5 million	5,000	15,000
Dunedin	Bus	2 million	20,000	60,000
Christchurch	Bus	15 million	225,000	450,000
Hamilton	Bus	2 million	20,000	60,000
Invercargill	Bus	0.5 million	5,000	15,000
Total			1,415,000	3,735,000

Beyond immediate operational forecasting, these models are increasingly used for long-term planning, climate change attribution, and specialized industry applications. Researchers have developed AI-powered and high-resolution modeling systems to quickly analyze extreme weather events, such as the 2021 Canterbury and Buller floods, to quantify how anthropogenic climate change has increased their severity. Additionally, sub-seasonal to seasonal models like NIWA35 and the Seasonal Climate Outlook provide long-range forecasts, assisting sectors like agriculture and energy in planning for drought, rainfall, and temperature variations.

These applications ensure that, despite the challenges of simulating atmospheric conditions in data-sparse regions, New Zealand can achieve improved accuracy in predicting hazardous weather conditions. Table 1 shows that there is current demand for approximately 1.5 million Cycle-PT trips per annum in New Zealand.

Current forecasts indicate that Cycle-PT (the integration of cycling with public transport) demand closely follows trends in general public transport patronage, with high-growth scenarios in PT patronage driving a similar rise in Cycle-PT usage. As substantial investment continues to flow into Auckland's passenger rail networks, such as the City Rail Link and electrification improvements, the resulting increase in rail patronage is expected to drive up demand for complementary cycling infrastructure, such as secure cycle storage at hubs.

Given this anticipated significant surge in public transport patronage over the coming decade, there is high potential for the actual Cycle-PT demand in the Auckland region to exceed the current forecast of 660,000 annual trips. As Auckland shifts towards more integrated transport solutions, the enhanced convenience and increased capacity of the rail system provide substantial upside potential for cycling-public transport uptake, allowing the city to move beyond initial conservative estimate.

7. PROBLEM FORMULATION

The problem formulation for predicting the societal benefits of integrated cycling and public transit networks centers on modeling the synergistic, rather than merely additive, effects of these modes to overcome the "first-and-last mile" challenge and replace private car trips. Key challenges include mapping the uneven distribution of benefits across spatial contexts (urban vs. suburban), quantifying non-linear impacts on health and emissions, and accounting for the behavioral shift from, or towards, transit as accessibility changes. The objective is to establish a framework that captures how improved infrastructure, such as secure parking at stations or bike-friendly transit vehicles, translates into increased patronage and reduced traffic congestion.

A robust formulation requires identifying specific metrics, such as accessibility to jobs and services, reduction in greenhouse gases, health benefits from physical activity, and increased safety for cyclists. It must account for various factors that influence the success of this integration, including the quality of public transport, the connectivity of the cycling network, and the built environment's ability to facilitate safe transfers. Crucially, the problem formulation needs to address social equity, examining if the benefits are accessible to all income levels and demographic groups or if the improvements primarily serve already advantaged, high-income areas.

Finally, the predictive model must incorporate dynamic, stakeholder-oriented, or system-dynamics approaches to account for feedback loops and long-term behavioral changes. This involves translating qualitative, "soft" factors like user safety perception and convenience into quantitative, actionable data for planners, balancing user-centric preferences with system-centric, sustainable, and equitable outcomes. The goal is to move beyond mere modeling to provide actionable insights for decision-makers to implement better infrastructure and policies. The integration of cycling and public transit networks offers significant, multifaceted societal benefits, primarily by expanding the accessibility and reach of mass transit systems. By enabling bicycles to cover the "first/last mile" between stations and final destinations, these networks significantly reduce total travel times for commuters, with some studies showing an expansion of accessible opportunities by up to 70%. This synergy allows cyclists to travel further and faster by combining, for instance, a 5 km bike ride with a high-speed train, making public transit a more viable alternative to private cars for medium-to-long-distance trips.

8. PROBLEM SOLUTION

Predicting the societal benefits of integrated cycling and public transit networks poses a significant data-modeling challenge, as it requires analyzing complex, multi-modal travel behaviors rather than just individual transit or cycling trends. A primary issue is the lack of standardized, real-time data that captures the "first mile/last-mile" connection between bike-sharing, private bicycles, and transit stops, making it difficult to accurately calculate potential reductions in car usage or greenhouse gas emissions. Furthermore, predicting the equitable distribution of these benefits is difficult, as infrastructure improvements often benefit already connected areas, neglecting underserved communities and potentially exacerbating spatial accessibility inequalities.

To solve these challenges, advanced modeling techniques that incorporate high-resolution spatial data-such as GTFS (General Transit Feed Specification) real-time data and GPS tracking of bike-sharing trips-are essential to simulate realistic, dynamic, and multi-modal travel patterns. A proactive solution involves developing predictive frameworks that simulate "what-if" scenarios, enabling planners to measure how improvements in infrastructure, such as protected bike lanes near transit hubs, affect the catchments area and increase overall public transit ridership. These models can quantify societal benefits, such as decreased travel times and reduced vehicular pollution.

To ensure that the predicted benefits are maximized across all demographics, solutions must focus on integrating equitable policy with infrastructure development. This includes prioritizing investment in cycling infrastructure in low-income or marginalized communities to reduce the "biking accessibility gap" and ensure that the benefits of integration are not limited to affluent areas. Using digital tools and AI to monitor bike-transit usage allows authorities to dynamically adjust services to meet real-time demand, while also enabling the integration of e-bikes for longer, more efficient trips, ultimately creating a more robust, accessible, and sustainable urban transport system.

9. CONCLUSION

The integration of cycling and public transit represents a transformative shift in urban planning, offering a sustainable alternative to car dependency. By bridging the "first and last mile" gap, these multimodal networks significantly expand the reach of existing infrastructure, making low-carbon travel viable for a broader demographic. The primary societal benefit lies in the creation of a more resilient and equitable transportation system that reduces congestion, lowers carbon emissions, and fosters a more efficient use of urban space.

Beyond logistical efficiency, the fusion of these two modes yields profound public health and economic dividends. Increased physical activity from cycling, combined with the reduced air pollution of transit-oriented development, directly correlates with lower rates of chronic illness and improved mental well-being. Furthermore, such systems enhance socio-economic mobility by providing affordable access to employment hubs, education, and essential services for those without private vehicles, effectively narrowing the gap between underserved communities and economic opportunities.

Ultimately, predicting the long-term success of these integrated networks requires a commitment to data-driven design and continuous policy support. As cities evolve, the focus must remain on user safety and seamless connectivity-such as secure bike storage at rail stations and protected lanes that lead directly to transit hubs. When executed effectively, the synergy between cycling and public transport does more than just move people; it cultivates more livable, vibrant, and human-centric cities prepared for the environmental challenges of the future.

Acknowledgement:

The author(s) wish to express their sincere gratitude to Nieves U. Ganiron and Tommy U. Ganiron for their invaluable technical assistance and insightful discussions throughout the development of this work. Their expertise significantly contributed to the analysis of integrated cycling and public transit systems, helping to clarify the project's focus on societal benefits.

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