

Bike Sharing:

A Review of the Literature

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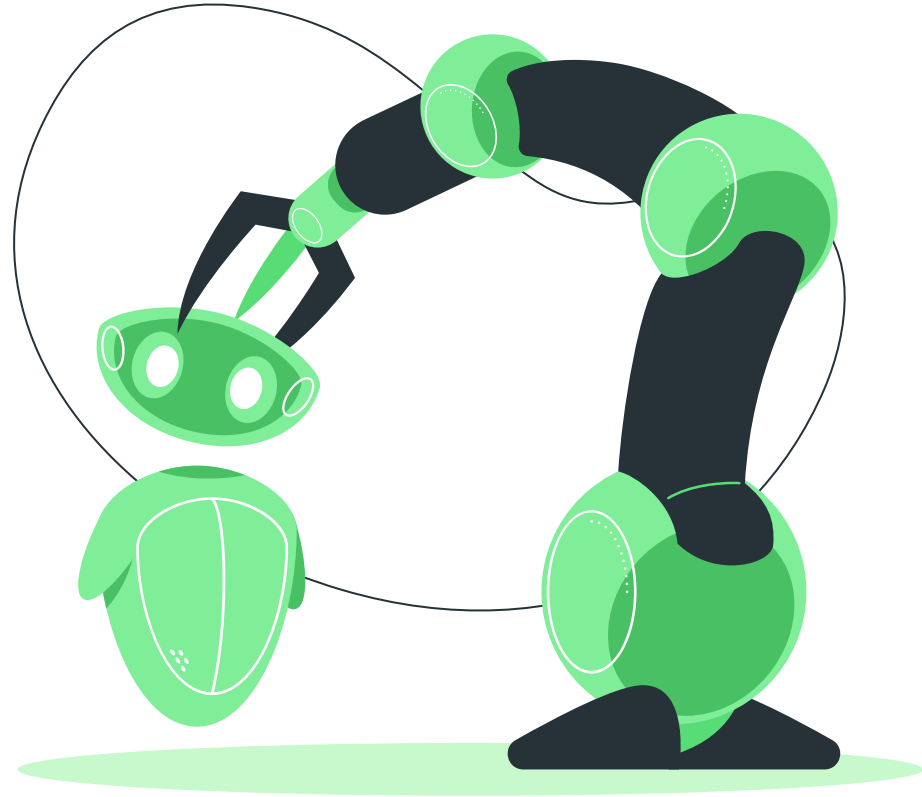


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1

Introduction


Introduction

Bike sharing is the shared use of a bicycle fleet.

Some benefits of a bike sharing system (BSS) are:

- Enhanced mobility
- Lower implementation cost compared to other modes of transportation
- Reduced traffic congestion, fuel consumption and greenhouse gas emissions
- Improved public health
- Increased environmental awareness.

Their primary travel purposes are:


- Work-related activities
 - Leisure across different age groups
 - Residential consumption
- 

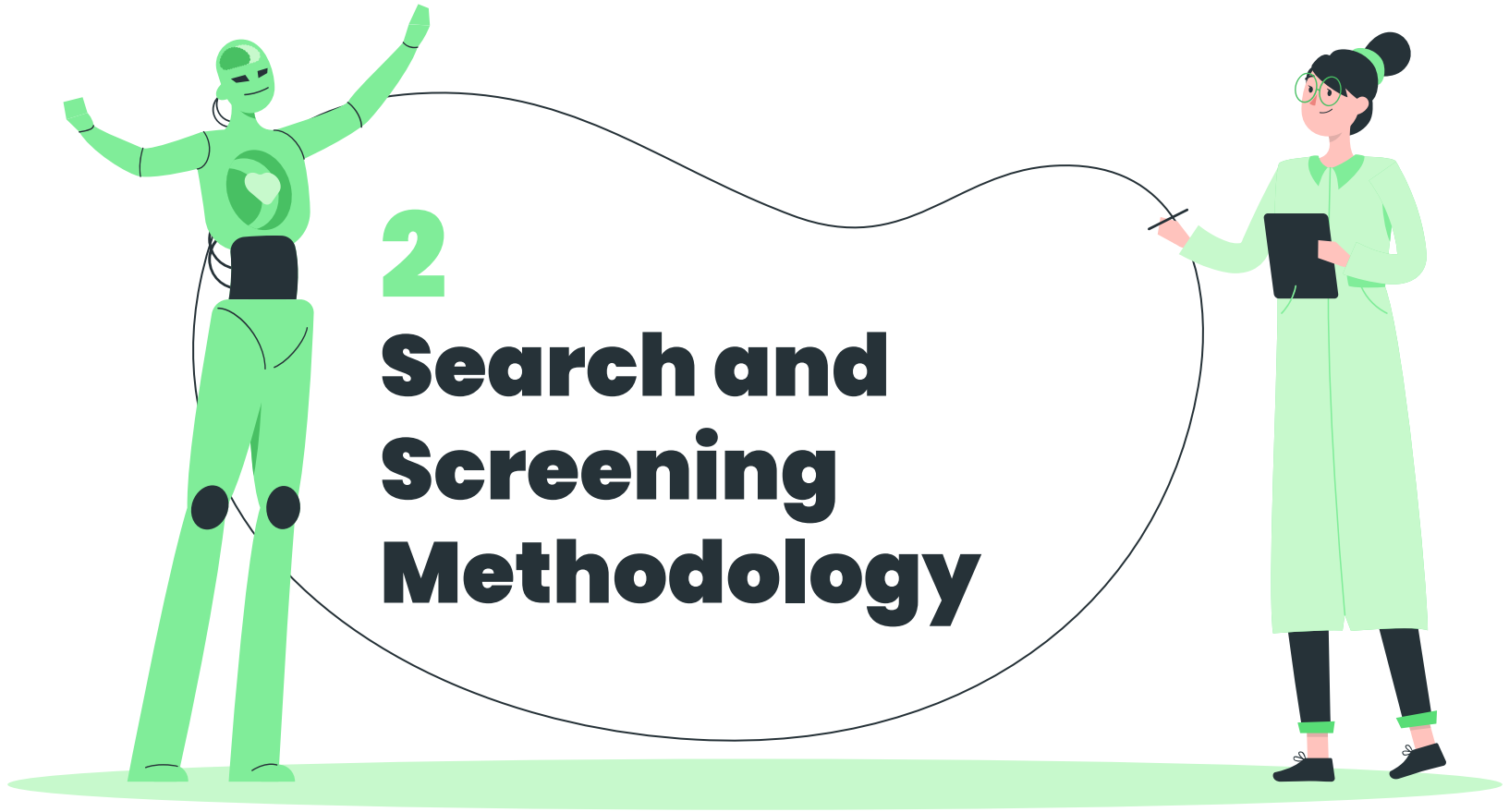
Bike Sharing Systems

There are two types of BSSs:

- Standard BSSs allow users to rent a bike from a docking station and return it to another
- Dockless BSSs provide rapid and flexible mobility, as bikes can be found and parked anywhere

The literature can be divided into two domains:

- The analysis, review and synthesis of different BSSs
 - Comparison of different rebalancing models for the bike repositioning problem
- 



Methodology

Complete Search

Search based on all the literature



Advanced Search

Search with more specific parameters

Screening Process

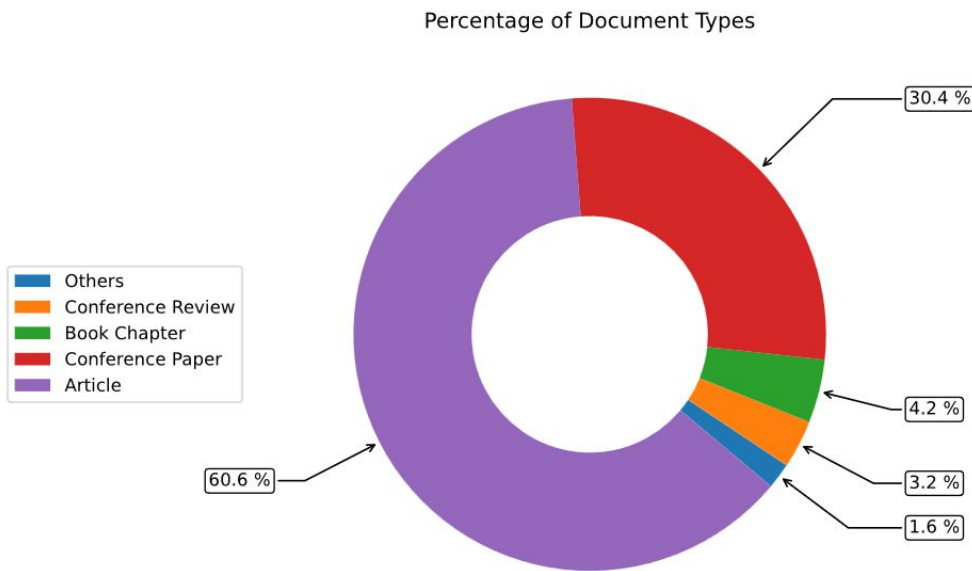
Check the resulting papers by relevance



Complete Search

TITLE-ABS-KEY (bike-sharing OR bikesharing OR "bike sharing") AND LANGUAGE (english) AND PUBYEAR < 2025

Resulting in 2,738 papers

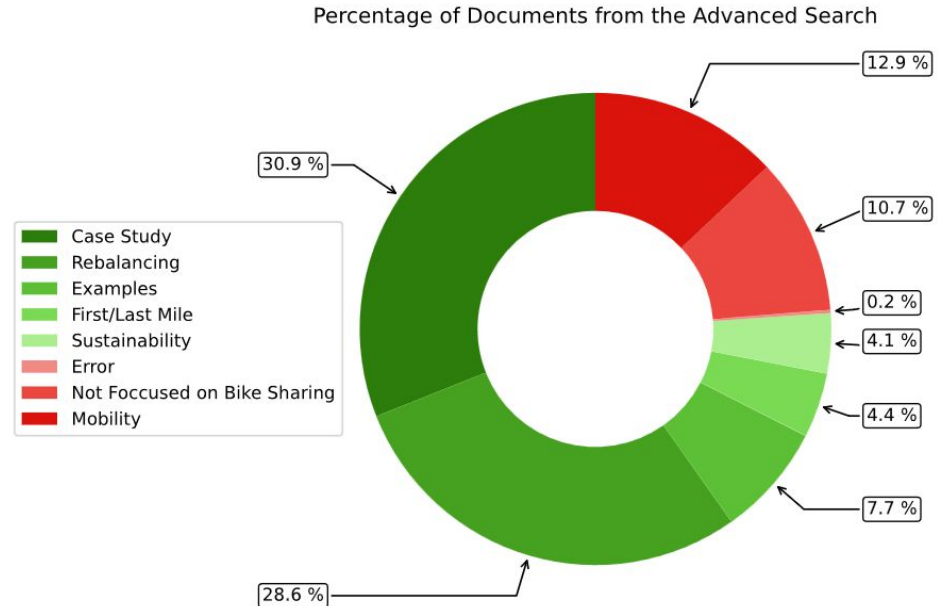


Advanced Search

TITLE-ABS-KEY (bike-sharing OR bikesharing OR "bike sharing") AND SRCTYPE (j) AND LANGUAGE (english) AND PUBYEAR < 2025

Resulting in 1,800 papers

Resulted in 1,316 Included papers



Inclusion Themes

Case Study: Studies on the impacts, problems, etc, and analysis of different types of BSSs.

Rebalancing: Studies on prediction models and algorithms for BSSs.

Examples: Studies on BSSs in a specific city, of their failure or success.

First/Last Mile: Studies on the correlation between BSSs and any transportation method.

Sustainability: Studies on the green impact of BSSs.



Excluded Conditions

Mobility: Studies on mobility as a service problems and mobility sharing.

Not Focused on Bike Sharing: Studies that use BSSs as an example or use data from a BSS.

Error: Any Erratum study.

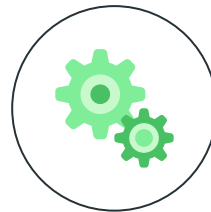


Screening Process



Impact of the

$$impact = \frac{number_of_citations}{number_of_years_since_publication}$$



**500 Selected
Papers**



Generations

1

White Bikes

Introduced in Amsterdam
1965-1994

2

Coin-Deposit

Introduced in Copenhagen
1995-1997

3

IT-Based

Introduced in Rennes
1998-2009

4

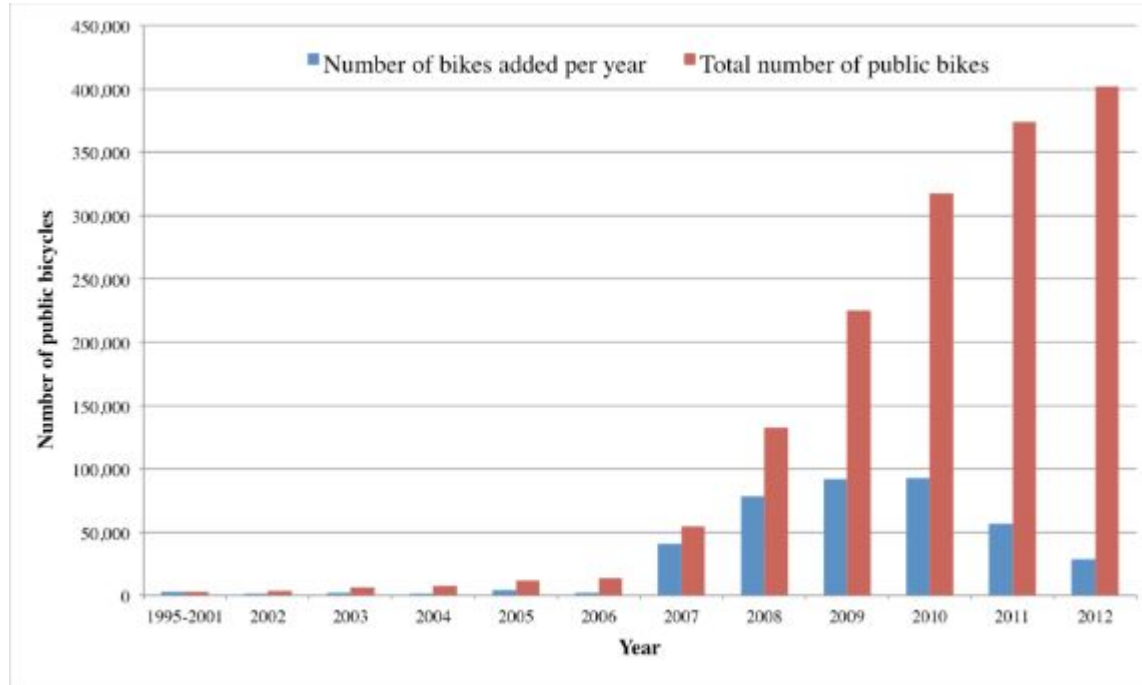
Demand-Responsiv

2009-2015

Generations

White Bikes	Coin-Deposit	IT-Based	Demand
Free bikes	Unlocked with a coin	Innovative technologies	Electric bikes
Bikes randomly parked	Docking stations	Bikes with locks	More efficient stations and a redistribution system
Bikes thrown into channels and stolen	Continued theft and vandalism	Anti-theft program	GPS to reduce theft and vandalism

Evolution – Until 2012



The Fifth Generation

The Dockless Bike Sharing
Systems

The Dockless Station:

- Easier for a user to pick up or drop off a bike
- Bikes can be parked anywhere and the users can access them instantly
- A reallocation team is needed to gather all randomly parked bikes into different parking areas

Increased cycling in urban areas and significantly reduced car usage and carbon emissions.

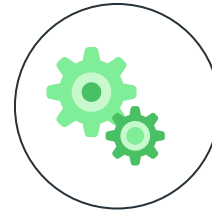
Modal Shift

Defined as, mode of transportation substituted.



Positive

Replacing private vehicles or taxis



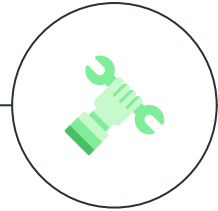
Negative

Replacing schedule based transportation, like bus, or walking

Parked Bikes Problem

In Dockless BSSs

Repositioning randomly parked bikes into different areas

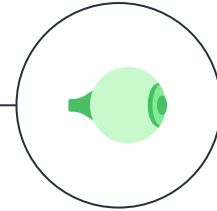


In Standard BSSs

Relocating existing or new bikes into stations

Bike Repositioning Problem

Calculation of ideal truck routes at each station



Demand Prediction

Prediction of the number of bikes left in a station at a certain point in time

Demand Prediction

1

City-Level

Focused on predicting the total bike usage for an entire city

Doesn't utilize all available data

2

Cluster-Level

Focused on grouping stations into clusters

Clusters based on geographical locations or temporal demand patterns

3

Station-Level

Focused on predicting demand for each station

Much more challenging to implement

Demand Prediction Models

	Name	# Papers	Metrics
LSTM	Long Short-Term Memory	11	RMSE, MAE, MAPE, MSE, PCC, SMAPE
ARIMA	Autoregressive Integrated Moving Average	11	RMSE, MAE, MAPE, R ² , RMSEr, MAER, RMSLE, ER
HA	Historical Average	9	RMSE, MAE, MAPE, R ² , PCC, RMSLE, ER
RF	Random Forest Algorithm	9	RMSE, MAE, R ² , CV, MAE, RMSE, RMSLE
LR	Linear Regression	9	RMSE, MAE, R ² , CV, RMSEr, MAER

Demand Prediction Models

	Name	# Papers	Metrics
XGBoost	Extreme Gradient Boosting	6	RMSE, MAE, MAPE, PCC
KNN	K-Nearest Neighbours	5	RMSE, MAE, MAPE, R ²
SVM	Support Vector Machine	5	RMSE, MAE, MAPE, R ² , CV
ANN	Artificial Neural Network	4	RMSE, MAE, MAPE, MAE, RMSE, RMSLE
RNN	Recurrent Neural Network	4	RMSE, MAE, MAPE, R ² , PCC, SMAPE, MSE

Models

ANN

One of the most common, but doesn't account for temporal dependencies

RNN

Can overcome the Limitation, but it is still not fit for the time series data

LSTM

Solves the Problem, can be used with neural networks and different layers

SVM

Developed to seek the finest hyperplane classification of data

ARIMA

Commonly used for forecasting time series, applied in traffic prediction

HA

Uses historical average data for prediction

RF

Used to overcome the problem of a model having large amount of predictors

LR

The simplest model, taking one or more input features, and outputting one value

XGBoost

Based on gradient boosted decision tree

KNN

Predicts based on it's K closes neighbours

Metrics

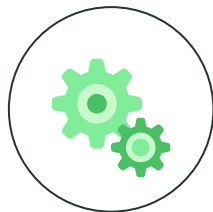


MSE and RMSE

$$MSE = \frac{1}{n} \sum_{i=1}^n (x_i - y_i)^2$$

$$RMSE = \sqrt{\frac{1}{n} \sum_{i=1}^n (x_i - y_i)^2}$$

Mostly applied when large errors must be penalized



MAE and MAPE

$$MAE = \frac{1}{n} \sum_{i=1}^n |x_i - y_i|$$

$$MAPE = \frac{1}{n} \sum_{i=1}^n \left| \frac{x_i - y_i}{x_i} \right| \times 100$$

Used when all the errors are treated equally

Infrastructure

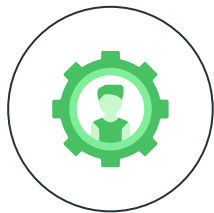
ESs incorporate more than just bikes, stations, trucks and demand prediction models.

The Infrastructure:

- Dashboards with various information on bikes, stations, users, trips, etc.
- Communicating bikes suggest a service oriented architecture
- Database to save all the information, including the messages exchanged
- Mobile application for the user to rent a bike
- Reallocation application for the redistributor to know where to load and unload the bikes



Summary



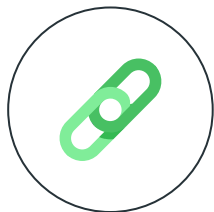
Screening Process

From 2700 searched to 500 screened papers



Literature Review

Five generations of BSSs were devised and studied



Reallocation Problem

Divided into two different categories: BRP and DP



Demand Prediction

Different models and metrics were analysed



Thanks!

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