

Planning for nodes, places, and people in Flanders and Brussels: An empirical railway station assessment tool for strategic decision-making

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Abstract: Against the backdrop of current policy discussions in Flanders dealing with urban development schemes for strategic railway stations, this paper develops an empirical railway station assessment tool. We build on the node-place modeling literature, and more specifically on the tradition of empirical station assessment models that has emerged from it. First, we propose a number of methodological contributions in which we aim to improve the analytical strength of some standard node-place parameters, to expand the model with a user-based accessibility account and to broaden the appraisal of a station's accessibility with a temporal component. Second, we apply the conceptual model to Flanders and the Brussels-Capital Region (Belgium). Drawing on factor and cluster analysis, we produce two intelligible station typologies for both the node-place and user-based data. Both typologies are interpreted and complemented with station-specific rose diagrams, summarizing a station's accessibility profile. These diagrams inform about station-specific accessibility characteristics, some of which are not captured by conventional node-place analyses. Lastly, we elaborate on five exemplary cases and illustrate what the results of these analyses may mean for planning practice.

Keywords: Accessibility, railway stations, node-place model, integrated planning, planning support tool, Flanders

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1 Introduction

1.1 Flanders in 2050: an ambitious spatial development program organized around strategic railway nodes

“In 2050, every citizen of Flanders can travel easily each day. We will have organized our space in such a way that the need to travel around is reduced. More Flemish citizens can travel in a sustainable manner. They will take the bicycle or train to commute and leave the car in the garage more often.”

Flemish Government (2017, p. 23, own translation)

The regional government of Flanders (the northern, Dutch-speaking part of Belgium) has recently engaged itself to develop a “metropolis Flanders,” an urbanized region that “ought to be large and efficient enough to position itself successfully in the urban economic network of the north-western European delta” (Boussauw et al., 2018, p. 8). It is clear that high levels of both external and internal connectivity are paramount for meeting this objective. The region nonetheless experiences a variety of related pressures that may impede achieving this objective, including heavy road congestion, air pollution and landscape fragmentation. All of these pressures are strongly associated with Flanders’ highly peri-urbanized and suburbanized spatial structure. Today, this diffuse urbanity (see Figure 1) hinges on, and is mutually enhanced by, a mobility system that is dominated by car use and ownership (Blondia, 2014; Fransen et al., 2015). Although enhancing collective transport is often proposed as an important solution to the mobility problem, land use and public transport developments are rarely successfully integrated in Flanders. A major reason for this is that the critical mass to organize a well-functioning public transport system is missing in most locations given a “nebular” settlement morphology (Blondia & De Deyn, 2012). Furthermore, policy levers in the policy realms of spatial planning and mobility are fragmented across various political-administrative levels, often hindering effective integration and coordination (De Vos & Witlox, 2013; Boussauw & Boelens, 2015).

Against this backdrop, the Flemish Government recently put forward an ambitious outlook on the future development of the built (and unbuilt) environment in Flanders by 2050 in the preparatory documents for the new Flemish Spatial Policy Plan (“Beleidsplan Ruimte Vlaanderen,” hereafter BRV). In 2012, a BRV Green Paper was released, followed by a BRV White Paper in November 2016. Twenty years after the approval of the first comprehensive Spatial Structure plan for Flanders, these policy papers put forward a renewed mid- to long-term vision extending most of the earlier spatial planning principles, albeit with a more strongly pronounced focus on railway-based accessibility as the backbone for future spatial developments. This strategic vision was recently approved by the Flemish Government (2018) and is being translated in a series of operational frameworks for implementation, of which one bears specific relevance in the context of this research: the “spatial backbone” policy paper.¹

A central objective of this policy paper is to designate strategic public transport nodes which have the highest potential for the allocation of additional urban development. This potential is determined by (1) the extent to which a location is accessible by public transport, and (2) the extent to which jobs, residents and amenities are present. Both criteria have recently been operationalized and mapped for Flanders and Brussels by Verachtert et al. (2016). Drawing on this study, the policy paper puts forward a conceptual typology of railway stations, discerning four types: international nodes, metropolitan nodes, urban-regional nodes, and rural-regional nodes. The first two types of nodes are defined based on a strict lower threshold with respect to both criteria, and unsurprisingly include the largest railway stations of the central and very urbanized area of Flanders alongside the equally urbanized corridors in the directions of Bruges and the Kortrijk-Lille and Maastricht-Aachen-Hasselt-Liège conurbations. Referring to

¹ We draw on the draft version of this framework (June 2017).

van Meeteren et al. (2015), the policy paper argues that these “strategic metropolitan regions” constitute a coherent metropolitan labor and consumer market with a strong presence of internationally competitive knowledge-oriented economic activities. The railway network connecting these international and metropolitan nodes is coined the “metropolitan transport system” (MTS, see Figure 1). Drawing on a flanking policy goal of increasing agglomeration economies in the Flemish polycentric region, the MTS network is put forward as Flanders’ spatial backbone for future economic development. It is furthermore stated that additional strategic railway nodes and growth areas should be identified along these MTS corridors, both in urban and rural areas, which translates into the two other railway station types mentioned: urban-regional and rural-regional nodes. While the policy documents clearly lay out thoughtful and detailed efforts for defining and selecting the international and metropolitan nodes, the interpretation of the two other types of strategic nodes remains opaque. This is because the exact definition and operationalization is outsourced to the urban-regional governance level. The consultative bodies operating on this smaller scale are expected to select these additional strategic nodes by stimulating cooperation between representatives of the municipalities, the public transport operators, the Flemish Government, and other parties involved.

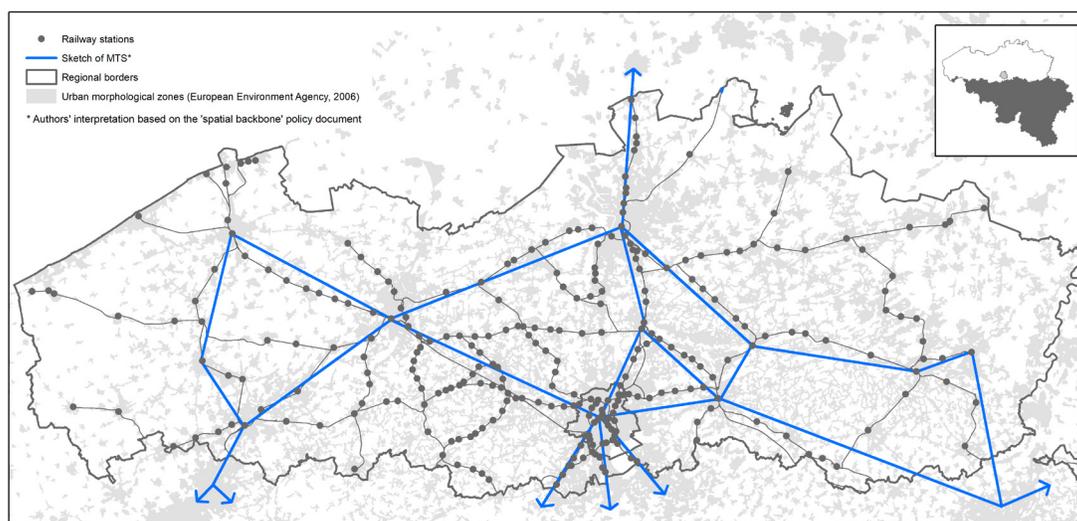


Figure 1. Flanders, its urban morphology and railway network

1.2 Strategies for stations: Empirical assessment models

Against the backdrop of this quest to identify strategic railway stations and their differentiated development opportunities, the objective of this paper is to conceptualize, operationalize and analyze a more comprehensive set of empirical parameters that may prove relevant when developing railway station typologies on the regional scale. This objective resonates with the literature on “node-place” modeling (Bertolini & Spit, 1998; Zweedijk & Serlie, 1998; Bertolini, 1999). In its most basic guise, the node-place model is an analytical framework in which a railway station (area) is operationalized as both a “node” within transport networks and as a “place” within settlement structures. The model allows to describe the relationship between both characteristics for groups of stations in a network and allows to classify these stations into empirically informed typologies (Peek, 2006; Chorus & Bertolini, 2011).

During the last two decades, the node-place model has been extensively applied and modified in different geographical contexts, both within academic and consultancy contexts. These models range

from empirical node-place analyses and station typologies to conceptual models in which the underlying mechanisms of node-place interactions are examined by means of, for example, actor analysis (see Peek, 2006, for a series of examples in the Dutch context). Of particular interest here are the type of empirical assessment models which serve the dual purpose of generating station typologies and visualizing the performance of stations on the different node and place criteria in order to allow visual comparisons between stations. These visualizations generally take the shape of polar graphs in which the performance of stations for a set of criteria is plotted along scaled axes with a common origin. Figure 2 provides a (non-exhaustive) overview of these type of node-place models, some of which include additional assessment criteria besides the standard node and place characteristics. The “kite model” (Stadsregio Arnhem Nijmegen, 2011), for example, includes a dimension combining transit ridership and the presence of services at the station (waiting rooms, shops, etc.). In a similar vein, the “node-place-experience” model (Groenendijk, Rezaei, & Homem de Almeida Correia, 2018) adds indicators reflecting the traveler’s experience at the station in terms of comfort (Wi-Fi, sheltered waiting etc.), ambient elements (type of architecture), and personnel presence. Vale, Viana, and Pereira (2018) on the other hand extend the model with a “design” dimension, in line with the commonly cited 3D’s (Cervero & Kockelman, 1997) within transit-oriented development literature. “Design” here refers to the walkability of the built environment in the station area. The web diagram introduced by Singh, Fard, Zuidgeest, Brussel, and van Maarseveen (2018) also measures walkability and bikeability of the station area, along with other extra dimensions such as “user-friendliness” of the station (in terms of, e.g., presence of information displays) and the “passenger load” or capacity utilization of the transit system. And finally, the “butterfly model” (Province of North Holland Noord-Holland and Deltametropool Association, 2013) adds a “proximity” dimension, reflecting the distance of the station to the nearest urbanized settlement.

Depending on the context in which these models were developed and applied, the way in which the models are conceptualized and operationalized varies (Peek, 2006). However, irrespective of the exact criteria included, the shared objective of these assessment models remains to empirically inform policy discussions dealing with the identification of differentiated development opportunities for railway stations from a regional perspective.

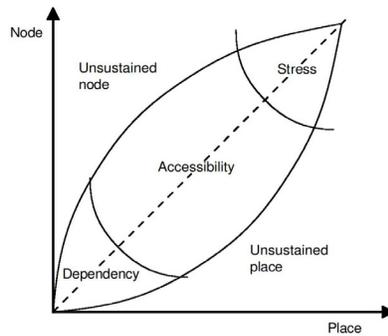
1.3 Research objectives

The research presented here has a double objective. First, there is a methodological objective in that we further develop this strand of research by paying attention to three main considerations which will be further elaborated in section 2.1: i) improving the analytical strength of some conventional node and place indicators; ii) incorporating information about the people who make use of the station, which reflects the actual demand for accessibility to and from each railway station; and iii) including temporal variations in public transport accessibility in the model. In addition to these methodological refinements to the literature, there is also, second, an empirical and related policy-support objective in that we apply the model to the case of Flanders and Brussels, and this in the broad spirit of the approved BRV strategic vision detailed above.

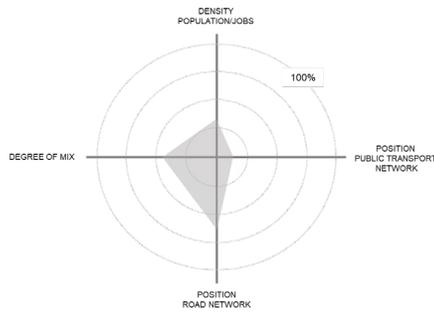
The remainder of this paper is structured as follows. The specific research objectives are described in section two, and this along with a detailed account of our methodology. Section three then reports on the main findings, while the fourth section deepens our insights into the practical relevance of the results by discussing a selection of exemplary cases. The conclusion reviews our major findings and reflects on avenues for further research.

Node-place model

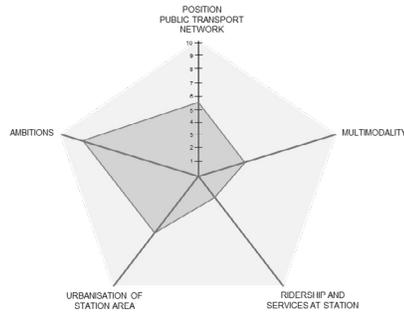
(Bertolini 1999)



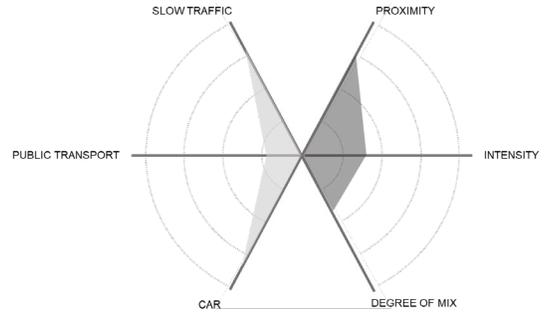
Node-place diagram
after Atelier Zuidvleugel (2006)



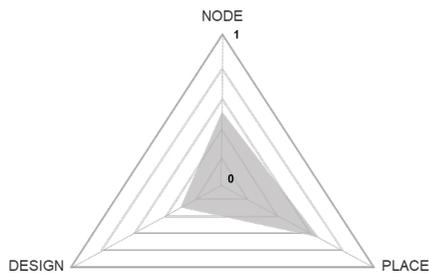
Kite model
after Stadsregio Arnhem
Nijmegen (2011)



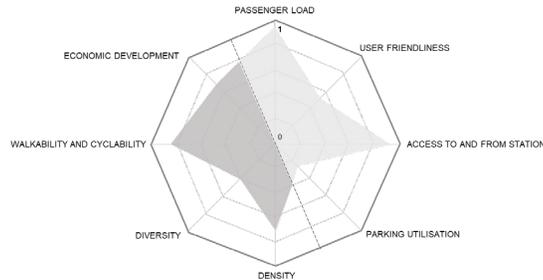
Butterfly model
after Provincie Noord-Holland and
Vereniging Deltametropool (2013)



Extended node-place model
after Vale et al. (2018)



Web diagram
after Singh et al. (2018)



Node-Place-experience model
after Groenendijk et al. (2018)

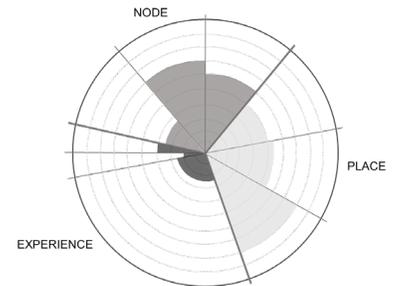


Figure 2. Overview of empirical station assessment models

2 Methodology

2.1 A modified station assessment model for Flanders and Brussels

In light of the policy-support objective of this paper, an assessment model for the case of railway stations in Flanders and Brussels requires a focus on node and place characteristics. Site-specific indicators regarding facilities or user experiences (presence of kiosks, sheltered waiting and the like) are deemed less relevant in the context of this research. In this way, the “butterfly model” developed and operationalized for all railway stations in the Dutch Province of North Holland offers a good starting point. This model was recently also modified and applied to the case of the Brussels Regional Express Network by Caset, Vale, and Viana (2018). The original butterfly model is composed of two “wings”: a node wing (on the left hand side), quantifying the accessibility of the station by bike, public transport and car; and a place wing (on the right hand side), quantifying the proximity of the station to the urban core, the number of inhabitants, jobs and visitors of nearby attractions and the functional mix of jobs and inhabitants. The model qualifies as a location-based accessibility instrument because it quantifies accessibility characteristics of a location (Geurs, 2006). The node and place characteristics included in the model furthermore capture two of the accessibility components discerned by Geurs (2006): the transport and land-use components.

However, Giannopoulos and Boulougaris (1989) among others contend that the notion of accessibility to and from railway stations extends well beyond these supply-side characteristics, since the accessibility of railway stations is also related to temporal constraints and individual needs and capabilities of its users. In Geurs’ (2006) framework, temporal constraints relate to differences in travel time and cost depending on time of the day or day of the week, whereas the individual component accounts for stratifications of the population under scrutiny (such as age or income groups). In the case of railway stations, the temporal component mainly relates to the transport component (temporal variations in public or other transport services). The individual component however requires new information to be added to the empirical assessment models discussed above. It requires relevant user- or demand-specific information which may improve particular insights about a station’s functioning in the railway network. As a corollary, including all four accessibility components may render a more comprehensive and diversified account of a station’s level of accessibility, both from the perspective of the node and the place dimension, but also from the perspective of its users.

The assessment model that resulted from these considerations takes the shape of a rose diagram (see Figure 3). Below, the structure of the diagram is explained, after which the operationalization of the dimensions and indicators is detailed. In the process, we will also discuss where we aim to improve the analytical strength of certain indicators.

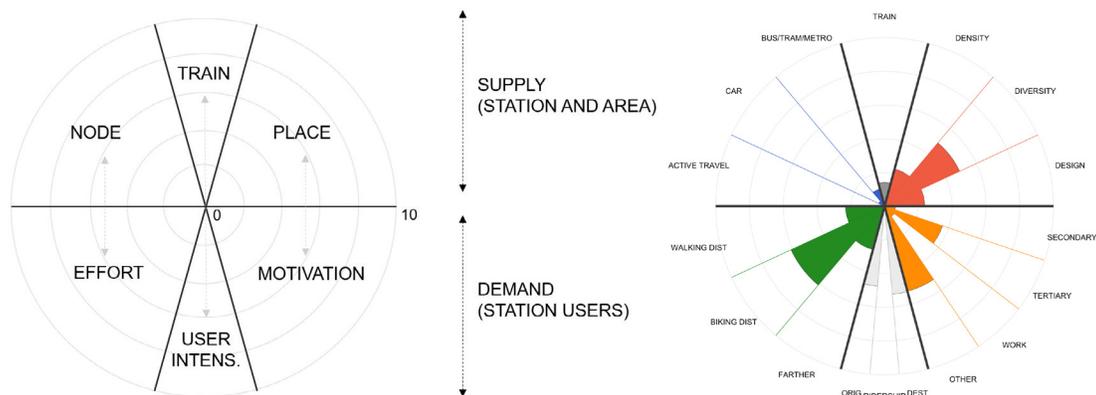


Figure 3. A station area assessment model for Flanders and Brussels: overall structure (left) and dimensions (right)

The lower part of Figure 3 presents a demand-side perspective to each station’s accessibility by visualizing information about the users of a station. Three fields are discerned: effort, ridership and motivation. The first field captures how far people live from the station they use as their origin station, and therefore relates to the effort it takes for people to reach their station of origin. This field relates to the node field in that opportunities to reach a particular station by car or by public transport (reflected in the node dimensions) arguably are more important when most people live far away from their origin station, while for example the bike parking capacity of a station is likely to be more instrumental when most people live within a walkable or bikeable radius from the station. The second field, ridership, reflects the frequency of passengers boarding on a regular working day (the actual ridership) and the extent to which the station functions as an origin or a destination station. Just as with the relation between the effort and node fields, the performance on the ridership field may be matched against that of its opposed train field (confronting supply with demand). In a similar vein, the last field, motivation, relates to the place field, as it depicts the different motives (education, work or other) of the people who employ a particular station as their destination station.

Each indicator is unity-based normalized to vary between 0 and 10. Then, for each dimension, a multi-criteria analysis is conducted in which all normalized indicator scores are summarized and again normalized per dimension. The visualized scale thus always varies between 0 and 10, and there will always be at least one station scoring 0 and another scoring 10 for a certain dimension or indicator. A descriptive code is given to each indicator detailing its field (N for node, including the train indicators, P for place, and PP for people), followed by its dimension (CA for car, AT for active travel etc.) and an indicator code (freq for frequency etc.), along with the percentage of missing values (MV). Calculations are done for all railway stations (287) with a weekday service in Flanders and Brussels.

2.2 Operationalization

2.2.1 Node dimension

Table 1. Indicators of the node dimension

Code	Indicator description	Source (year)	MV (%)
<i>ACTIVE TRAVEL</i>			
N_AT_park_f	Number of free bike parking places	NMBS (2018)	0
N_AT_park_p	Number of paying bike parking places		0
<i>CAR</i>			
N_CA_park_f	Number of free car parking places	NMBS (2018)	4
N_CA_park_p	Number of paying car parking places		4
<i>BUS/TRAM/METRO</i>			
N_BTM_freq	Frequency of B/T/M departures on a day-basis	Based on GTFS data by De Lijn,	0
N_BTM_rout	Number of unique B/T/M routes to and from the station on a day-basis	TEC and MIVB/STIB (June 2018)	0

Both the active travel and the car dimensions focus on the parking capacity for these feeder modes, discerning between free and paying parking services. In order to calculate the accessibility to and from the railway station by public transport, the stops considered very close to the station (within a 300 m walkable network distance from all possible station exits) were filtered from publicly available GTFS (Gen-

eral Transit Feed Specification) data using R statistical programming and R studio. For these selected bus, tram and metro stops, two indicators were measured and afterwards grouped (summarized) per railway station, resulting in the indicators listed in Table 1. By drawing on frequently updated GTFS data, this method allows to easily update and calculate these indicators for different time windows. In this research, a typical Tuesday is selected for the calculation of the B/T/M (Bus/Tram/Metro) indicators.

Table 2. Indicators of the train dimension

Code	Indicator description	Source (year)	MV (%)
<i>TRAIN</i>			
N_TR_freq_tue	Frequency of departures on a day-basis on Tuesdays	Based on	0
N_TR_freq_sat	Frequency of departures on a day-basis on Saturdays	GTFS data by	0
N_TR_freq_off	Frequency of departures off-peak (between 10 and 11 AM, Tuesdays)	NMBS (June	0
N_TR_amp	Amplitude between the earliest departure/arrival and latest departure/arrival (Tuesday)	2018)	0
N_TR_tcentr	Travel time centrality		0
N_TR_trcentr	Transfer centrality		0

2.2.2 Train dimension

The first four indicators (Table 2) analyze characteristics of the railway service at the station (as is usually done in node-place analyses). However, over the last decade, more advanced accessibility indicators for public transport networks have been developed which assess the position (or “centrality”) of a station from a network perspective.

Different methods exist to calculate these types of “network measures” (Curtis & Scheurer, 2010). Some scholars (e.g., Papa, Moccia, Angiello, & Inglese, 2013; Caset et al., 2018) employed the open source Urban Network Analysis toolbox for ArcGIS (see Sevtsuk & Mekonnen, 2012) to calculate these measures, whereas Curtis and Scheurer (2010, 2016) developed a series of multimodal public transport centrality measures as part of the SNAMUTS decision tool. As some of these latter measures were previously operationalized for the Flemish and Brussels railway network (see Verachtert et al., 2016), we opted to take the same approach. We operationalized the SNAMUTS “closeness centrality” (equation 1) and “degree centrality” (equation 2) measures (see Curtis & Scheurer 2016, p. 304). Centrality was measured relative to all Belgian stations and the foreign stations included in the GTFS dataset (those on the railway lines towards Lille, Amsterdam, Maastricht, Aachen, and Luxembourg). In order to enhance the interpretation of the indicators for the users of the tool (and to avoid possible confusion with the standard language used in classic network analysis (see Newman, 2010), we renamed the indicators into “travel time centrality” and “transfer centrality,” respectively.

The travel time centrality indicator calculates the minimum cumulative impediment ($L_{min, ij}$) between station i and all other stations j in the network (with N = all railway stations), in terms of travel time (t) and service frequency (f):

$$travel\ time\ centrality_i = \sum_{j=1, j \neq i}^N \frac{L_{min, ij}}{N - 1} \quad (1)$$

with: impediment value of route segment between stations i and $j = 4 \times \sqrt{\frac{t_{ij}}{f_{ij}}}$

The transfer centrality indicator calculates centrality for a station i in terms of the average minimum number of transfers (p) required to reach all other stations j in the network:

$$transfer\ centrality_i = \sum_{j=1, j \neq i}^N \frac{p_{min,ij}}{N-1} \tag{2}$$

Similar to the B/T/M indicators, drawing on GTFS data allows to easily update and calculate these indicators for different time windows. Again, Tuesday is selected for the calculation of these indicators.

In order to calculate the place dimensions, the extent of “the place” needs to be defined. As pointed out by Bertolini and Spit (1998, p. 12), “any delimitation of the station as place is destined to be somewhat arbitrary,” as the influence of a station may go far beyond its immediate surroundings. Conversely, entities located nearby may have no apparent relationship with the station. As this paper draws on empirical assessment models which focus on the walkable area of stations, we delineate “place” as the accessible area covered by a walkable street network distance of 1200 meter (roughly 15 minutes walking). Table 3 illustrates the indicators that are part of the place dimensions.

First, the “Density” dimension refers to the concentration of jobs, residents and amenities. Its contribution to the walkability of a transit-oriented neighborhood is detailed in Cervero and Kockelman (1997). The rationale is that density provides the potential to reduce distances between people and the places they need to access. There are a variety of analytical approaches to measure density (for a recent overview see Dovey & Pafka, 2018a). Here we adopt the most commonly used measures of population density (residents/hectare and jobs/hectare) along with three measures reflecting the density of amenities (basic amenities/hectare, regional amenities/hectare and metropolitan amenities/hectare). The difference between the three types of amenities is specified in Verachtert et al. (2016). In general terms, basic amenities are those considered necessary to organize daily life (e.g., a kindergarten, a pharmacy, a general practitioner). Regional amenities are assumed to have a larger catchment, serving different urban areas in the region (e.g., a shopping mall, a cultural center, offices), while metropolitan amenities have the largest catchment (e.g., touristic attractions, a university). Based on the coordinates of the individual amenity locations, distance decay functions were applied (depending on the assumed amenity catchment size) and rescaled to a raster with 100 m x 100 m cells.

Second, “Diversity” or land-use mix is a key ingredient of walkability (Dovey & Pafka, 2018b). In the node-place modeling literature, this dimension is often operationalized by employing the entropy measure used by Zweedijk and Serlie (1998), the dissimilarity index introduced by Cervero and Kockelman (1997), the MXI (Mixed-use Index) proposed by van den Hoek (2008), or other types of functional mix measures (such as the “Mixed-ness Index” applied by Singh, Luckman, Flacke, Zuidgeest, & van Maarseveen, 2017). While these measures capture “functional” land-use mix, they do not capture the spatial configuration of the land-use types (a critique that was already raised by Hess, Moudon, & Logsdon, 2001). Given this, we draw on the work of Hess et al. (2001) in which a landscape ecology approach to measuring “patch” diversity is applied within the context of transport and land use interaction studies. Using the Fragstats software (see McGarigal & Marks, 1995), two indicators are measured, which reflect the functional and the spatial diversity of land-use types within each station’s precinct: the Shannon’s Diversity Index (SHDI, equation 3) and the Contagion Index (CI, equation 4).

$$SHDI = - \sum_{i=1}^m (P_i \cdot \ln P_i) \tag{3}$$

with: i the type of land use and Pi the proportion of land-use type i present within the station’s precinct. SHDI increases as the number of different land-use types increases and/or the proportional distribution of area among types becomes more equitable.

CI measures both the land-use “interspersed” (the spatial intermixing of raster cells with different types of land use) and its “dispersion” (the spatial distribution of a land-use type with respect to the station precinct), at the level of individual raster cells. Higher CI values may point to precincts with a few large and contiguous land-use patches, whereas lower values generally characterize precincts with many small and dispersed land-use patches. CI, detailed mathematically below, represents the observed level of contagion as a percentage of the maximum, given the total number of land-use types:

$$CI = 1 + \frac{\sum_{i=1}^m \sum_{k=1}^m \left[P_i \frac{g_{ik}}{\sum_{k=1}^m g_{ik}} \cdot \ln(P_i) \frac{g_{ik}}{\sum_{k=1}^m g_{ik}} \right]}{2 \ln(m)} \cdot 100 \quad (4)$$

More specifically, CI consists of the sum, over land-use types, of the product of two probabilities: the probability that a random chosen raster cell belongs to type i (estimated by the proportional abundance of type i), and the conditional probability that given a cell is of type i , one of its neighboring cells belongs to type j (estimated by the proportional abundance of type i adjacencies involving type j) (see McGarigal & Marks, 1995). Equally crucial here is the way in which the land-use types, functioning as proxies for walking trip origins and destinations, are defined. The approach to land-use categorization adopted here draws on the triangular model of functional mix in which urban functions are divided into just three primary categories of housing, work and amenities (see van den Hoek, 2008; Nes, Berghauer, & Mashhoodi, 2012; Dovey & Pafka, 2018b). We draw on land-use data on a 10-meter raster scale containing 39 land-use types, which we assigned to the three categories.

The third “D,” design, is added to the station assessment model. This design dimension aims to measure the ways in which walkable and bikeable access (or “active accessibility,” see Vale, Saraiva, & Pereira, 2016) is mediated by the urban morphology of public space and by the built environment. As regards walkable access, in line with Pafka and Dovey (2018), two key approaches are applied, i.e. walkable catchments and permeability. The first indicator measures the “catchment” of a station, not in terms of its walkable surface (or “Pedshed,” which is frequently used in node-place modelling studies), but in terms of “how much” it gives access to. In line with Pafka and Dovey (2018), we focus on the extent of public/private interface within the station’s walkable precinct as a proxy measure for how much is actually “caught.” More specifically, this “interface catchment” (IC) is calculated by summarizing the length of all walkable street segments (the public realm) that are also flanked by buildings (the private realm). The second indicator, permeability, measures the extent to which the urban morphology is permeated by publicly accessible space (see also Marshall, 2005) by means of mapping the total number of street crossings per station area (see also Ryan & Frank, 2009). This measure relates to the ease of movement through an urban area as well as the multiplicity of route choices between any pair of points. Both indicators are complemented by a third one mapping the walkable and bikeable street networks within a station’s precinct.

Table 3. Indicators of the place dimension

Code	Indicator description	Source (year)	MV (%)
<i>DENSITY</i>			
P_DE_res	Summarized residential densities for all raster cells in the station's precinct	VITO (see Verachtert et al. 2016)	0
P_DE_job	Summarized job densities for all raster cells in the station's precinct		0
P_DE_bas	Summarized basic amenity densities for all raster cells in the station's precinct		0
P_DE_reg	Summarized regional amenity densities for all raster cells in the station's precinct		0
P_DE_met	Summarized metropolitan amenity densities for all raster cells in the station's precinct		0
<i>DIVERSITY</i>			
P_DI_shan	Shannon Diversity Index for all raster cells in the station's precinct	Based on VITO (2013)	0
P_DI_CI	Contagion Index for all raster cells in the station's precinct		0
<i>DESIGN</i>			
P_DG_IC	Interface catchment: total length of street segments flanked by buildings	Based on OSM (2018)	0
P_DG_perm	Permeability defined as the ratio of the number of street segments and street intersections within the CA		0
P_DG_netw	Network length of walkable and bikeable street infrastructure in station precinct	Based on OSM (2018) and Verachtert et al. (2016)	0

2.2.4 Motivation, ridership and effort dimensions

The data for these three dimensions was provided by the Belgian national railway company NMBS and is based on a comprehensive analysis of origin-destination season ticket pairs. As for the effort field, NMBS calculated the percentage of people living within a certain Euclidean distance from their origin station: walking distance (closer than 900m), biking distance (between 900 and 3000m) and a farther distance (> 3000m). The second field, ridership, is structured around three dimensions: origin (the estimated percentage of people using the station as their origin station), ridership (the frequency of passengers boarding trains on a regular working day), and destination (the estimated percentage of people using the station as destination station). We emphasize that both origin and destination do not reflect quantities of travelers (ridership) but specify the proportion of tickets for which the station functions as an origin or a destination station. Thirdly, the motivation field informs us about the (assumed) motivation of people traveling to a particular station. Four categories (dimensions) were discerned: secondary education, tertiary education, work and other. The first three dimensions draw on season ticket data and were categorized by NMBS based on age groups (respectively <19 years old, [19 – 25] and > 25). The latter dimension “other” reflects individual ticket sale figures.

Table 4. Indicators of the motivation, ridership and effort fields

Code	Indicator description	Source (year)	MV (%)
<i>MOTIVATION</i>			
PP_MOT_sec	% of people using the station as destination station for secondary education purposes	NMBS (2018)	20
PP_MOT_tert	% of people using the station as destination station for tertiary education purposes		20
PP_MOT_work	% of people using the station as destination station for work purposes		20
PP_MOT_other	% of people using the station as destination station for other purposes		20
<i>RIDERHIP</i>			
PP_RID_orig	% of people using the station as an origin station	NMBS (2017 and 2018)	0
PP_RID_rid	The frequency of passengers boarding trains on a working day		0
PP_RID_dest	% of people using the station as a destination station		0
<i>EFFORT</i>			
PP_EFF_walk	% of people using the station as origin station who live within a Euclidean walking distance (< 900 m)	NMBS (2018)	21
PP_EFF_bike	% of people using the station as origin station who live within a Euclidean biking distance (900 – 3000 m)		21
PP_EFF_far	% of people using the station as origin station who live farther (> 3000 m)		21

2.3 Analyzing patterns and developing a typology of stations

Taken together, our model consists of 32 indicators. Most indicators have no missing values, except for those of the motivation (20%) and effort (21%) fields. This is due to the existence of tariff zones,² which leaves us with some uncertainty regarding the specific station travelled to and/or from when analyzing sold tickets and passes.

A first step in a series of descriptive analyses consists of a two-sided Spearman correlation analysis. This will shed light on the direction and strength of the relations between the indicators belonging to different dimensions and fields, allowing for a validation of the conceptual model from a statistical point of view. As the proportion of missing values for the motivation and effort fields is fairly high, this correlation analysis is based on a list-wise deletion of missing values, resulting in a subset of 221 stations. Drawing on these descriptive statistics, a procedure to estimate the missing values in the remainder of the dataset can be set up, in order to develop a typology of stations inclusive of the effort and motivation fields. The procedure used is a multiple imputation (MI) algorithm and is conducted in SPSS. Since the values are missing in a non-random way, the monotone MI procedure is used. The “pooled” result (the average of the imputed values over 5 runs) is retained as a basis for the following analyses. The findings are detailed in section 3.1.

Second, in order to verify to what extent the demand-side indicators add meaning to the findings of a conventional node-place type of classification, we conduct two cluster analyses. The first one only includes the supply-side accessibility fields (node, train and place), while the second one focuses on the demand-side fields. Both cluster analyses draw on an a priori exploratory factor analysis in order to generate a classification based on uncorrelated variables. The findings are detailed in section 3.2.

² There are 12 tariff zones in Flanders and Brussels: Aalst, Antwerp, Bruges, Brussels, Denderleeuw, Dendermonde, Ghent, Halle, Hasselt, Knokke, Leuven, Mechelen.

3 Findings

3.1 Correlation analysis

When analyzing the direction and strength of correlations between the indicators (see Appendix 1), the following insights emerge.

First, the overall logic of the model seems justified: indicators belonging to the same field generally correlate strongly with each other and exhibit similar patterns with respect to the other fields. The decision to separate the train accessibility indicators from those of the feeder modes (bus, tram and metro) also seems justified, as correlations are clearly contained within both dimensions (e.g., the bus indicators have strong mutual correlations but exhibit weak correlations with the train indicators and vice versa). Furthermore, the assumed associations between the node and effort fields, the train and ridership fields, and the place and motivation fields seem supported by these findings. First, there is a clear positive relation between the distance people live from their station and the supply of car parking (and to a lesser extent bike parking) facilities and feeder public transport mode accessibility. Second, the ridership indicator is strongly positively correlated with most of the train indicators, especially with the frequency indicators. Third, the motivational factors, and in particular the secondary education profile, are generally significantly and strongly positively correlated with the place indicators. The ridership indicator “destination” is furthermore strongly positively correlated with all motivation indicators, justifying its location on the right-hand side of the ridership dimension in Figure 3.

On the indicator-level the following findings are noteworthy. The three design indicators are very strongly correlated ($>.936$). It thus seems that the extent of the public-private interface, the permeability of the street network and the total length of walking and cycling infrastructure to a large extent contain similar information about station area walkability. Further research could therefore look into alternative ways to assess walkable and bikeable access to stations mediated specifically by the built environment. The inclusion of measures of street network integration or connectivity as proposed by Pafka and Dovey (2018) may be a sensible addition here. The work by Nes and Stolk (2012) and Liu, Wu, Hidetosi, and Gao (2015) in which the spatial configuration of the local street network in railway precincts is assessed using Space Syntax analysis may prove instrumental. The open source “Urban Network Analysis” toolbox for ArcGIS (Sevtstuk & Mekonnen, 2012) arguably also provides an interesting basis to refine centrality analyses of street networks with respect to railway stations (see, e.g., Sun, Zacharias, Ma, & Oreskovic, 2016).

In a similar vein, the two diversity indicators are very strongly correlated, indicating that the intended differences in functional and spatial land-use diversity are not yet sufficiently captured. More experimentation with alternative Fragstats measures (and parameters) or alternative software packages is a sensible next step.

Furthermore, the two centrality indicators do not correlate strongly with other indicators, which suggests that both contain specific information about the network structure that is not covered by the other train indicators. Travel time centrality in particular does not exhibit any strong correlations, whilst transfer centrality is quite strongly correlated with the other train indicators.

A final reflection concerns the ridership indicator. Although it is beyond the scope of this paper to investigate the determinants of transit patronage in Flanders and Brussels (see, for example, Cardozo, García-Palomares, & Guitiérrez, 2012; Chow, Zhao, Liu, & Li, 2006), the correlation patterns do allow to explore some preliminary relations. Ridership is strongly correlated with (most indicators of) all other fields. Especially (free) car and bike parking supply, feeder public transport services, the design dimension, both educational motivations and the size of the catchment area are strongly related. Also, transfer centrality seems more important in explaining ridership than the travel time and frequency-based centrality measure.

3.2 Factor and cluster analysis

In this section the results of two cluster analyses are discussed: a first one based on the supply-side accessibility characteristics (node, train, place) and a second one based on the demand-side accessibility characteristics (people). Clustering based on the full set of indicators did not result in an intelligible typology, hence the separate analyses discussed in two subsections.

3.2.1 A typology for nodes and places

The factor analysis (orthogonal, varimax rotation) results in 4 interpretable factors (see Appendix 2a) with an eigenvalue larger than 1 and explaining 81% of total variance. Factor 1 has strong loadings for nearly all place indicators, factor 2 for the train frequency indicators, factor 3 for most of the node indicators and the train amplitude and factor 4 for both centrality measures. Based on these factors, a Two Step cluster analysis (log-likelihood distance criterion and BIC clustering criterion) is conducted resulting in 5 interpretable groups of stations (see Figure 4). Brief descriptions for all station types are also provided in Figure 4.

The majority of stations in Flanders and Brussels classify as rural stations (very low scores on all place indicators) with weak bus and tram accessibility levels and weak railway accessibility in terms of frequency (on average 64 departing trains per Tuesday), but favorable in terms of network centrality. In other words, based on the actual NMBS timetables, these smaller stations with low public transport service frequencies are nonetheless located on the most important railway lines connecting the largest cities in Flanders, giving them robust and strategic potential in terms of future urban developments. Although most of these stations are located along the MTS spatial backbone sketched in Figure 1, the stations located both in the Western part and the Eastern part of Flanders exhibit a distinct typology. In these areas, types 2 and 4 are more abundant. Type 4 is characterized by very low bus and tram accessibility levels, and the lowest railway accessibility in terms of frequencies, amplitude and centrality; these stations are not at all or only weakly connected to the Walloon and foreign stations that are part of the network analysis. The level of “urbanity” of type 4 is nonetheless stronger than the type 1 stations which, from the perspective of a node-place equilibrium, arguably leads to a development scenario prioritizing higher railway accessibility in the network. The type 2 stations, in turn, are all located in urbanized areas and—in contrast with the previous two types—exhibit sizable levels of bus and tram accessibility and have by far the highest car parking supply. Their railway-based accessibility is moderate, except for the service amplitude which is sizable in most cases. Geographically speaking this type is scattered, but there are nonetheless some clusters in and around the large cities of Kortrijk and Antwerp. The Brussels Capital Region is in turn predominantly characterized by stations of type 3. The main features of this type, differentiating it from the Flemish stations, are its very high scores on all place indicators and (mainly travel time) centrality measures. However, given the metropolitan context, the very low scores for the (mainly bike) parking facilities are striking. And finally, two of the three most important railway stations in the Belgian railway network (Brussels South and Brussels North) along with the main stations of the largest cities in Flanders, classify as type 5. The main distinctive factors are the very high frequencies and (mainly transfer) centrality scores and the very high bus, tram and metro accessibility levels.

3.2.2 A user-based typology

The factor analysis (orthogonal, varimax rotation) results in 4 factors (see Appendix 2b) with an eigenvalue of 1 and explaining 78% of total variance. Factor 1 has strong positive loadings for destination, and the work and leisure motivations, while factor 2 strongly loads on the tertiary (and to a lesser extent

secondary) education motivation and moderately loads on ridership. The third and fourth factor both cover the effort dimension, therefore only the third factor is retained. It loads strongly on large catchment areas where most people live further than a 3000 m radial buffer. The first three factors together explain 70% of total variance.

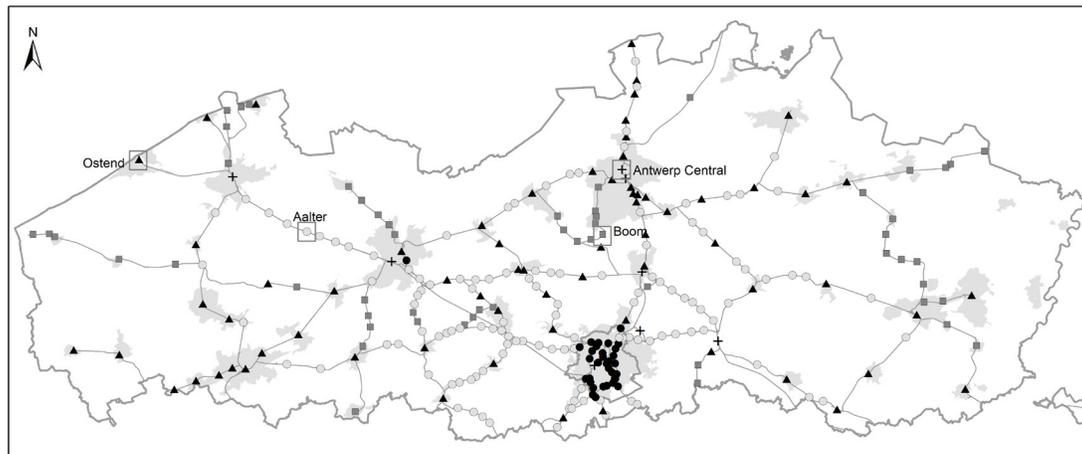
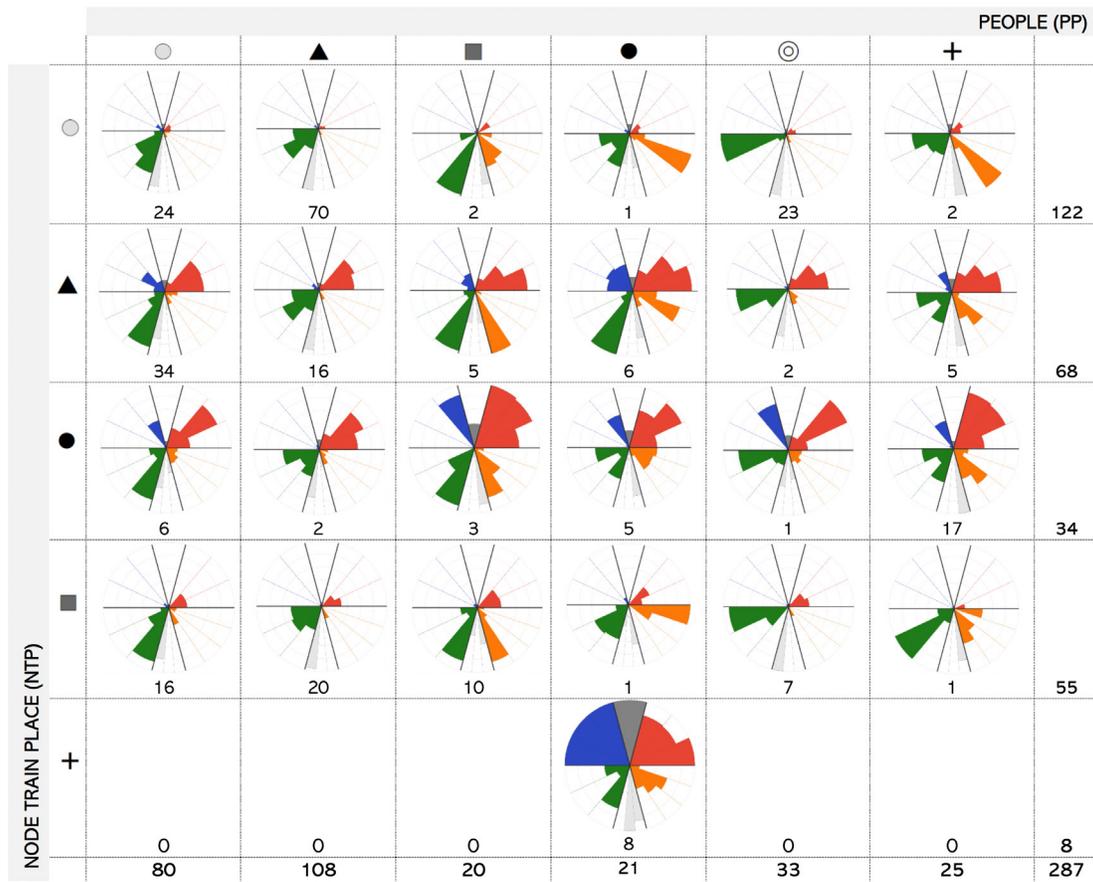
Based on these factors, a Two Step cluster analysis (log-likelihood distance criterion and BIC clustering criterion) is conducted resulting in 6 interpretable groups of stations (see Figure 5). The majority of stations in Flanders and Brussels classify as strong origin stations with semi large catchment areas (type 2). These stations are abundant across the region and are often located one after another along certain corridors in between the larger stations. The western and eastern parts of Flanders nonetheless mainly exhibit station types 1 and 3. The first type is quite similar to type 2 but is characterized by large catchment areas. Unsurprisingly, these stations are often located quite far from their neighboring stations, arguably explaining their larger catchments. Type 3 stations in turn have the largest catchment areas and are also characterized by strong leisure motivations (high destination motivations in terms of individual ticket sales). These are mainly destination stations and are located nearly exclusively in the periphery of the network, i.e. along and near the coastline in the West and at the end of some rail corridors in the East. The stations with the smallest catchment areas (type 5, strong origin stations) are in turn mainly located along some corridors in the center of the network. Finally, the station types with the least number of stations are types 4 and 6. The type 4 stations are mixed origin/destination stations with a strong educational motivation and high to very high ridership. They are located at the cornerstones of the MTS sketched in Figure 1 and across the Brussels Capital Region. The strongest destination stations (type 6) are also mainly found in the urban regions of Brussels and Antwerp. The classifications of those stations located in the tariff zones should nonetheless be interpreted with prudence, as most of the people-based indicator scores (except for those part of the “ridership” field) are based on the imputation estimates described above.

3.2.3 Blending typologies

Table 5 illustrates by means of a cross-table how the stations are sorted across both typologies (NTP = Node, Train and Place, PP = People) and visualizes a rose diagram for each of the cluster intersections. The numbers indicate how much stations are included in the different intersections and the symbols across both sides pertain to those used in Figures 4 and 5. The dimensions in the diagrams are calculated as averages of the underlying indicators, and afterwards the dimension scores were rescaled to vary between 0 and 10 for each of the intersections.

Table 5 illustrates how the majority of railway stations in Flanders and Brussels classify within a limited number of cluster intersections. A clear example is the general overlap between the rural stations (type 1 in NTP) and the moderate to strong origin stations (1, 2 and 5 in PP). Likewise, the urban stations of type 2 in NTP are most often also moderate to strong origin stations with semi large to large catchments (types 1 and 2 in PP). A similar observation holds for the type 4 stations (NTP), while a distinct pattern is true for the metropolitan stations of type 3 in NTP, which classify predominantly as strong destination stations in the PP typology. The metropolitan stations of type 5 in NTP exclusively classify as the mixed origin/destination stations of type 4. These strong cluster intersections may not surprise given the earlier demonstrated correlations between the supply- and demand-side fields in the model. The confrontation of both typologies nonetheless allows for a more nuanced and differentiated account of a station's functioning in the Flemish and Brussels railway network when compared to a standard node-place typology, given that each NTP category diversifies into a range of distinct PP categories and vice versa. This comparative framework furthermore allows to identify stations that fit into unexpected or underrepresented “boxes” within Table 5, which can be submitted to further scrutinization.

Table 5. Cross-table of cluster intersections with averaged rose diagrams



Groups

- 1 - rural with low frequency, average centrality, low parking supply and very low BTM accessibility
- ▲ 2 - urban with low frequency, average centrality, high parking supply and high BTM accessibility
- 3 - metropolitan with high frequency and centrality, very low parking supply and high BTM accessibility
- 4 - mix rural/urban with very low frequency and centrality, low parking supply and low BTM accessibility
- + 5 - metropolitan with very high frequency and centrality, very high parking supply and very high BTM accessibility

- railways
- urban areas *
- regional border

* Flemish Government (2016) and European Environment Agency (2006)

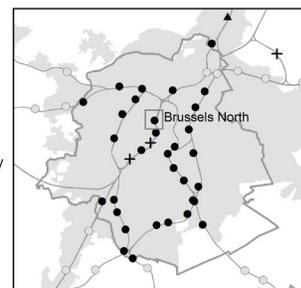


Figure 4. Typology of stations based on node, train and place dimensions

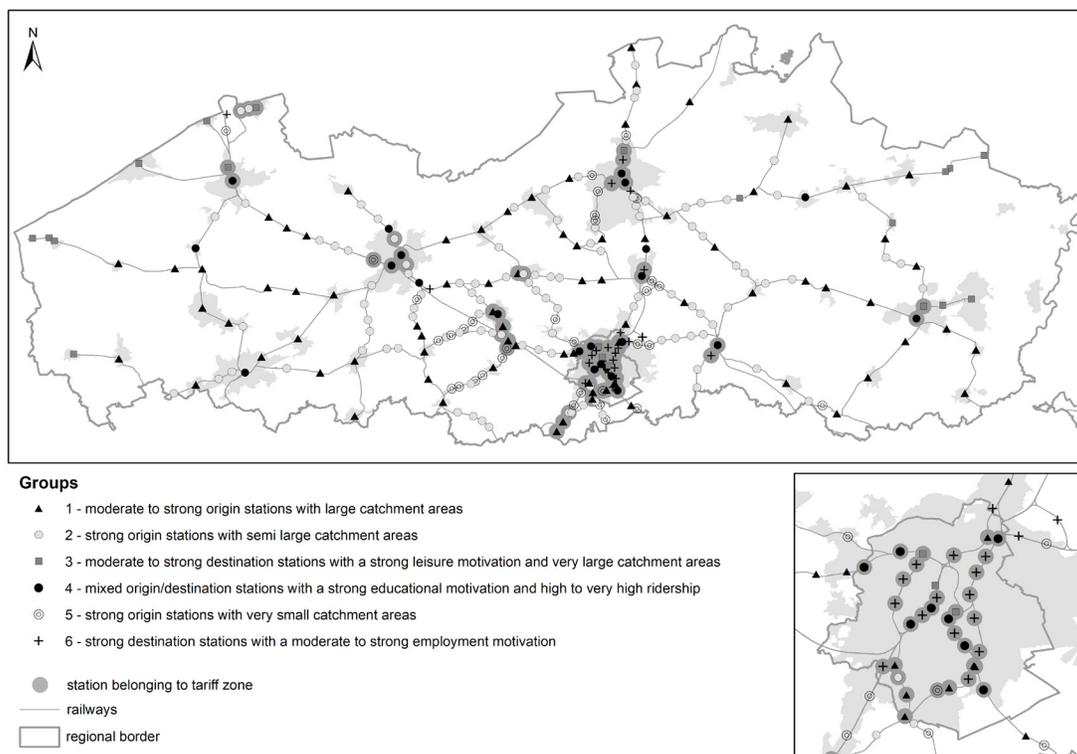


Figure 5. User-based typology of stations

4 Practical application to five cases

In order to clarify what the station-specific results of these analyses may mean for planning practice, this section discusses the characteristics of five railway stations. Each of these stations belongs to a cluster derived from the NTP typology (see Figure 4 where the five stations have been designated and Table 5 indicating their cluster membership for both typologies). We will also explain the position of the station at hand within the PP typology and relate our findings to the objectives of urban planning policies in Flanders and Brussels.

4.1 Aalter station

As indicated in Figure 4, Aalter station is located halfway between the larger cities of Bruges and Ghent. The strong growth of the municipality of Aalter since the 1970s, is a typical consequence of the suburbanization trend of that era, which was initially fueled by the smooth access of Aalter to the nearby motorway E40, but which is currently also complemented with reasonable access to the railway network. Judging from Figure 6, Aalter is a predominant origin station with a large catchment area where most people live farther than 3 km away from the station. The motivations to reach Aalter station as a destination station are very limited, with the exception of a small proportion of season tickets used for secondary education purposes. Although the number of daily passengers boarding (over 2,000) is quite sizable compared to the size of the municipality (more than 20,000 inhabitants), the ridership indicator scores very low when compared to all other stations in the network. Combined, these characteristics fit the station's classification as a type 1 station in the PP typology.

timetable of trains departing from this terminus is adapted to domestic traffic, which includes an important flow of long-distance commuters. According to the NTP typology, this station can be described as “urban with low frequency, average centrality, high parking supply and high BTM accessibility.” This can be explained by, among other things, its location in a regional urban area, its status as a terminus, and its importance as a hub for regional public transport, the backbone of which is the coastal light railway. Although most train indicators score weak or moderate, the amplitude is very high.

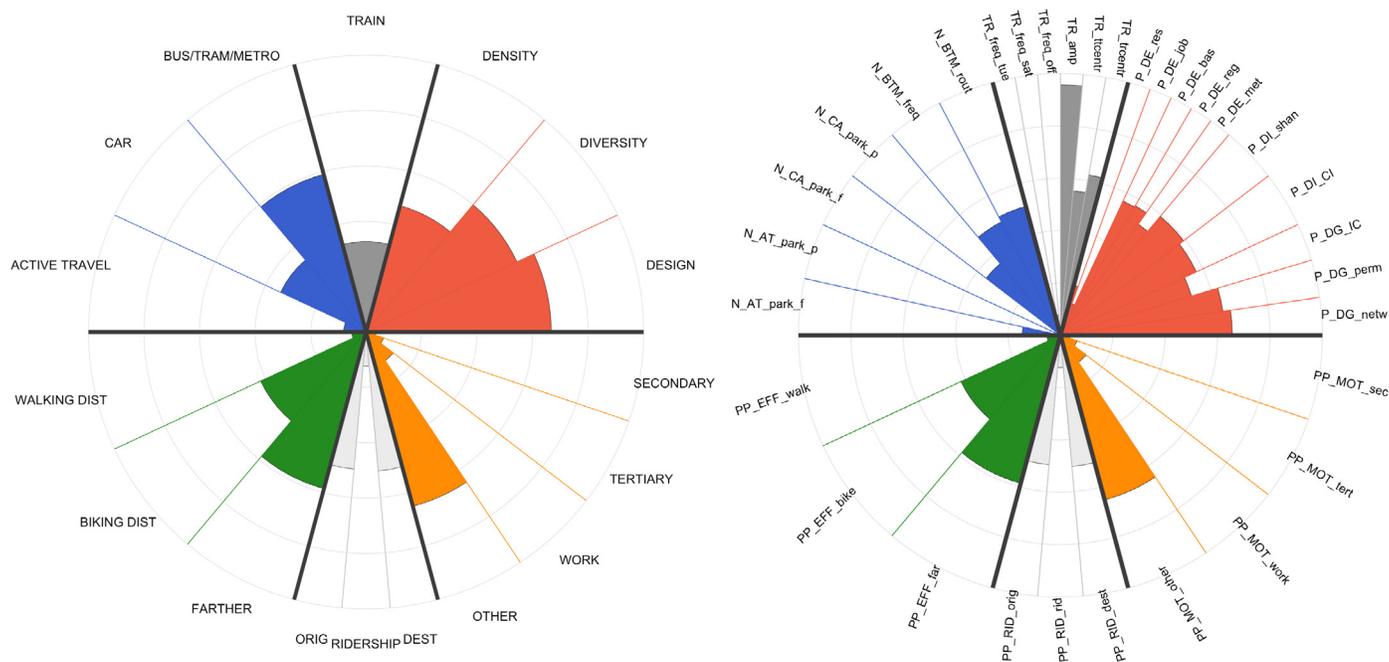


Figure 7. Ostend station scores (left: dimensions, right: indicators)

According to the PP typology, Ostend station is a “moderate to strong destination station with a strong leisure motivation and very large catchment area.” This description seems to align with the accessibility profile provided in Figure 7, although the station in fact exhibits a “perfect” balance between an origin (long-distance commuters) and a destination (presence of the beach as a tourist attraction and sizable presence of regional and metropolitan amenities) character. The remote location of Ostend station in the national rail network may furthermore explain its large catchment area. The station’s ridership on a Tuesday is about 7,600, which is significant but pales in comparison to the Brussels main stations (with over 61,000 boarding passengers).

Given the status of Ostend as a regional urban area, strengthening the current concentration of a mixed residential, professional, educational and recreational environment here would be an obvious policy option. The immediate vicinity of the station, which is located next to the currently underused seaport, is hardly developed and may well be suitable for compaction.

4.3 Brussels North station

Brussels North station is one of the three major stations of the Brussels north-south corridor, the busiest railway line in Belgium receiving around 1,200 trains per working day. The station is located centrally in the Brussels metropolitan area, which includes the Brussels Capital Region comprising 1.2 million inhabitants, and is situated between a modernist office district and a densely populated residential area. Judging from Figure 8, the train frequencies are maximal, both off-peak and in the weekend. Besides the

discussed how both typologies intersect geographically, and how they may each provide unique information about the functioning of stations within the Flemish and Brussels railway network. By doing so, our research serves as a refinement of the conceptual typology of strategic railway stations put forward in the BRV strategic vision and the “spatial backbone” operational framework. Additionally, our analysis is comprehensive in that it includes all railway stations in both regions, allowing to deduce meaningful observations for those stations that fall outside the scope of the BRV typology (which strictly focuses on the stations situated along the MTS).

We nonetheless point out that the rose diagrams developed in this paper may lose their informative capacity when used in isolation from cartographic material and/or when absolute figures underpin the relative scores in the diagrams. As this station assessment model is intended to help developing a “useful” and “usable” (see Pelzer, 2017) planning support system (PSS), we presume that additional maps that serve to clarify and increase the interpretation of specific indicator scores, as well as additional tables detailing the absolute figures behind the relative scores will be crucial. After all, several studies dealing with interdisciplinary communication processes facilitated by PSS (see Geertman & Stillwell, 2009; Pelzer & Geertman, 2014; Papa, Coppola, Angiello, & Carpentieri, 2017) stress the importance of spatial visualizations and of transparency in data and methods to render results more easily understandable and relevant for the end users of the tool. In these earlier studies, the added value of node-place analyses as perceived by their intended end users is nonetheless rarely evaluated; to the best of our knowledge, there are no such studies with the exception of Gilliard et al. (2018), that critically validated a node-place model application in a design studio setting with urban design students. This observation has a broader significance, as few planning support instruments commonly discussed in the literature are explicitly validated by their intended users (te Brömmelstroet, 2010; Straatemeier, Bertolini, te Brömmelstroet, & Hoetjes, 2010; Pelzer, Geertman, vander Heijden, & Rouwette, 2014; see also Bertolini, 2017). This lack of cross-fertilization between the output of applied academic research and actual planning instruments hampers the integration of scientific and practical knowledge (Balducci & Bertolini, 2007).

In order to help bridge the gap between planning research and planning practice, a next step in this research will therefore consist of a qualitative validation of the usefulness of this model in the Flemish context. Given the growing importance of the urban-regional governance level in integrating transport and land use in Flanders, a sensible strategy would be to focus on the recently (2018) established “transport regions.” The objective of these new regional partnerships (15 in total) is to stimulate cooperation between municipalities, public transport operators, the Flemish Government and other stakeholders around the organization and coordination of public transport in the region, and this in close cooperation with spatial interventions. A validation of the model in this context should inform us about the extent to which particular building blocks of the model require modification to better fit the needs of its end users. It might turn out that our concern with the pursuit of rigor (or “soundness,” see Bertolini, Clercq, & Kapoen, 2005) in the operationalization of certain indicators, proves less directional for the future development of our planning support tool. In this way, the frequently raised contention that PSS developers should seek for an effective balance between scientific rigor and practical relevance (see among others, Papa, Silva, te Brömmelstroet, & Hull, 2016; Silva & Larsson 2018) may be put to the test.

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Appendices

Appendices are available as supplemental files at <https://www.jtlu.org/index.php/jtlu/article/view/1483>