Labor Mobility and Commuting

The timeliness of commuting research is closely related to the necessity of monitoring and controlling of commuter routes. Qualitative management and control of labor mobility is a key factor in efficient economic development at any level (a company, a region, and a state) since implementation of any project is highly dependent on the presence and quality of the available human capital. It is uncontroversial that range and intensity of commuting in large modern agglomerations demonstrates a significant macro influence on the socioeconomic situation in the region, as had been demonstrated in a set of our earlier research [12-15]. But on the other side there is a huge lack of experimental data related to developing countries in this field. The main problem is prohibitively high costs of any studies aimed to collect commuting data in conditions of limited funding. As a result, contrary to intensive research on commuting in the cities worldwide, there are no empirical studies of commuting in Russian cities.

The present study is aimed to fill this gap analyzing working behavior of almost 60% of working age population of Moscow region using unique self-produced micro data set, which has been reconstructed from a set of Russian database data leaked to Internet. Living and working locations linked with personal data (age, gender, and income) are available for each person in the database. The present paper develops a new approach to the analysis of commuting that involves implementation of GIS (Geographic Information System) processing technologies on the geospatial data-living and working locations of commuters in MR. This elevates commuting studies to a new qualitative level in order to address a number of empirical and theoretical problems. Next section gives a short review of implementation of GIS-approach
to commuting studies, and the visualized patterns of commuting are presented in the third section. The fourth section is aimed to describe the GIS-method which was used, while statistical results obtained by it are discussed in fifth section. Final section concludes the results.

GIS Technologies in the Analysis of Commuting

GIS technologies are usually used in two following approaches of commuting studies.

Spatial Approach

Visualization of commuter home-workplace routes enables the creation of a spatial pattern of population labor mobility (direction, length and intensity), identification of districts of labor attraction and outflow, analysis of the disproportion between supply (place of residence) and demand (place of work), and estimation of the traffic intensity on transport routes.

Analytical Approach

An analytical approach is possible if GIS information is linked to the individual characteristics of commuters (gender, age, income, etc.) and/or their employers (industry, form of ownership, enterprise size, etc.) at the micro level. In this case, it is possible to analyze the social structure of the commuters and the impact of individual factors on labor mobility as well as characteristics of the businesses attracting commuters.

However, desperate shortage of data remains the main problem of application of GIS analysis techniques. Therefore, despite generally high interest in the analysis of commuting (about one hundred research articles per year) GIS technologies are only used in a tiny number of them. Furthermore, in most of these papers GIS is applied for aggregation of spatial variables, such as comparative weight of districts according to area, average time estimation, route lengths over the region, etc. The calculated variables were then used in multiple regression analyses. As examples one can mention the research of the spatial interaction of housing and labor markets in North West England [5] and intra-urban variation of commuting time and distance in Columbus Ohio, USA [18]. However, in these studies GIS was used as a subsidiary tool.

A direct spatial approach was applied to map commuting transport routes in New York and Amsterdam using the full power of GIS technologies [1]. A good example of an analytical approach is the research of the interregional migration of British students, where the geography of school-university-first job shifts was estimated using GIS platform [3]. One of the most extensive studies is an analysis of ecological influence (carbon dioxide emission) on the environment depending on commuter journeys and alternative transport movement models and scenarios based on GIS data (residential and working addresses of 1829 people surveyed) from a special transport survey made by the Vodafone company [20]. It is also worth mentioning the GIS analysis of commuter flows using intraday bicycle commuting data [16].

GIS-based studies related to the topic of the present research mostly focus on the following two directions.

Analysis of GIS Maps

Analysis of GIS maps via graphical identification of natural and artificial landscape objects on the maps with the purpose of reconstructing nature or human activity patterns and analyzing them further. Examples include comparative analysis of structure and economic force of the regions of China as a whole and the development of agricultural districts in the province of Guangxi in particular [18, 19]; research (simulations) of Madrid agglomeration growth scenarios [9]; comparative analysis of economic development of the administrative districts of Romania [7]; analysis of land use for ecotourism planning in Thailand [2].

GIS analysis of GPS Data

GIS analysis of GPS data, gathered from moving objects under study (cars, public transport, etc.), for example analysis of the interaction of participants with the transport network in Minnesota [6, 11, 21]. However, such surveys are more related to transport assessment and modeling than commuting and are aimed to study driver behavior.

The research method developed in this paper combines microanalysis with GIS technologies to provide an innovative and unique approach to fine details of the commuting process that still remains an open question. Combining analysis of commuting patterns in Moscow region with individual behavior of commuters is the key to the power of the method proposed and discussed further.

Method of GIS Processing

The basis of this research is a unique micro dataset for the year 2001 covering 60 per cent of working-age population of Moscow region. The original data were from databases (DB) of the Tax inspectorate of the Russian Federation (RF), the Pension Fund of the RF and the State Register of Russian Enterprises. These confidential data have leaked to internet due to illegal actions of unknown persons. With these data in the public domain, and our methods of statistical analysis preventing disclosure of personal information, it has been decided worth to use this unique information (unavailable by any other means) for scientific purposes. The primary data from different DBs have been merged using key field-employee’s TIN (Tax Identification Number) present in all DBs. It was used to link the places of residence (PR) and work (WP) with employee personal data (PD: annual income, age and gender) and employer characteristics (EC: activity status, form of ownership, capital, etc.). As a result, a secondary DB has been created, which contains almost one million records of employees in the following format:

Employee (WP, PR, PD, EC)-----(1)

At the next stage GIS technologies are used to reconstruct the house-workplace route of each commuter. A special web application for Yandex maps platform (http://api.yandex.ru/maps) had been created for this purpose. For each commuter this program reconstructs a path on the regional transport network between the commuter’s PR and WP determined by postal addresses in the
In fact, the program solves the navigation task in the same way as a usual (car or hand) navigator does. The principal difference is that developed application is working in semi-batch mode processing large scale datasets and returning quantitative data on reconstructed routes for thousands of records. In comparison, a typical navigator works in a single interactive mode with manual input and graphical output. Examples of the results obtained by our program are shown in Figure 1. After GIS-processing the data have been obtained in the following format:

\[
\text{ROUTE (X1=PR, X2, ..., X1, ..., XN=WP) = GIS(PR,WP)}\]---(2)

Where XI are parameters of the commuting path (time, distance, and geo-coordinates) reconstructed as a polyline consisting of N route segments between commuter’s PR and WP. The total commuting distance and time are the most obvious and easily calculated GIS parameters which have been used for a basic analysis of commuter behavior. Finally routes for ~0.8 million commuter have been reconstructed (~80% reconstruction efficiency of the GIS-program) in order to obtain the final DB of the following structure:

\[
\text{Employee (PD, EC, ROUTE(X1,X2,...XN))}-----(3)
\]

The DB (3) contains unique information on the majority of commuters in Moscow region. The results of analysis of DB (3) in both geospatial (commuter patterns) and analytical (analysis of commuter individual characteristics) approaches will be demonstrated further on.

**Visualized Patterns of Commuting**

**Home and Work Location Maps**

In order to visualize geospatial commuting patterns in the Moscow region contour maps of commuters’ places of residence and work have been plotted (see Figure 2) using corresponding first and last geo points of the reconstructed commuter routes. From a comparison of these maps it becomes clear that commuters of the whole region come to work mainly to the Moscow central area with weaker flux directed to the regional cities of the second order in close vicinity to the center.

**Commuters’ Transportation Map**

Using reconstructed home-work routes (2) one can count the amount of commuters who crossed each particular geo coordinate on the roads in Moscow region. This commuters’ loading map of highways is shown in Figure 3. Radial structure of commuter fluxes is clearly visible on this map.

Let us proceed, moving from qualitative visual geospatial GIS analysis to quantitative assessment of the individual characteristics of commuters.

**Individual Characteristics of Commuters**

Analysis of the individual characteristics gives an insight into factors and mechanisms underlying the phenomenon of commuting. The individual behavior of commuters is a subject of intensive study, a detailed review of which had been presented in our previous review [15]. The use of DB (3) data makes it possible to perform a detailed analysis of the individual characteristics of commuters.

**Gender**

Figure 4 demonstrates the distributions of distance and journey times for commuters of Moscow region according to gender.

It should be pointed out that the average time spent by commuters for a single trip is one hour, while the average distance travelled is 50 km. One can see though that the majority of commuters spend around 35 minutes for a distance of 30 km (the main
peaks of top plots on Figure 4). At first sight the distributions of commuting distance and duration for women and men are almost identical. The differences in average values (Mean) are essentially less than their variance (RMS). However, a more detailed analysis shows that the discrepancy between men and women distributions is essentially asymmetric (see men-women difference curves under main distributions shown on top plots on Figure 4). They show that women dominate on short trips less than 40 minutes or 30 km (the points lie below zero), while men prevail on the rest longer trips. Thus, it can be emphasized that the data of this study is in favor of a lower commuting mobility of women compared with men. This hypothesis is supported by most of commuting theories. This is a perfect example, when the aggregate evaluation of the effect is too rough and does not provide an opportunity to

Figure 2. Commuter's transportation map of the whole Moscow region (top) and its central part (bottom). Gradient scale shows daily number of commuters crossing the color-coded points on highways.

Figure 3. Commuter's transportation map of the whole Moscow region (top) and its central part (bottom). Gradient scale shows daily number of commuters crossing the color-coded points on highways.
see the fine effect, the detection of which was only able thanks to unique data used for analysis.

The effect of Public transport

In this approach, all commuters are considered motorists. In fact, about 30% of commuters use public transport and this can introduce a systematic error in the calculation of the transport map, as well as commuting time distributions (Figures 2-4). We cannot estimate this effect precisely and this is a lack of methodology. However, we believe that this effect is small. First, the self-driving commuters dominate and contribute main part in the transport patterns. Secondly, travel times between public and personal transport do not differ essentially due to serious traffic jams on the roads of the region during rush hours. As a result commuting time distributions should not be violated dramatically due to this effect.

Age

The age distribution of commuters as well as average time of home-workplace routes in relation to age are shown in Figure 5. It should be noted that the age distributions of commuting men and women are almost the same and this once again confirms the gender equality in relation to participation in commuting. At the same time the age distribution of all citizens differs from that of the commuters. Comparing the two, it is evident that individuals of the “golden working age”, 20-50 years, dominate the group of commuters with a corresponding lack of older workers (>55 years). It is also clear that the youngest commuters (20-30 years old) are making the longest commuting journeys (see Figure 5, right). Hereafter the commuting time steadily decreases with age.

Income

The income of commuters declared to the Pension Fund of the Russian Federation (PFRF) is a decreasing exponential curve, as is shown in Figure 6, left. Whereas it should be noted that there is an anomalously large number of low declared incomes below the living wage of 1581 RUB (average value in the Moscow region in 2001). This indicates an enormous amount of cash-in-hand wages. It is most probable that this group includes workers with shadow wages as, based on common sense, it is hardly possible to assume that 50-60% of commuters work for a wage that is below the poverty line.

In addition, it was found that the average income of women was 2111 RUB which is 18.6 percent less than that of men whose average income is 2503 RUB. Thereby on the scale of Moscow region there are signs of gender discrimination according to the level of income.

The distribution of the average length of commuting journeys depending on commuter income is shown in Figure 6, right. The first data point on the graph includes commuters with cash-in-hand wages and hence with unknown actual income.

[Figure 4. Top: Distributions of time (left) and distance (right) of commuter journeys in Moscow region with men - women differences shown on the bottoms of figures. Bottom: 2D distribution.]
the obvious conclusion from this data point is that it confirms the fact that a cash-in-hand wage is a global phenomenon happening at all working places and, as a result, does not depend on commuting length. Due to the nature of the data the last bins also spoiled by essential amount of commuters with unknown wages, but high income from non-wage assets (shares, property, dividends, etc). After exclusion of these two problematic boundary points the average length of commuting journeys grows with increasing income. It means that high-skilled workers (HSW) commute farther and longer than low-skilled workers (LSW), which is in line with most theories of commuting. There are various explanations of this hypothesis. The mathematical approach [4] shows that the inclusion of travel times in the utility function of the Standard Urban Model (SUM) [8] of individual choice leads to an increase in the length of trips for HSW. In the theory of job search [10], the situation is explained by the spatial factor (coverage effect) of job search efficiency. Spatial spread of HSW work locations is higher, therefore HSW are forced to scan bigger areas in job searches. In addition, the benefits of higher wages at a new place of work are higher for HSW, wherefore they are ready to cover longer distances.

Figure 7 shows the distribution of average incomes of commuters depending on their age. The maximum income is earned by workers of 25-55 years old. Inside this group a peak is observed at 27-28 years, and a slight increasing slope between 30 and 50 years. The steep slope on the left of the figure corresponds to the careers of youngsters starting paid jobs, while the slope on the right might correspond to wage discrimination of aged workers forced to take less well paid positions. It is interesting that the shape of the curves is the same for both men and women, while the difference in amplitudes undoubtedly demonstrates the aforementioned gender discrimination.

The present work demonstrates application of GIS technologies to the analysis of commuting in Moscow region. The use of such an approach became possible due to the ability to reconstruct the commuting journeys of 0.8 million workers using GIS analysis.

The geospatial patterns of home and work locations of commuters, as well as their transportation maps, have been visualized. Reconstructed GIS information in combination with micro data have been used to analyze commuting from a point of view of the individual characteristics of workers: gender, age and income. On average, commuters cover 50 km between home and work, spending one hour one way. It was found that women travel to smaller distances than men. At the same time, women’s wages are 20 percent lower than those of men. The farthest and the longest trips are made by commuters of 22-30 years old, after which mobility decreases steadily with age. Finally, highly skilled commuters with high wages commute farther and longer than low skilled and low paid workers.

The approach proposed in this work has great potential for further...
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References