

1 **Massachusetts Regional Household and Labor Force Projections and Subregional Allocation**
 2 **Documentation**
 3 **Prepared for MassDOT Office of Transportation Planning**
 4 **August 2023**

5 **Contents**

6 INTRODUCTION..... 2

7 REGIONAL POPULATION, HOUSEHOLDS, AND EMPLOYMENT PROJECTIONS 3

8 UMDI Population Projections 3

9 Regional Households and Labor Force 4

10 *ACS PUMS Geographies 4*

11 *Household Population..... 4*

12 *Labor Force 5*

13 Household Types 6

14 PUMS Reweighting 6

15 Household Control Totals 6

16 Employment Control Totals 7

17 SUBREGIONAL ALLOCATION AND SMALL-AREA FORECASTS 8

18 About UrbanSim..... 8

19 Model Geography 9

20 Model Inputs 9

21 *Existing Households and Housing Units 10*

22 *Employment 10*

23 *Block-level Capacity (Density and Developable Land) 10*

24 *Travel Times and Job Accessibility 11*

25 *Massbuilds Development Pipeline..... 11*

26 UrbanSim Structure and Sub-Models 12

27 *Household Location Choice Models..... 13*

28 *Employment Location Choice Models..... 13*

29 *Residential Development Project Location Choice Models..... 14*

30 *Real Estate Price Models 14*

31 Model Calibration 15

32 Adjustments 15

33 Vacancy Rate Targets and Unavailable Vacant Units 16

34 Post-Processing of Household Population 16

35 *Household Size Adjustments..... 16*

36 *Age Shift Adjustments 17*

37 *Person Record Duplication/Deletion 17*

38 Group Quarters Population..... 18

39

40

41 **INTRODUCTION**

42 In May 2023, the Massachusetts Department of Transportation (MassDOT), the University of Massachusetts
43 Donohue Institute (UMDI), and the Metropolitan Area Planning Council (MAPC) completed new population,
44 household, and employment projections out to the year 2050 to support regional transportation modelling,
45 project planning, and design. Population and industry projections at the regional level were created by
46 UMDI, while household and labor force projections were developed by MAPC under contract to MassDOT.
47 MAPC then allocated households and jobs to census blocks across the state using a land use model called
48 UrbanSim. The resulting products paint a highly detailed picture of demographic change, industry shifts, and
49 land development over the coming thirty years if current trends continue.

50 This document details MAPC's methods for the development of regional household and labor force
51 projections, and for the allocation of households and jobs to sub-municipal geographies. UMDI's methods for
52 the creation of regional population and industry forecasts are described elsewhere in documentation
53 prepared by UMDI.

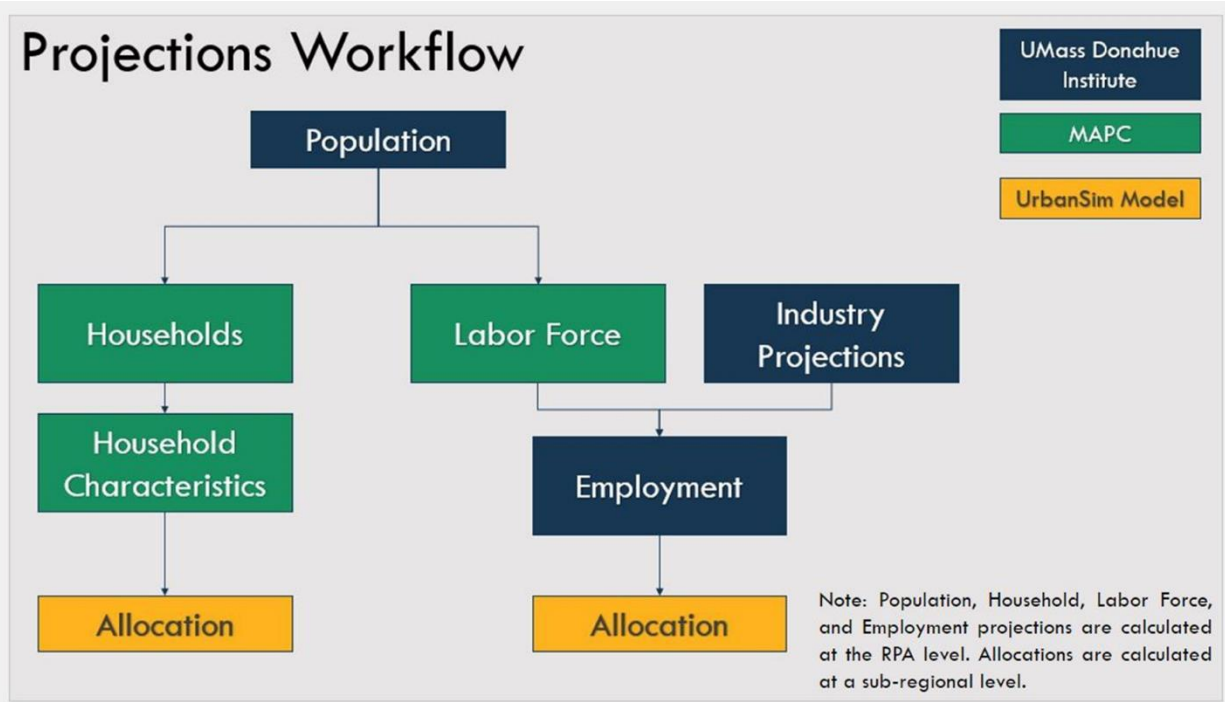
54 A simplified version of MAPC's projections process is as follows:

- 55 • Based on UMDI's regional population projections, MAPC first estimated the number of people
56 in households by age group, and the number of people in the labor force for each Metropolitan
57 Planning Organization (MPO) region and forecast year using values computed using data from
58 the U.S. Census Bureau's American Communities Survey (ACS) 2015-2019 Public Use Microdata
59 Sample ([ACS PUMS](#)). These estimates assume, generally speaking, that the likelihood someone
60 of given age is in group quarters, in the labor force, or head of a household remains relatively
61 steady over time.
- 62 • Labor force projections were further cross classified by age, educational attainment, and the
63 labor force status of individuals. Household projections were further cross classified by age of
64 householder, household type, number of persons in a household, and the presence of children in
65 a household.
- 66 • The initial household estimates are further adjusted to match data from the 2010 and 2020
67 Decennial Censuses as closely as possible. This includes global headship rate adjustments to
68 match 2020 household counts, targeted adjustments to headship rates for specific age groups
69 and types, and formulation of synthetic households for each municipality in 2010 so that the
70 base year households match 2010 census counts on all key characteristics.
- 71 • Regional labor force estimates, by age and educational attainment, were provided to UMDI to
72 serve as a constraint on employment growth in the state and its regions. Overall employment in
73 Massachusetts therefore grows at the same rate as the available labor force implied by the
74 population projections.
- 75 • Once regional household and employment totals are finalized, the UrbanSim model allocates
76 employment and households to sub-regional geographies. UrbanSim is a comprehensive urban
77 simulation model that forecasts future growth of households and jobs based on location choice
78 models that incorporate information about existing and planned development, access to
79 amenities and labor, zoning and environmental constraints, and resident demographics. The
80 model includes a "population synthesizer" that takes the aggregate household estimates
81 prepared by MAPC and breaks them down into individual households with specific
82 characteristics for size, workers, income, and children. Unlike prior models created by MAPC,
83 growth is not allocated to municipalities and then to sub-municipal geographies; households and
84 jobs are allocated from the regional control total directly to census blocks. As a result, there are

- 85 no fixed municipal totals—the forecast for any city or town may change as inputs to the model
 86 are modified.
- 87 • Following completion of the UrbanSim allocation, household population counts are reviewed and
 88 adjusted to ensure consistency with the regional UMDI Population projections for each age
 89 group. This process entails adding or removing people from selected microsimulated households
 90 to meet population targets without changing essential household characteristics.
 - 91 • After this post processing, the microsimulated households are grouped by Transportation
 92 Analysis Zone (TAZ) and shared with the Central Transportation Planning Staff for travel
 93 demand modeling; and are summarized and published by municipality for other purposes.

94 The products can support the work of other state agencies and local governments in the preparation of other
 95 regional planning applications given the detail of the household and labor force characteristics. Data
 96 created via this process can be modified through changes to the various methodological assumptions (e.g.,
 97 headship rates, labor force participation rates, education levels, etc.) to test alternative projection scenarios.

98 The following diagram summarizes the workflow for production of the 2023 LRTP projections:



99

100 **REGIONAL POPULATION, HOUSEHOLDS, AND EMPLOYMENT PROJECTIONS**

101 **UMDI Population Projections**

102 As with the last two rounds of statewide socioeconomic projections, MassDOT contracted with UMDI to
 103 produce population projections by sex and five-year age groups (e.g., 20–24-year-old) for each decade,
 104 out to 2050. UMDI’s population projections represent the total population in each sex and age group for
 105 each Metropolitan Planning Organization (MPO) geography. The projections are based on a [cohort](#)
 106 [component](#) methodology that uses data about births, deaths, and migration to forecast future demographic
 107 change. Put simply, the model assumes the rate at which people of each age group in each MPO region
 108 move in, move out, have kids, and pass away stays about the same as it was during the period from 2015
 109 – 2020. Over this period annual births declined while deaths, even prior to the pandemic, were rising.
 110 Domestic migration into Massachusetts declined throughout the period while international immigration was
 111 rising until 2017 before falling precipitously afterwards. On net, the state saw only small growth in its natural

112 increase (births minus deaths) in the population and a decrease in net migration by the end of the period
113 used to anchor the population projections. This method neither assumes anything about major changes
114 regarding where people choose to live or when they have children nor does it account for economic and
115 employment factors such as housing prices, available jobs, or remote work. While we know all these things
116 have changed since the pandemic, we neither know exactly how nor whether those changes will endure.

117 Given these caveats, it is useful to think of the UMDI projections as a “trends extended” scenario of
118 Massachusetts’ demographic future; one that represents *where the state was headed prior to the pandemic*. It’s
119 not necessarily the *most likely* outcome, but it provides a plausible and documented baseline for thinking
120 about the future. With this baseline, we can better estimate how Covid-19 (or any number of future
121 disruptions) will have effects distinct from those driven by massive demographic forces such as the aging
122 Baby Boomers and the relatively small size of Gen Z compared to Millennials.

123 Information about the UMDI’s population projection methodology and the MPO level population projections
124 can be found [here](#). It should be noted that UMDI methods use a cohort component method applied at the
125 municipal level, the results of which are aggregated to the 13 MPO regions. In other words, UMDI produces
126 municipal population projections and then compiles them into the regional totals needed by MassDOT and
127 MAPC for this process. While UMDI’s *municipal* population projections were created through the course of
128 the MassDOT-funded effort, they are based on recent patterns of migration, births, and deaths and do not
129 account for anticipated development, zoning constraints, or the impact of amenities and access on location
130 choice – so they do not match municipal population in household projections prepared by MAPC. The
131 municipal projections published separately by UMDI forecast the population for each municipality if recent
132 trends are unconstrained and unaffected by land use and development factors.

133 **Regional Households and Labor Force**

134 Following receipt of the regional population projections from UMDI, MAPC’s next step is to estimate the
135 number and type of households likely to form in each region, as well as the number of people in the labor
136 force. Households are classified by the age of the householder (i.e., the head of household), number of
137 people in the household, presence of children under age 18, and income group (all income data is adjusted
138 to a 2013 USD base year). MAPC also estimates the number of civilians in the labor force in each MPO
139 region. MAPC’s work achieves these ends using the Census Bureau’s American Community Survey (ACS) data
140 to construct multipliers which are applied to the population projections produced by UMDI. The following
141 sections provide a detailed account of how these multipliers are created, implemented, and adjusted to
142 create the input file for the UrbanSim microsimulation model.

143 *ACS PUMS Geographies*

144 Household and labor force multipliers for each region are based on ACS Public Use Microdata Sample
145 (PUMS) data. Each PUMS release covers a five-year period and is available at a geographic level called
146 a Public Use Microdata Area (PUMA) which contains at least 100,000 residents. This is done to maintain the
147 privacy of ACS respondents, since the data made public in PUMS contains specific information about
148 individual persons and households. PUMAs are usually larger than municipalities and generally do not
149 adhere to county or regional boundaries (with a few exceptions). The geographies of PUMAs also change
150 every decade; PUMAs boundaries for 2000, 2010, and 2020 are not always aligned with each other.
151 Therefore, MAPC has constructed crosswalks that assign each decade’s PUMAs to MPO regions. PUMAs for
152 each region are then pooled to create a regional group of PUMS records used for household and labor
153 force estimation. Please refer to [Appendix A](#) for the PUMA 2000 and 2010 to MPO crosswalks.

154 *Household Population*

155 To produce the projected number of households and people in the labor force, MAPC first converts the
156 projected total population, created by UMDI, into the total population of civilians in households, excluding

157 military personnel and residents living in group quarters (dorms, correctional facilities, nursing homes, etc.)
158 We adhere to the definition of “household” used by the Census Bureau (more information [here](#)) which is
159 included as a variable in the PUMS data (see the PUMS data dictionary). To convert the UMDI data, we first
160 determine the proportion of the population living in households as a share of total population by age group
161 and MPO. Using data from the 2010 Decennial Census SF1 file, we subtract the total number of people (by
162 five-year age group) in the military (since they do not count in the civilian population or live in households
163 while on active duty) and in “group quarters” (more information [here](#)) from the total population and then
164 divide the number of people in households by the total population by age and MPO. The resulting rate is
165 the proportion of the population in households by age and MPO in 2010. To produce the **projected**
166 household population, first, the total number of people in the military (using the 2010 data) are subtracted
167 from the population in the forecast years. Then, the household population proportions are applied by age
168 and MPO to produce the projected population in households for each projection year.

169 *Labor Force*

170 After estimating the projected household population by age group, MAPC applies educational attainment
171 and labor force participation rates to each age group. Following consultation with MassDOT and the
172 Massachusetts MPOs, MAPC used 2015-2019 ACS PUMS data to develop rates for the labor force model.
173 This is the most recent pre-pandemic data available and represents what many economists consider a point
174 in time the nation was at “full employment” (i.e., where virtually all residents willing and able to work are
175 either employed or in between jobs). During this period, the statewide labor force participation rate was
176 [equal to levels that existed before the Great Recession](#) and [unemployment was at a 20-year low](#). Like the
177 [Bureau of Labor Statistics](#), MAPC uses these “full employment” conditions as an assumption for future labor
178 force projections.

179 One factor in the labor force growth since 2010 was increasing participation rates for older individuals. The
180 Bureau of Labor Statistics published a [research report](#) which concludes that the labor force participation rate
181 (LFPR) for older adults is likely to continue increasing between 2020 and 2030. In consultation with
182 Northeastern University Professor and UMDI partner Alan Clayton-Matthews, MAPC decided to adjust the
183 LFPR for members of older age groups to reflect projected structural changes to the labor force. The
184 adjustment factors for those are located [here](#). It should be recognized that changes in labor demand,
185 retirement fund volatility, and other factors may lead to other outcomes for the labor force participation
186 rates of older adults. Furthermore, the projections do not make assumptions about cyclical factors that affect
187 economic activity or further evolution of educational attainment or labor force participation rates for younger
188 workers, both of which are highly uncertain given potential impacts of automation, artificial intelligence,
189 climate change, international turmoil, and other factors. These topics can be explored in future economic
190 scenarios.

191 The labor force model begins by estimating the likelihood (in 2015 – 2019) of a person in each age and
192 sex category to achieve a given educational attainment level, classified as “High School or less”, “Associate’s
193 degree or some college”, “Bachelor’s Degree”, and “Master’s or higher.” These rates were applied to the
194 future population by age and sex categories; then labor force participation rates for each age and
195 educational attainment were applied to the number of persons at each education level. This rate is also
196 derived from the 2015-2019 ACS PUMS data about “Employment status,” for each age and educational
197 attainment group, using records where the individual responded that they are either working for pay or
198 seeking/available for work, and incorporating the older worker rate adjustments described above.
199 Educational attainment and participation rates are constructed for each MPO region and applied to the
200 household population data by age group and sex. The resulting output is the number of people in households
201 in the labor force by sex, age group, and educational attainment.

202 It’s important to note that the labor force participation rate derived from ACS PUMS data is not the same
203 as the one derived from the Current Population Survey (CPS) which is generally used as the “official” rate

204 published by the Bureau of Labor Statistics (BLS). The [PUMS-based definition diverges from the CPS](#)
205 [definition](#) in one key regard: The ACS instrument asks people if they are actively looking for work and
206 available to take a job if offered one, but does not ask about the nature of the job search. The CPS
207 questionnaire probes to see if people are actively looking for work (interviewing, calling contacts, etc.) versus
208 passively looking for work (for example, looking at want ads in the newspaper). In the CPS, a person is
209 unemployed only if that person responded affirmatively to having engaged in one or more active methods
210 of job search. As a result, ACS estimates a larger number of unemployed respondents in the population—
211 and a correspondingly higher labor force participation rate—than the CPS.

212 **Household Types**

213 For these projections, MAPC defines three general Household Types: a person living alone, households with
214 two or more adults and no children, and all households with one or more children under 18. These types
215 were chosen because they are assumed to have very distinct housing needs and household location choice
216 preferences. Following consultation with MassDOT and the Massachusetts MPOs, MAPC used 2015-2019
217 PUMS data to develop region- and age-specific headship rates, defined as the probability of a person of
218 a given age being the head of a given type of household. Headship rates for the three groups of households
219 described above are applied to the total population in households for each forecast year. The resulting
220 product is an initial estimate of households by age of head of householder and household type for each
221 region and forecast year (2020, 2030, 2040, 2050). These household forecasts are used as targets in the
222 PUMS “reweighter”.

223 **PUMS Reweighting**

224 After determining the number of households and persons in the labor force, additional work is needed to
225 create detailed household estimates that are needed for UrbanSim inputs. This detail is created using a
226 process called “PUMS reweighting.” The ‘reweighter’ is a method created by Alan Clayton-Matthews,
227 Associate Professor of Economics and Public Policy at Northeastern University and Senior Research Associate
228 at the Dukakis Center. This technique adjusts the household weights of PUMS data so that the resulting
229 collection of weighted households matches a set of user-defined targets. The user can specify known
230 characteristics of future residents/households and use the method to determine what more detailed
231 characteristics of a synthetic future population would be. This method ensures consistency with projections
232 created in different parts of the model. For a more detailed explanation of how the PUMS reweighter works
233 see the UMDI employment projections documentation. The ‘reweighter’ has three input tables which act as
234 targets for the future reweighted sample to hit: persons by age group; persons by educational attainment,
235 age group, and labor force status; and households by householder age and household type. Age categories
236 for the input tables are classified as 5-year age groups. The household types match the three household
237 types created in the household model. The microdata sample to be reweighted are the [2015-2019 ACS](#)
238 [PUMS data](#), grouped by the PUMAs that make up each MPO (see Appendix A).

239 **Household Control Totals**

240 The outputs of the reweighter are a list of PUMS “serial numbers” (unique identifiers for households),
241 “sporder” designations (a number indicating the person record within a household), and the adjusted person
242 and household weights for each forecast year. Those are merged to the original 2015-2019 ACS PUMS
243 records. We then create a cross tabulation of the total number of households by the number of people in
244 the household, presence of children in the household, the age of the head of household, and the household
245 income group of the household for each year and MPO region. To interpolate data for the interdecadal
246 years we use the linear interpolation method. UrbanSim uses this cross tabulation as the input to its household
247 allocation model. The household control totals for the Statewide UrbanSim model have Duxbury, Hanover,
248 Pembroke, and Stoughton assigned to the OCPC MPO/RPA whereas in the MAPC UrbanSim model they are
249 included in the MAPC MPO/RPA. Additional adjustments to these estimates are described below.

250 The first adjustment harmonizes MAPC’s household estimates with counts from the 2010 and 2020 Decennial
 251 Censuses. For a variety of reasons (geographic misalignment, changing headship rates, sampling errors, etc.),
 252 household estimates produced with PUMS data don’t exactly match enumerated households in 2010 or 2020,
 253 even when using the same base population. Therefore, we compare the total number of households
 254 enumerated in the decennial census to the number estimated by our method. The ratio between these two
 255 figures becomes an adjustment factor applied to every age- and type- headship rate for those decades. As
 256 a result, MAPC’s total household estimates match census data for 2010 and 2020 at the regional level.

257 For purposes of land use allocation (described below), MAPC interpolates the decadal household
 258 estimates/forecasts to develop an annual estimate of household counts for each household type for each
 259 year in the forecast period. For the post-2020 period, a linear interpolation between decadal household
 260 forecasts is adequate for modeling purposes. For the period from 2010 – 2020, adjustments were made to
 261 align household growth with observed housing unit construction activity. Household formation was slower in
 262 the earlier part of the decade and then accelerated as the pace of housing unit production increased to
 263 meet demand. In order to match estimated household growth to housing delivery, MAPC used the [Massbuilds](#)
 264 development pipeline database to estimate the cumulative total of new units built in each region in each
 265 year between 2010 and 2020. We calculated the proportion of housing units developed in or before each
 266 year compared to the total number of housing units in Massbuilds for the whole decade. The resulting yearly
 267 rates are applied to the total household change between 2010 and 2020, so that household growth matches
 268 the rate of housing production in Massbuilds. While Massbuilds is only a sample of housing units and a proxy
 269 for household formation, this method ensures that estimated household growth does not rapidly outpace
 270 observed housing production during the early part of the decade. Adjusted interpolation rates for each
 271 region can be found [here](#).

272 **Employment Control Totals**

273 Employment forecasts for each region were prepared by the UMDI Population Estimates team and Alan
 274 Clayton-Matthews. Forecasts were prepared for each decade and for detailed industrial categories, then
 275 summarized to ten large “Super Sectors.” To infer interdecadal data in years after 2020, we linearly
 276 interpolate the data between each decade and each Super Sector. For interdecadal years between 2010
 277 and 2020, the data produced by the process described below is used instead of a linear interpolation. The
 278 ten Super Sectors are listed below and the constituent 2-Digit NAICS codes for each are listed in [Appendix](#)
 279 [B](#).

SuperSector	NAICS Codes	Constituent Sectors	UrbanSim Model Segment
Construction	23	Construction	1
Education and Health Services	61, 62	Educational Services, Health Care and Social Assistance	2
Finance	52, 53	Finance and Insurance, Real Estate Rental and Leasing	3
Public Administration	92	Public Administration	4
Information	51	Information	5
Retail, Leisure, and Hospitality	44, 45, 71, 72	Retail Trade; Arts, Entertainment, and Recreation; Accommodation and Food Services	6
Manufacturing	31, 32, 33	Manufacturing	7
Other Services	81, 11, 21	Other Services, agriculture, mining	8
Professional and Technical Services	54, 55, 56	Professional, Scientific and Technical Services; Management of Companies and Enterprises;	9

		Administrative and Support and Waste Management and Remediation Services	
Trade, Transportation, Warehousing	22, 42, 48, 49	Utilities, Wholesale Trade, Transportation and Warehousing	10

280

281 The launch year for the employment projections is 2020; specifically, employment estimates for each region
 282 at the beginning of 2020, before any Covid-related job losses. These employment estimates were based on
 283 municipal-level employment (by sector) from [ES-202](#). However, ES-202 data suppresses employment
 284 estimates for sectors with a small number of establishments in any given geography. As a result, the sum of
 285 the sector-level employment for a city or town may not match up to the total employment for that
 286 municipality; and the sum of sector-level employment for all cities and towns don't match up to statewide
 287 totals for that sector. To ensure that existing jobs in a given municipality or region were accounted for in the
 288 base year control totals, MAPC adjusted ES-202 employment estimates as follows: we used establishment-
 289 level data from Data Axle to augment sectoral municipal employment estimates where suppression was
 290 suspected; we compared municipal, county, and state data for each sector to impute how much employment
 291 was suppressed at each level; and then we distributed the imputed employment to municipalities based on
 292 the augmented sectoral totals. The result is a dataset in which municipal employment in each sector adds up
 293 to the state total employment. These municipal totals were then summed for each MPO region and provided
 294 to UMDI as the launch year employment for their forecasts.

295 **SUBREGIONAL ALLOCATION AND SMALL-AREA FORECASTS**

296 **About UrbanSim**

297 To develop small-area forecasts of population, employment, and land use change, MAPC has developed a
 298 customized implementation of the UrbanSim land use allocation model. [UrbanSim](#) is an established and
 299 validated land use model specifically designed to explore how changing demographics, economic growth,
 300 land use policies, and infrastructure investments may affect the development and character of cities and
 301 regions. As a modeling framework, UrbanSim has been adopted by nearly two-dozen cities and regions in
 302 the United States for land use forecasting. UrbanSim, Inc. makes the model available as a cloud-based
 303 subscription service called UrbanCanvas and provided support to MAPC for the estimation, calibration, and
 304 implementation of models customized to the Massachusetts and Metropolitan Boston context.

305 UrbanSim simulates the interactions among households, businesses, and developers within real estate markets.
 306 By modeling how households trade off housing costs, ease of access to jobs, and neighborhood amenities,
 307 UrbanSim simulates households' choices to select a housing location, housing type, and whether to rent or
 308 own. UrbanSim also simulates employment dynamics, including firms' location choices, as well as real estate
 309 developers' choices of what kind of buildings to build, where, and when, and whether to redevelop existing
 310 properties. Zoning and other policies constrain what developers can build, and assumptions about how
 311 transportation plans may affect the speed or cost of travel influence the attractiveness of different locations
 312 for households and firms. This influences prices, rents, and the market conditions for new development or
 313 redevelopment.

314 UrbanSim models individual households, jobs, and building construction, as well as their changes from one
 315 year to the next due to economic changes, policy interventions, and market interactions. UrbanSim simulates
 316 changes for each year, and the results of one year provide the starting point for the next simulation year.
 317 This method closely replicates the way that urban areas evolve, year over year, with mismatches between
 318 the supply and demand of housing and jobs.

319 Unlike prior land use models used by MAPC, growth is not allocated from regions to municipalities and then
 320 to sub-municipal geographies in a hierarchical manner. Rather, new households and jobs are allocated from

321 the regional control total directly to census blocks. As a result, there are no municipal totals that can be
322 specified or maintained. When inputs to the model are changed (new development information, updated
323 zoning, etc.), the entire regional allocation is re-run; even small, localized updates may have ripple effects
324 that spread across municipal boundaries. In this way, the model simulates real-world conditions more
325 accurately than an approach relying on fixed municipal 'control totals,' since major development projects
326 and transportation improvements influence market and neighborhood conditions without regard to city and
327 town boundaries.

328 **Model Geography**

329 MAPC developed two different UrbanSim models: one covers the 101 cities and towns in the MAPC region;
330 the other encompasses the entire state. Location choice and price models for the two different model sets
331 were specified and estimated separately; the MAPC model is tailored to the conditions at the core of the
332 Metro Region, while the statewide model is better suited to forecast growth through the diverse regions of
333 the state.

334 The statewide model is structured so that the user can provide regional control totals for each of the 13
335 Metropolitan Planning Organization (MPO) regions, and growth is allocated within each region. Four
336 municipalities (Duxbury, Hanover, Pembroke, and Stoughton) are within the MAPC region but belong to the
337 Old Colony Planning Council MPO. Regional totals, prepared at the MPO geography, can be input without
338 alteration into the statewide model region for the OCPC MPO. For the MAPC model, additional steps are
339 needed to account for the discrepancy between the 101 cities and towns in the RPA model region and the
340 97 cities and towns in the MPO control total region. To resolve this discrepancy, MAPC applies a proportional
341 adjustment as follows. For the household controls, the MAPC model (MAPC101) cross-classified household
342 data are multiplied by a factor that represents the proportional increase in the number of households in the
343 MAPC101 geography relative to the MAPC97 geography. Factors differ in each year for which
344 administrative data of the number of households in each region is available (2010-2020). Data in the
345 forecast years have the 2020 factor applied. A similar track is taken for the employment control totals. ES-
346 202 employment data is used to fill in the 2010-2019 baseline data for the MPO by Super Sector, since
347 that data is available at the municipal-level. Using the 2020 ES-202 employment data, we construct an
348 adjustment factor for the MPO level projections. As before, we calculated the proportion of employment in
349 each Super Sector using the MAPC101 and MAPC97 aggregations. That proportion is then applied to the
350 decadal forecasted data. For operational purposes, both models are run and the results of the statewide
351 model for the Boston MPO's 97 municipalities are discarded and replaced with the results from the MAPC
352 region model. (The full 101-municipality MAPC model results are used for regional analysis and planning
353 not directly tied to the Long Range Transportation Plan.)

354 Within each region, the models allocate households and employment at the 2010 Census Block level, a total
355 of nearly 55,900 blocks in the MAPC model region and 118,700 blocks in the statewide model. The model
356 can output results at multiple summary geographies including counties, municipalities, 2010 Census Tracts,
357 and Transportation Analysis Zones (TAZs) used for the CTPS Travel Demand Model. When summary
358 geographies do not follow block lines completely (e.g., TAZs, some municipalities) an area-weighted division
359 of the census block is used to divide the results.

360 The MAPC region model can accommodate specified sub-regional control totals for each of the [MAPC](#)
361 [Community Types](#). This allows MAPC to create scenarios with specified allocation of growth across the
362 Community Types as opposed to one control at the regional level.

363 **Model Inputs**

364 *Existing Households and Housing Units*

365 The launch year of the model is 2010. It is essential, therefore, to have an accurate picture of housing units
366 and households in existence at that time. Information about households and housing units was drawn from
367 the 2010 Census and the 2009 – 2013 American Community Survey. MAPC used the reweighter described
368 above to provide municipal level control totals by age of householder, household size, income, and household
369 type to the UrbanSim model so that the household characteristics in the launch year match records from the
370 Decennial Census. Launch year housing unit estimates for each block set to match the 2010 Census.

371 *Employment*

372 2010 Employment data by sector is drawn from record-level establishment data provided by InfoGroup
373 (now Data Axle) in 2011. This dataset was reviewed and edited by MAPC and CTPS shortly after receipt
374 in 2011 to resolve major discrepancies in employment and location. However, it was not reconciled with
375 2010 ES-202 employment data, so job counts do not match at the municipal, regional, or sector level.

376 MAPC also acquired 2019 employment data from Data Axle and performed extensive Q&A on this data
377 to improve its accuracy. Data was provided to all MPO partners and extensive review and feedback was
378 provided by MPO and municipal staff. The main applications of this information was to provide sub-municipal
379 estimates of employment by sector for model calibration, and to fill gaps in the suppressed municipal ES-
380 202 employment data. Data Axle data was used to help reconcile municipal ES-202 data with county and
381 state estimates, so that the employment counts are aligned at all geographies and across all sectors. The
382 2019 employment estimates are used as the pre-pandemic employment counts for each municipality; in other
383 words, the 2020 employment is effectively employment as of January 1, 2020, prior to the pandemic.

384 *Block-level Capacity (Density and Developable Land)*

385 A key assumption in the UrbanSim model is the number of housing units and jobs that could be accommodated
386 within each census block. This assumption is termed “capacity” and is a function of two inputs: density
387 assumptions (as dwelling units per acre or floor area ratio) and area of developable land. Multiplying the
388 density times the developable land area yields units and commercial area capacity. MAPC’s [Zoning Atlas](#)
389 serves as the principal source of density assumptions in the MAPC model. This database contains information
390 for each of the 101 MAPC municipalities about maximum dwelling units/ acre and maximum commercial
391 floor area ratio (FAR). Allowable density estimates were based on interpretation of local zoning
392 codes/ordinances with a multi-step process to identify the highest allowable density if there are multiple
393 development options. Detailed description of the interpretation methods can be found in the [Zoning Atlas](#)
394 [technical appendix](#). In January 2023, MAPC solicited input from all cities and towns in the region, asking
395 municipal staff and boards to review the density estimates in the Zoning Atlas and provide feedback or
396 updates as necessary. This feedback was incorporated into the assumptions used for the LRTP projections. It
397 should be noted that not all overlay zoning districts were interpreted and incorporated in time for the LRTP
398 projections and, therefore, the assumptions may over- or under-estimate allowable density in blocks where
399 overlay districts provide more flexibility or impose additional restrictions stronger than the underlying base
400 zoning.

401 To estimate developable land area in each block, MAPC removed areas with permanent and inviolable
402 constraints that prevent any future development. These areas include the following:

- 403 • Permanently protected open space
- 404 • Water bodies and wetlands (source)
- 405 • Cemeteries

406 • Public Rights of Way

407 The presence of existing development was not considered to be a permanent constraint on development,
408 since reuse and densification of most land uses is feasible if market conditions warrant. It should also be
409 noted that the model estimates total capacity and compares this to existing development to determine the
410 net additional units or commercial floor area that could be accommodated. If the existing development
411 already meets or exceeds total capacity, no additional units can be placed by the model. An extensive
412 review process was conducted by all cities and towns in MAPC to confirm the accuracy of these capacity
413 constraints.

414 For the statewide model (SWM), MAPC collaborated with the state's other RPAs to compile a database of
415 zoning district boundaries for non-MAPC municipalities. While some cities and towns provided current zoning
416 data, others were unresponsive and MAPC resorted to using data originally published by MassGIS in 1999.
417 Given the available timeline and budget, it was not feasible to manually interpret zoning bylaws and
418 ordinances from the 250 non-MAPC municipalities. Therefore, MAPC and UrbanSim staff developed methods
419 to estimate residential and nonresidential development densities based on recent development as recorded
420 in MassBuilds and municipal assessor data.

421 We first looked at the existing residential densities in each zoning district using the [MassGIS Level 3 Assessor](#)
422 [Parcel Database](#). We then considered all recent developments over the last decade in each zone ([MassBuilds](#)
423 [developments](#)) as well any parcels in each zoning district where the 'built_date' was 2010 or later).
424 Developments that fell under 40B permitting were excluded from this analysis. Also excluded were any
425 outliers with densities or FAR greater than the 95% percentile of the initial estimates. These initial estimates
426 based on "currently built densities" were then provided to all RPAs for detailed review. After extensive
427 feedback and review from municipal and RPA staff, many additional direct manual adjustments were made
428 to revise the initial estimates. The final spatial data on zoning district boundaries and assumptions regarding
429 allowable densities used for the LRTP can be found here: [Residential dwelling units per acre \(maxDUA\)](#).
430 [Nonresidential Floor to Area Ratio \(maxFAR\)](#).

431 For nonresidential development, the maximum FAR for commercial development in a zone was converted
432 into a maximum employment capacity for each Census Block using the developed square foot-to-job factors
433 described in more detail in the section below on the Massbuilds Development Pipeline. The final estimates
434 for Census Block level capacities for residential units and employment used for the 2023 LRTP projections
435 can be found [here](#).

436 *Travel Times and Job Accessibility*

437 Estimated travel times and costs to different locations around the region are used as one input to the location
438 choice models described below. This information comes from the CTPS TDM19 travel demand model. The
439 outputs used for UrbanSim include travel times between each Transportation Analysis Zone pair for each
440 mode and time of day. Travel times and costs are combined into a single measure expressed in dollar terms,
441 with each hour of travel time equivalent to 18 dollars.

442 Once travel times are input, UrbanSim calculates the total number of jobs (of a given type) or households of
443 a given type that are within a specified travel time/cost threshold. This value is recalculated for each model
444 year/decade so that as new households and jobs are added, or as travel times increase or decrease, the
445 accessibility to opportunities at a given location change over time.

446 *Massbuilds Development Pipeline*

447 While the UrbanSim model forecasts the location of future homes and job growth based on historic patterns,
448 it can also incorporate information about development that is already planned or proposed (“scheduled
449 development” in model terms.) Fortunately, MAPC already has a robust picture of recent and anticipated
450 development, recorded in our collaborative online development inventory, [MassBuilds](#). This database
451 includes information about over 4,000 projects across the state either completed, underway, or planned.
452 Information about each project includes residential unit count, commercial square footage (by various types),
453 year complete, and point location, among other details. For the current forecasting effort, MAPC solicited
454 input from all the municipalities in our region as well as the other twelve regional planning agencies in
455 Massachusetts. Hundreds of records were created or updated through this process in early 2023. The
456 Massbuilds export used for the LRTP projections can be found [here](#).

457 To provide this data to the model, MAPC specifies the year the project is anticipated to be available for
458 occupancy, the count of units, and the number of on-site permanent jobs it is likely to support. Jobs are
459 estimated from commercial square footage using [a range of factors](#) (jobs per square foot of developed
460 space) that depend on the community type, local job density, type of development (commercial, industrial,
461 institutional), and employment sector, based on MAPC’s analysis of pre-pandemic occupancy patterns and
462 consultation with municipal and agency staff. At this time, UrbanSim accepts only total employment
463 associated with scheduled development, not jobs by sector.

464 When added to the UrbanSim model, these developments increase the number of units or employment spaces
465 available for occupancy, though whether they are filled with households or employees is determined by the
466 household and employment location choice models. Scheduled development can also exceed the unit or job
467 capacity for each block that was calculated from the zoning densities and developable land. In these cases,
468 the model overrides the existing capacity and adds the development project as specified. Each project can
469 be tagged with different scenarios, which allows us to create different model runs including/excluding certain
470 projects based on assumptions about policy choices or likelihood of happening. While scheduled
471 development adds units to a targeted block, there is no guarantee that those units will be occupied by
472 households if other available units in the region are more attractive to the household location choice models.
473 In a few select cases where it was judged that future development projects had a high likelihood of occurring
474 and of being occupied, MAPC used “adjustments”, in a procedure described below, to ensure that households
475 were allocated to the relevant census blocks.

476 **UrbanSim Structure and Sub-Models**

477 A general overview of the UrbanSim block model (as well as details about data needs and specifications)
478 can be found here: <https://cloud.urbansim.com/docs/index.html>

479 There are four main models in the UrbanSim system. Three of these are “location choice” models that drive
480 the placement of new agents (households, jobs, development projects) in the simulation. Then there is a “price
481 regression” model that estimates the variation of buildings’ values. These models are further subdivided into
482 submodels with different specifications depending on the category of the agent being modeled. For
483 example, for the household location choice model (HLCM), households are grouped into 18 different
484 ‘segments’ based on the age of the householder, the number of adults and/or children, and the household’s
485 income. Each segment’s sub-model has a slightly different specification (predictor variables), and estimated
486 coefficients for these variables.

487 Data sources for predictor variables include the Decennial Census, ACS, and PUMS data, Data Axle
488 employment data, municipal assessor parcel data, and others. Measures of access to jobs and amenities are
489 based on travel time and travel cost estimates (by mode) generated by the CTPS travel demand model,
490 TDM19.

491 *Household Location Choice Models*

492 As described above, MAPC prepares cross-classified household projections for each decade year and
493 interpolates the count of households with each unique combination of attributes for each year between the
494 decadal regional totals. UrbanSim prepares synthesized households and population for the base year and
495 then allocates new households for each one-year time step increment. When there is a net increase in the
496 number of households in a given category, those households are sampled from the microsimulated household
497 population and are placed into census blocks based on the HLCM. When there is a decline in the number of
498 households in a given category, a corresponding number of simulated households meeting those criteria are
499 selected randomly from across the region and are deleted from the model population.

500 Different household location choice models (HLCM) are specified for 18 distinct household segments defined
501 by age of householder, household type, and income. (Initially 21 different segments were defined for
502 modeling, but these were clustered for segments with similar characteristics and location choices.) The segment
503 groupings are:

- 504 • Age of Householder: <35, 35-64, and 65+
- 505 • Household Income: <\$35,000, \$35,000-\$100,000, and >\$100,000
- 506 • Household Type: Single adult living alone, Multiple adults no children, One or Multiple Adults with
507 children.

508 A full list of model specifications and coefficients is available [here](#).

509 Each household segment has a unique combination of variables and coefficients in the location choice model
510 used for estimating where new households will settle. The following variables are used in one or more of
511 the HCLMs for the MAPC model region:

- 512 • Household density (2010 Census Block Group)
- 513 • Proportion of units in single family vs. multi-family buildings (2010 Census Block, MA Land Parcel
514 Database)
- 515 • Mean rent (2008 – 2012 ACS Census Tract)
- 516 • Mean home value (2010 Census Block Group, Warren Group Real Estate Data)
- 517 • Mean year built, housing units (2010 Census Block Groups, MA Land Parcel Database)
- 518 • Total jobs within 45 minutes by transit, AM peak period (Traffic Analysis Zone (TAZ), CTPS Travel
519 Demand Model)
- 520 • Total households within [8 dollars in time terms] by auto (TAZ, CTPS Travel Demand Model)
- 521 • Total jobs within [7 dollars in time terms] by auto (TAZ, CTPS Travel Demand Model)
- 522 • Mean income (2008 – 2012 ACS Census Block Group)
- 523 • Census Block is within ¼ mile of transit station (2010 Census Block, MassGIS MBTA stops)
- 524 • Census Block is within 3000m of a major road (MassDOT Road Inventory)
- 525 • Percent of roads with sidewalks (MassDOT Road Inventory)
- 526 • Number of high-income households within 1500 meters (ACS 2008 – 2012 Census Block Group)
- 527 • Count of units built since 2010 within 1500 meters (SF-1 2010 Census Block Group, ACS 2008 –
528 2012 Census Block Group)

529 *Employment Location Choice Models*

530 The ELCM is comprised of 10 different employment location choice models, one for each SuperSector
531 described above. The model was estimated using 2011 establishment-level employment data from

532 InfoGroup (now DataAxle), which was reviewed and corrected by MAPC and CTPS for prior modeling
533 efforts. As with households, the decadal control totals for employment are interpolated linearly for each
534 sector to create projected employment counts for each individual forecast year. Similarly, the model places
535 any net increase in a sector's employment into blocks using the appropriate ELCM, whereas declines in a
536 given sector are resolved through random selection and deletion of jobs in that sector across the region.

- 537 • Residential unit density (2010 Census Block)
- 538 • Residential household density (2010 Census Block)
- 539 • Proportion of units in single family vs. multi-family buildings (2010 Census Block, MA Land Parcel
540 Database)
- 541 • Mean year built, housing units (2010 Census Block Groups, MA Land Parcel Database)
- 542 • Total households within 8 dollars in time terms by auto (TAZ, CTPS Travel Demand Model)
- 543 • Total jobs within 8 or 25 dollars in time terms by transit, AM peak period (TAZ, CTPS Travel Demand
544 Model)
- 545 • Census Block is within ¼ mile of transit station (2010 Census Block, MassGIS MBTA stops)

546 Model specifications and coefficients are available [here](#).

547 *Residential Development Project Location Choice Models*

548 The MAPC Region UrbanSim model also relies on a pair of residential development project location choice
549 models (RDPLCM). There is one model for Single Family development and another for Multi Family
550 development. These models are used to estimate the likelihood that a hypothetical developer will build new
551 units of either type in a given census block. The RDPLCM is run as the last sub-model in a given year, so the
552 resulting housing units are available in the subsequent year for households to choose in the household location
553 choice model. The following variables are used in either the single family or multifamily RDPLCM:

- 554 • Proportion of units in single family vs. multi-family buildings (2010 Census Block, MA Land Parcel
555 Database)
- 556 • Household density (block groups, 2010 Census)
- 557 • Mean year built, housing units (2010 Census Block Groups, MA Land Parcel Database)
- 558 • Number of units built since 2010 within 800 meters (block groups, 2010 Census, 2013-2017 ACS)
- 559 • Census Block is within ¼ mile of transit station (2010 Census Block, MassGIS MBTA stops)
- 560 • Total jobs within 8 dollars in time terms by auto (TAZ, CTPS Travel Demand Model)
- 561 • Number of high-income households within 1500 meters (Census Block, ACS 2008 – 2012)
- 562 • Mean rent (Census Tract, 2008 – 2012 ACS)
- 563 • Ratio of households to residential units AKA vacancy rate
- 564 • Total households within \$35 travel time via transit during AM peak (CTPS travel demand model)

565 Model specifications and coefficients are available [here](#).

566 *Real Estate Price Models*

567 The MAPC Region UrbanSim model relies on a pair of real estate price models (REPMs) to estimate
568 residential sales prices and rental rates. The value model was estimated using Warren Group transaction
569 data and the Rent model was estimated using ACS data.

- 570 • Residential unit density (2010 Census Block Group)

- 571 • Mean household income (2008 – 2012 ACS Block Group)
- 572 • Proportion of units in single family vs. multi-family buildings (2010 Census Block, MA Land Parcel
- 573 Database)
- 574 • Total households within 8 dollars in time terms by auto (TAZ, CTPS Travel Demand Model)
- 575 • Census Block is within ¼ mile of transit station (2010 Census Block, MassGIS MBTA stops)
- 576 • MAPC Community Type
- 577 • Mean year built, housing units (2010 Census Block Groups, MA Land Parcel Database)
- 578 • Census block is within 3000m of a major road (MassDOT Road Inventory)

579 Model specifications and coefficients are available [here](#).

580 **Model Calibration**

581 The models above have all been calibrated using data on the change in households and employment
582 between 2010-2020. The household location choice models are calibrated using the observed household
583 growth estimates between Census 2010 SF1 and Census 2020 PL-94 redistricting data, at the municipal level.
584 In the context of UrbanSim, 'calibration' involves making small adjustments to model coefficients to minimize
585 the value of an objective function (in the case of households, for example, the calibration seeks to minimize
586 the difference between the model's predicted household growth from 2010 to 2020 with the household
587 growth reported by the Census Bureau, for each municipality between 2010 and 2020). After calibration,
588 the model estimates of total households by municipality in 2020 are within a few percent for nearly every
589 municipality in the state.

590 The employment location choice models were calibrated using the combination of 2019 ES-202 data on
591 employment by municipality and 2019 DataAxle business location and employment data described in the
592 **Employment Control Total** section earlier in this document. The residential development project location
593 choice model was also calibrated using ACS 2009-2013 data.

594 As stated above, UrbanSim's calibration methods do not involve adding an error term to achieve a perfect
595 fit at every geography. Instead, coefficients of the estimated models are iteratively changed to achieve a
596 better fit with the calibration targets. This means that the 2020 forecast does not exactly match 2020
597 observed households or jobs at the sub-municipal level. When using the projections, it is advised that users
598 look at the change between 2020 and a forecast year, rather than using absolute numbers for 2020 or
599 afterwards.

600 **Adjustments**

601 While households and job locations are primarily determined by the location choice model, a user can also
602 specify a particular number of households or jobs of a given type be allocated to a specific block in a
603 specific year. These are termed 'adjustments' and are specified for a specific start year and end year range.
604 Adjustments can be specified on a single variable such as job sector, age of householder, or household
605 income. Cross classified household adjustments (e.g. 100 senior-headed households with incomes <\$50,000)
606 are not possible at this time.

607 An adjustment applies to the total number of households and jobs in a given block, so existing households
608 and employment must be taken into consideration when applying adjustments. For the year range provided,
609 the model will hold the block households or employment at the specified levels in the adjustment. After the
610 end year of the adjustment, the model is then free to add or remove households or jobs using the standard
611 allocation models.

612 MAPC used adjustments for two purposes: to improve the post-calibration fit with municipal and sub-
613 municipal households and jobs for 2020; and to ensure that households are allocated to large future
614 developments that are not fully occupied by the HLCM. A full enumeration of all adjustments applied to the
615 L RTP projections can be found [here](#).

616 **Vacancy Rate Targets and Unavailable Vacant Units**

617 UrbanSim allows users to specify a target regional vacancy rate, which determines the number of units built
618 by the RDPLCM. For purposes of the 2023 L RTP projections, we applied a regional vacancy rate of 4.5%
619 in both the MAPC and statewide models. However, the model does not distinguish between available vacant
620 units (those that are available for sale or for rent) and those that are vacant but effectively unavailable to
621 home-seekers. This latter category includes seasonal or second homes, vacation rentals, units that have been
622 sold or rented but not yet occupied, units used for commercial purposes, and housing for migrant workers. In
623 its default state, the model will allocate households to these “unavailable vacant” units even though they are
624 not actually available for occupancy, which then results in curtailed production of new units through the
625 RDPLCM. Since there are so many vacant units that are, as far as the model is concerned, available for
626 occupancy, it does not then build enough units to keep up with household growth.

627 To address this issue, MAPC undertook a process to distinguish available from unavailable vacancies at the
628 municipal level and remove the latter from the inventory of existing units provided to UrbanSim. We used
629 information from the 2010 Census and the 2015-2019 ACS to determine the share of vacant units in each
630 municipality that were classified as seasonal or occasional units, sold or rented but not yet occupied, or other
631 vacant units. We then applied this share to the total vacancy in each census block and removed that many
632 units from the total housing stock. This estimate of unavailable vacant units is held constant at the block and
633 municipal level over the forecast period and is added back to the projections of housing units to produce
634 estimates of total housing units in a forecast year. This prevents the model from filling existing non-available
635 vacant units with new households, resulting in vacancy rates for the ‘available’ unit counts that are more
636 reflective of reality.

637 **Post-Processing of Household Population**

638 As described above, MAPC provides detailed household forecasts to the UrbanSim population synthesizer,
639 which enumerates individual households matching the characteristics of the target totals. The provided
640 households are cross-classified by age of householder, presence of children under 18, household size (1, 2,
641 3, and 4 or more people), and other characteristics. While household counts and characteristics are matched,
642 the population contained within the synthetic households may exceed or fall short of the population in
643 household forecasts for any given forecast year and age group. **UrbanSim does not utilize population
644 control totals.** As a result, the microsimulated population in households that are allocated to match the
645 **household control totals** will always differ somewhat from the population totals provided by UMDI.

646 Due to the nature of the reweighter and population synthesizer, it is not possible to configure these tools so
647 they converge on a population that exactly matches the UMDI forecasts. The biggest discrepancies were
648 underestimates of the population in the 15-34 and 65-74 year old age groups, though there were also
649 over-and under-estimates for other age groups. As a result, MAPC took a multi-step and iterative process
650 to align the synthetic household population with the forecasted population.

651 *Household Size Adjustments*

652 The first step entailed adjusting the distribution of household sizes provided as inputs to the population
653 synthesized. Based on a review of the synthetic population age distribution and tests of the
654 shift/delete/duplicate methods described below, we determined that adjustments to household sizes were
655 needed to achieve convergence without resulting in unreasonably large household sizes. Specifically, we
656 reduced the number of single-person households and increased by a corresponding amount the number of

657 households with multiple adults and no children. The effect is that the same number of households incorporate
658 a larger population. This process was applied to two age ranges for the head of household: the 15-34 and
659 65-74 age groups.

- 660 - For the 15-34 age group the shift is implemented as such:
 - 661 ○ For households with three persons and no children in the household, the number of households
 - 662 across all income group segments was multiplied by 1.3 (a 30% increase).
 - 663 ○ For all households with four or more people and no children in the household, the number of
 - 664 households across all income group segments was multiplied by 1.5 (a 50% increase).
- 665 - For all single households, the number of households was decreased by the sum of the increase in
- 666 three- and four or more-person households. The decrease is distributed proportionally across all
- 667 income group segments based on the pre-adjustment income group distribution of households.

668 For the 65-74 age group the shift is implemented as such:

- 669 ○ For households with two-, three-, and four or more-person households with no children, the
- 670 number of households across all income group segments was multiplied by $(1 + [\text{Headship rate for households with two or more persons without children}])$
- 671
- 672 ○ For all single households, the number of households was decreased by the sum of the
- 673 increase in two-, three-, and four or more-person households. The decrease is distributed
- 674 proportionally across all income group segments based on the pre-adjustment income group
- 675 distribution of households.

676 The next adjustment is applied to the microsimulated population after allocation. The regional control total
677 household population and the UrbanSim microsimulated population are compared to determine
678 discrepancies for each five-year age group, for each MPO, and each decadal forecast year.

679 *Age Shift Adjustments*

680 In regions and forecast years where there are adjacent age groups with discrepancies different in sign (one
681 microsimulated age group has a surplus compared to the regional control totals and an adjacent age group
682 has a deficit of population), then population is shifted to the adjacent age group. In this process, the age
683 variable for individual person records from an age group with a surplus of population are altered by plus
684 or minus five years to shift them into an adjacent age category. For example, if the 25-29 age group has
685 a surplus population and the 30-34 has a deficit, enough 25-29 year old records are randomly selected
686 and 5 years are added to their age to eliminate either the surplus or the deficit. After this process is run,
687 MPO age group segments with remaining surpluses or deficits are passed to the "Duplication/Deletion"
688 script.

689 *Person Record Duplication/Deletion*

690 In the next step of the process, MPO age group segments with outstanding surpluses and deficits are brought
691 into alignment with the UMDI household population data by randomly selecting household members for
692 duplication or deletion from a subset of persons who represent the characteristics in the delinquent MPO age
693 group segments. Before the duplication/deletion process occurs, a subset of household members in each
694 segment is created. Depending on the age group, the criteria for inclusion in the subsets differ. For example,
695 the duplication subset for children is constrained to households with fewer than four previously existing
696 children (so as not to create households with an inordinately large number of children) and the deletion
697 subset for children is constrained to households with more than one child so as not to alter the household type
698 of said household (i.e. to go from a household with children to a single person household or a multiple adult
699 household without children). Duplication/deletion subsets for adults in households are less constrained. The
700 addition subset for each MPO age group is dictated by household type (multiple adult households without
701 children) and age of the household reference person (persons can be added to households with a reference

702 person who is 12 years younger or older than them). The deletion subset is constrained by the household
703 type and the number of preexisting persons in the household. Then the microsimulated data for each MPO
704 age group segment has person records either duplicated or deleted until parity with the UMDI household
705 population data is reached. The result is microsimulated data that reflects the original MPO-level household
706 population targets used in development of the household and labor force data used at the outset of the
707 projections process.

708 **Group Quarters Population**

709 As described previously, group quarters residents are subtracted from the total UMDI projections prior to
710 estimating and allocating future households. In order to estimate total population at the municipal level in
711 future years, it is necessary to allocate forecasted group quarters population from the region to
712 municipalities. The method for this allocation is quite simple: MAPC calculates the share of the MPO group
713 quarters population, by age group, residing in each municipality in the year 2010 (using decennial census
714 data.) This share is applied to the future year forecasts of MPO-level group quarters population by age;
715 the result is an estimate of group quarters residents by age in each municipality in each forecast year, which
716 sum to the MPO-level group quarters forecasts by age.