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4 **Main Manuscript for**

5 **Beyond efficiency: Sufficiency unlocks deep decarbonization of U.S.**
6 **residential sector**

7

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22 P.V. advised on all matters related to the building energy model, while A.M. and X.Y. advised on
23 the stock turnover model; S.N. performed analyses, created graphics, and wrote the initial draft;
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29

30 **This PDF file includes:**

31 Main Text
32 Figures 1 to 6
33 Table 1

34 **Abstract**

35 Decarbonization strategies in the residential sector have largely focused on lowering the carbon
36 intensity of energy supply and improving end-use efficiency. Sufficiency, defined as avoiding
37 unnecessary energy demand while maintaining well-being, remains largely unquantified in
38 national energy system analyses. We quantify how structural sufficiency (dwelling size and
39 housing form) and behavioral sufficiency (thermal comfort preferences via thermostat settings),
40 alongside efficiency and electrification, contribute individually and in combination to emissions
41 reductions. Using an integrated modelling framework that links U.S. housing stock turnover to
42 hourly building energy demand, we find that floorspace expands by 55% from 2020 to 2050 under
43 business-as-usual trends, increasing energy demand by 11%. Relative to this 2050 trajectory,
44 efficiency and structural sufficiency reduce demand by 18% and 31%, and by 45% when
45 combined. Adopting conservative, health-informed thermostat setpoints further reduces energy
46 demand by 31–33%. Sufficiency measures offer substantial mitigation potential, enabling deep
47 decarbonization while maintaining household well-being.

48

49 **Main Text**

50

51 **Introduction**

52

53 The U.S. residential sector accounts for approximately 18% of national energy-related CO₂
54 emissions¹, with space heating and cooling representing roughly half of household energy use².
55 Addressing residential energy demand for space-conditioning is therefore critical to U.S.
56 decarbonization efforts. Choices about housing infrastructure create long-lived lock-ins³ by fixing
57 dwelling size and built form, constraining the extent to which efficiency measures can reduce
58 energy demand. Over recent decades, U.S. residential floorspace per capita has increased by
59 roughly 56%, rising from about 600 square feet per person in the 1980s to more than 900 square
60 feet today.⁴ Public policies have shaped these trends in meaningful ways. As one notable
61 example, the Servicemen’s Readjustment Act of 1944 (also known as the GI Bill)—provided low-
62 interest mortgages that rapidly accelerated suburban expansion.² This contributed to a housing
63 stock dominated by single-family homes, which use approximately 30–60% more energy per
64 dwelling than multi-family units^{2,5}.

65 Mitigation strategies within the residential sector have largely focused on improving
66 energy efficiency—reducing energy use per unit of service through measures such as improved
67 building envelopes, more efficient equipment, and electrification of end uses—whereas
68 sufficiency avoids unnecessary energy demand while ensuring wellbeing^{6–8}. Previous research
69 indicates that sufficiency-oriented strategies, particularly those affecting floorspace and housing
70 type, can reduce building-sector energy demand and associated emissions by roughly 20–45%
71 by mid-century in high-income contexts (including the U.S.), in some cases approaching the
72 magnitude of reductions achievable through efficiency measures⁶. Beyond structural sufficiency,
73 residential energy demand is also shaped by indoor temperature setpoints, underscoring the
74 importance of evaluating sufficiency under alternative behavioral assumptions.

75 Despite its mitigation potential, sufficiency remains inconsistently represented or omitted
76 in national decarbonization scenarios, limiting the availability of evidence needed to inform policy
77 design^{9–11}. This underrepresentation reflects, in part, limitations in modeling frameworks, which
78 often rely on aggregated or archetype-based representations that limit their ability to capture
79 heterogeneity in housing structure and household practices relevant to sufficiency. Most
80 integrated assessment models (IAMs) capture only aggregated residential energy use—typically
81 at annual time scales and without explicit representation of building-level heterogeneity—
82 obscuring variation across buildings and households^{12,13}. These models also typically rely on
83 degree-day methods that approximate space-conditioning demand using ambient temperature
84 alone, neglecting important weather factors such as humidity, wind speed, and solar irradiance¹⁴.
85 Beyond IAMs, much of the building efficiency literature assumes a static building stock,

86 overlooking construction, renovation, and retirement processes that shape demand over time^{15–17}
87 (Supplementary Table S1).

88 Here we evaluate how structural and behavioral sufficiency, alongside energy efficiency
89 and electrification (via heat pump adoption), shape future U.S. residential energy demand and
90 emissions, both individually and in combination. To address limitations of existing modeling
91 approaches, we develop an integrated framework that links a dynamic building stock–turnover
92 model (MESSAGEix-Buildings)¹⁸ with a climate-responsive reduced-form building energy model
93 (RC-BEM)¹⁵. The integrated framework captures heterogeneity in housing stock evolution—
94 including building types and urban form—while simulating hourly thermal demand under future
95 climate trajectories. We conduct simulations for 2020–2050 under Representative Concentration
96 Pathways (RCPs) 4.5 and 8.5 and quantify changes in residential energy demand and
97 greenhouse gas (GHG) emissions under alternative power-sector decarbonization pathways.

98 This framework enables comparative assessment of energy efficiency, structural and
99 behavioral sufficiency, showing that policy effectiveness depends on housing stock evolution and
100 indoor temperature setpoints, and that sufficiency is essential alongside efficiency for deep
101 emissions reductions. It supports evaluation and sequencing of residential policy portfolios to
102 meet climate policy goals.

103 104 **Policy Scenario Framework**

105
106 Our scenario framework distinguishes two demand-side levers (Fig. 1). Structural sufficiency and
107 energy efficiency collectively shape “what we build,” while setpoint-based behavioral sufficiency
108 governs “how we use buildings.” This framing aligns with the IPCC Assessment Report, which
109 emphasizes infrastructure, technology, and sociocultural practices as key drivers of low-energy
110 demand futures^{19,20}.

111 On the “what we build” dimension, structural sufficiency is represented through changes
112 in per-capita floorspace and housing form, reflected in shifts in the relative shares of single-family
113 and multi-family dwellings. Energy efficiency is represented through assumptions about
114 renovation rates and depth, as well as heating electrification via heat pump adoption.

115 The **Reference** scenario reflects business-as-usual conditions based on recent U.S.
116 housing trends, characterized by per-capita floorspace growth of 10% per decade², constant
117 housing type shares (Supplementary Tables S2–S4), and continuation of recent trajectories in
118 renovation activity and heat pump adoption. The **Efficiency** scenario accelerates renovation
119 rates (from 1.5% to 2% per year), assumes greater renovation depth represented by larger
120 efficiency gains, and substantially increases heat pump adoption, while holding floorspace growth
121 and housing form constant. These assumptions represent plausible improvements in retrofit
122 activity and electrification relative to recent U.S. trends, rather than maximum feasible rates. The
123 **Sufficiency** scenario stabilizes per-capita floorspace at 2020 levels and modestly shifts the
124 housing stock toward multi-family dwellings, while maintaining Reference efficiency assumptions.
125 This scenario contrasts with recent U.S. trends of increasing floorspace per person and illustrates
126 a structural sufficiency pathway rather than an upper bound on feasible change. The **Combined**
127 scenario integrates the strategies from Efficiency and Sufficiency scenarios. All scenarios use
128 identical state-level population projections from Shared Socioeconomic Pathway 2 (SSP2;
129 “Middle of the Road”; Supplementary Figs. S1–S2), ensuring that differences across scenarios
130 reflect only assumptions about housing stock evolution and energy service provision.

131 On the “how we use” dimension, scenarios represent contrasting heating and cooling
132 practices through different assumptions about thermostat setpoints, which define different levels
133 of space-conditioning energy service. The **Baseline Comfort** scenario samples empirically
134 observed, state-specific distributions of heating and cooling thermostat setpoints, reflecting how
135 U.S. households currently condition their homes². The **Sufficient Comfort** scenario applies
136 health-informed indoor temperature setpoints based on World Health Organization guidance^{21,22}

137 and a 2022 California heat protection law²³, representing a lower-bound standard that maintains
138 wellbeing while reducing energy demand.

139 We evaluate each combination of “what we build” and “how we use” scenarios and
140 assess the sensitivity of outcomes to alternative climate and power-sector pathways. Climate
141 pathways are represented using **RCP4.5**, a moderate-warming trajectory consistent with
142 approximately 2.5–3 °C of global warming by 2100, and **RCP8.5**, a high-emissions trajectory
143 projecting approximately 4–5 °C of warming under limited mitigation^{24–26}. Power-sector pathways
144 are represented using three electricity decarbonization cases from the 2024 Cambium dataset²⁷
145 produced by the National Renewable Energy Laboratory (now the National Laboratory of the
146 Rockies): a **Mid-case** pathway with central cost assumptions, a **Fossil-leaning** pathway with
147 higher renewable costs and lower natural gas prices, and a **Renewable-leaning** pathway with
148 lower renewable costs and higher natural gas prices.

149

150

151 **Results**

152 **1. Residential energy demand can be decoupled from floorspace growth**

153 Under the Reference scenario, floorspace expands by 55%, from 27,200 Mm² in 2020 to 42,100
154 Mm² in 2050 (Fig. 2A). This expansion is accompanied by an 11% increase in final residential
155 energy demand, from 4.9 EJ to 5.5 EJ by mid-century (Fig. 2B). Nearly all of this increase is driven
156 by space heating, which grows from 4.0 EJ to 4.6 EJ, while cooling demand changes little because
157 efficiency gains offset higher cooling loads from larger dwellings and a warmer climate.

158 In the Efficiency scenario, floorspace growth mirrors the Reference case; however, final
159 energy demand declines to approximately 4.5 EJ by 2050 (–18% relative to the 2050 Reference
160 value). This reduction is concentrated in space heating: deeper renovations and higher renovation
161 rates reduce useful thermal demand, while heat pump adoption delivers the same heating service
162 with substantially lower final energy requirements than combustion-based systems.

163 In contrast, the structural sufficiency scenario limits total floorspace growth to 17% by 2050,
164 reducing final energy demand to approximately 3.8 EJ (–31%), exceeding reductions from
165 efficiency alone. When efficiency improvements are combined with structural sufficiency, final
166 energy demand declines further to approximately 3.0 EJ (–45%) by mid-century, representing the
167 largest absolute reduction across all scenarios.

168 Across all scenarios, differences between the RCP4.5 and RCP8.5 climate trajectories
169 result in less than a 2% change in total final energy demand in 2050, indicating that mid-century
170 demand outcomes are robust to climate uncertainty (Supplementary Fig. S3).

171 **2. Residential peak loads decline under warming and housing transformations**

172 Resolving hourly residential energy demand reveals how reductions in total energy use affect peak
173 demand and daily load profiles that are not visible in annual metrics alone (Fig. 3 shows
174 representative winter and summer days). This temporal detail is relevant for understanding
175 interactions with power systems, including renewable integration and peak capacity requirements,
176 although these dynamics are not explicitly modeled here.

177 In January, residential load exhibits a pronounced diurnal structure with morning and
178 evening heating peaks. In the 2050 Reference scenario, peak hourly load is highest in the Midwest,
179 reaching 224 GW, compared with overnight minima of 110 GW. In the West, lower loads relative
180 to 2020 primarily reflect reduced heating demand under warmer 2050 climate conditions in the
181 weather inputs. Under the Combined scenario, peak winter load declines substantially relative to
182 the 2050 Reference case across regions, by roughly 35–40%, with reductions approaching ~80–
183 90 GW in colder regions, while the timing of daily peaks remains largely unchanged.

184 In July, residential load shifts to afternoon cooling peaks and is much lower in absolute
185 magnitude than in winter, with peak hourly loads reaching ~45 GW in the South and West,
186 compared with winter peaks exceeding 200 GW in cold-climate regions. Under the Combined

187 scenario, peak summer load reductions are more modest at approximately 25%, corresponding to
188 absolute reductions of ~4–11 GW depending on region, with the largest reductions occurring in the
189 South. In this region, improvements in building efficiency and cooling technologies offset warmer
190 climate conditions, resulting in lower cooling loads despite rising temperatures. The timing of
191 afternoon cooling peaks remains unchanged.

192 Together, these results demonstrate a strong seasonal asymmetry. Winter loads in 2050
193 are generally lower than in 2020 due to warmer climate conditions reducing heating demand,
194 whereas summer loads are similar or slightly higher as cooling demand increases. Further
195 reductions in the Combined scenario arise from housing and technology changes that lower overall
196 demand levels without substantially altering the diurnal structure of residential load.

197 **3. Behavioral sufficiency yields substantial additional reductions in energy demand**

198 Thermostat setpoints directly determine the level of space-heating and cooling services delivered
199 and therefore represent a key behavioral driver of residential energy demand. According to 2020
200 Residential Energy Consumption Survey (RECS) data, prevailing thermostat practices in the U.S.
201 result in energy use well beyond levels required to meet health-based comfort guidance. The
202 national median setpoints are 21.1°C and 22.2°C for heating and cooling, respectively (Fig. 4A);
203 fewer than 10% of homes use heating setpoints below 18 °C or cooling setpoints above 26 °C.
204 These distributions define the Baseline Comfort assumption in our model.

205 By 2050, adopting Sufficient Comfort, which assumes household heating and cooling
206 setpoints aligned with health-based temperature thresholds, reduces final residential energy
207 demand by 31 – 33% (1.6 – 1.8 EJ) relative to Baseline Comfort across housing stock scenarios
208 (Fig. 4B). The magnitude of this reduction is comparable to that achieved through structural
209 sufficiency alone under Baseline Comfort.

210 **4. Housing improvements narrow geographic variation in sufficient-comfort savings**

211 We disaggregate state-level energy savings from adopting Sufficient Comfort setpoints to assess
212 geographic variation and distributional impacts (Fig. 5). Across all U.S. states, adopting Sufficient
213 Comfort reduces final energy demand, with absolute savings dominated by population size. Under
214 the Reference scenario, the median state reduces energy use by 33 PJ/year (IQR: 16–66), with
215 the largest contributions from highly populous states such as California and Texas (Fig. 5A). In the
216 Combined scenario, absolute savings fall to a median of 21 PJ/year (IQR: 11–44) due to lower
217 baseline demand, while the interquartile range narrows relative to the Reference case.

218 Expressed on a per-capita basis, energy savings from Sufficient Comfort are largest in
219 colder regions of the U.S., particularly across the Midwest and northern states (Fig. 5B). Under the
220 Reference scenario, per-person savings reach up to 6.1 GJ/person-yr, with a national median of
221 4.6 GJ/person-yr (IQR: 4.0–5.2). These patterns reflect the dominance of space-heating demand
222 in colder climates, where heating loads are both larger and more energy-intensive than cooling,
223 making energy use more responsive to heating setpoints. In the Combined scenario, both the
224 magnitude and the spread of per-capita savings are reduced, lowering the national median to 2.9
225 GJ/person-yr (2.7–3.2) and substantially narrowing geographic variation across states.

226 **5. Deep residential decarbonization hinges on both housing and grid transformations**

227 National primary energy demand in 2050 exhibits strong seasonal variation, with January
228 representing the maximum month across all housing stock scenarios (Fig. 6A). Structural
229 sufficiency and efficiency reduce demand across all months, with the largest absolute reductions
230 occurring in winter. Under Baseline Comfort, structural sufficiency lowers the January peak by 24%
231 relative to the Reference case, while combining efficiency and structural sufficiency reduces the

232 peak by 32% (from 1,281 to 878 PJ). Adopting Sufficient Comfort further reduces the January peak
233 by 24% in both the Reference and Combined scenarios (to 974 and 667 PJ, respectively).

234 These reductions in primary energy demand correspond to substantial declines in
235 residential GHG emissions across housing stock and grid scenarios (Fig. 6B). Emissions fall rapidly
236 through 2030 as space-heating electrification expands and the electricity sector decarbonizes,
237 declining from 328 MtCO₂e in 2020 to 250 MtCO₂e (-24%) in the Reference scenario and 211
238 MtCO₂e (-36%) in the Combined scenario under Baseline Comfort.

239 After 2030, emissions trajectories increasingly reflect electricity-sector assumptions. By
240 2050, emissions range from 203–231 MtCO₂e in the Reference scenario to 108–133 MtCO₂e in
241 the Combined scenario, depending on grid pathway. Fossil-leaning grids yield persistently higher
242 emissions, while renewable-leaning pathways produce lower outcomes, introducing a 22–33
243 MtCO₂e (10–25%) spread within each housing stock scenario. Even the Combined scenario
244 approaches but does not fully achieve the deep decarbonization benchmark^{28,29}, suggesting that
245 additional mitigation measures, including faster electrification or carbon removal, may be required
246 to reach comparable targets. Shifting from Baseline to Sufficient Comfort further reduces 2050
247 emissions by 23–33%, bringing outcomes closer to the deep decarbonization benchmark but not
248 fully achieving it.

249 Discussion

250

251 This study presents an integrated U.S. assessment of residential decarbonization pathways,
252 jointly examining structural sufficiency and energy efficiency in the built environment (“what we
253 build”) alongside behavioral sufficiency in household operation (“how we use”). By linking a state-
254 level housing stock turnover model with a climate-responsive building energy model, we capture
255 geographic and climatic variation that national assessments often obscure while resolving
256 residential energy demand at hourly temporal scales. This framework links housing infrastructure
257 decisions to future residential energy demand and emissions, enabling evaluation of sufficiency-
258 based interventions within broader mitigation strategies.

259 Our results underscore the long-term consequences of today’s housing decisions for
260 avoiding infrastructural lock-in. Our Reference scenario mirrors historic patterns: since the 1970s,
261 average single-family home size has grown by over 20% while household size has declined,
262 resulting in roughly a 40% increase in floorspace per person^{2,30}. If these patterns persist, the
263 resulting structural lock-in could erode efficiency gains and make achieving climate policy goals,
264 including net-zero emissions by mid-century, more difficult. Even the most ambitious combination
265 modeled here approaches but does not fully achieve deep decarbonization benchmarks,
266 indicating that housing and demand-side transformations alone are insufficient and require
267 additional measures such as faster electrification and deeper electricity-sector decarbonization.

268 While our modeling framework advances understanding of the role of sufficiency in
269 building energy policy, it also has limitations. First, our model parameters for new construction,
270 renovation, and retirement are specified exogenously. This limits representation of household-
271 level life-cycle processes and market dynamics. Second, we restrict attention to space heating
272 and cooling. While these loads dominate climate-sensitive demand, omitting other end uses
273 understates the residential sector’s total footprint and constrains comparability with whole-sector
274 assessments. Third, household income distributions and affordability constraints are not
275 represented, precluding explicit analysis of equity, energy insecurity, or heterogeneous adoption
276 pathways. Fourth, occupant behavior is represented through fixed thermostat setpoint regimes
277 rather than dynamic responses to energy prices, technologies, or extreme events, meaning
278 behavioral conclusions should be interpreted as bounding cases rather than forecasts.

279 Future work can address these gaps by embedding endogenous household and market
280 decision-making into our integrated stock turnover and building energy modelling framework,
281 extending coverage to all household energy demand and incorporating socioeconomic
282 heterogeneity. Coupling behavioral models of adaptive comfort management with physical energy
283 demand would further capture real-world feedbacks. Together, these steps would enable a more
284 comprehensive, policy-relevant assessment of how residential transformation interacts with

285 equity, affordability, and resilience in the U.S. energy transition. These types of insights will be
286 important for designing responsive policies, including targeted incentive and investment programs
287 that reflect different thermal comfort needs of households (especially vulnerable groups) and
288 address systemic barriers to achieving energy sufficiency.

289 Translating modeled pathways into practice depends on local governance and planning
290 frameworks. In some states, building codes are set at the state level, which can limit municipal
291 ambition, as in 24 states where local governments are barred from exceeding state standards³¹.
292 Elsewhere, cities have greater autonomy but face zoning restrictions that constrain denser urban
293 form^{32,33}. Financial incentives for retrofits and electrification can help overcome these barriers,
294 although recent political efforts in the U.S. threaten long-term decarbonization^{34,35}. Because
295 housing form and dwelling size are shaped by planning and regulatory decisions, advancing
296 structural sufficiency requires reform through zoning, building codes, and housing policy,
297 including measures that enable greater density and smaller dwelling units in traditionally single-
298 family neighborhoods³⁶, supported by incentives for compact, multi-family development³⁷.

299 Ambitious housing transformation is essential for aligning U.S. climate change mitigation
300 goals with equitable access to thermal comfort. Upgrading building infrastructure can reduce
301 energy demand and flatten seasonal energy peaks, while easing strain on the grid and
302 addressing infrastructure barriers to electrification in underserved households. While there are
303 opportunities for sufficiency to reduce energy demand in aggregate, many U.S. households report
304 difficulty meeting their energy needs, and supportive policies for households may increase their
305 energy use.⁴⁰ As these challenges evolve, policymakers must prioritize housing that is safe,
306 efficient, and equitable, integrating sufficiency into codes, incentives, and planning frameworks to
307 enable deep decarbonization across communities.

308

309 **Materials and Methods**

310

311 **1. Stock Turnover Model**

312 We use the Stock TURnover Model (STURM), a module of the MESSAGEix-Buildings
313 framework, to project the evolution of the U.S. residential building stock. STURM tracks annual
314 new construction, renovation, and demolition, capturing changes in floorspace, vintage
315 composition, and energy efficiency at regional and national levels. Key parameters—new
316 construction, renovation, and retirement rates—are set exogenously, using parametric variations
317 chosen to span plausible but distinct trajectories (“what we build” in Fig. 1 for scenario-specific
318 assumptions).

319 Household size and urban–rural shares are initialized using the 2020 Residential Energy
320 Consumption Survey (RECS) and assumed to be constant for all future years. State-level
321 population projections follow the U.S. version for the Global Change Analysis Model (GCAM-
322 USA) under SSP2, which assumes moderate population growth, urbanization, and income with
323 no extreme challenges to mitigation or adaptation. Simulations are conducted at five-year
324 intervals from 2020 through 2050.

325

326 **2. Reduced-Complexity Building Energy Model**

327 We use the Reduced-Complexity Building Energy Model (RC-BEM) to estimate hourly
328 heating and cooling demand in U.S. single-family homes under different climates and comfort
329 conditions.¹⁵ A detailed description of the RC-BEM model structure and validation is provided in
330 Nawawi et al. RC-BEM is trained on EnergyPlus simulations generated for 54 representative U.S.
331 cities, spanning 16 climate zones and 24 states. To capture building heterogeneity, we use
332 ResStock, a building stock analysis platform developed by the U.S. National Renewable Energy
333 Laboratory (NREL), which synthesizes realistic building prototypes by sampling across
334 distributions of floor area, construction vintage, insulation levels, and HVAC system efficiencies.
335 This procedure yields a sampling pool of ~17,000 unique single-family homes, which forms the

336 basis for scaling results to the national level through integration with STURM building stock
337 projections.

338 EnergyPlus outputs are converted to heating and cooling demand by accounting for HVAC
339 performance (i.e., annual fuel utilization efficiency for furnaces and coefficient of performance for
340 heat pumps). RC-BEM is trained using simulations driven by Typical Meteorological Year (TMY3)
341 weather files. For the scenario analyses, however, the model is applied to Thermodynamic Global
342 Warming (TGW)⁴¹ climate datasets, which preserve local weather variability while incorporating
343 climate change signals from global models. Indoor temperature states are included as
344 independent predictors in the RC-BEM, and the EnergyPlus training dataset spans a wide range
345 of indoor temperature conditions, allowing the model to represent the alternative setpoints used in
346 the behavioral sufficiency scenarios. Multivariate regression with lasso regularization is applied to
347 prevent overfitting. Once trained, RC-BEM produces hourly demand profiles for each prototype
348 across years (2020–2050) and climate scenarios (RCP4.5, RCP8.5). These hourly profiles are
349 subsequently aggregated to annual totals and scaled to the national level using STURM-
350 projected floor areas.
351

352 **3. Proxies for Missing Efficiency Classes and Multi-family Homes**

353 RC-BEM represents residential energy performance using nine efficiency classes (s1–s9),
354 which correspond to building construction decades, from pre-1940 stock (s1) through 2010s
355 construction (s9). STURM stock projections include additional classes not directly simulated in
356 RC-BEM (such as new construction or renovation-derived classes). To close these gaps, we
357 applied parametric proxy rules derived from RECS median site energy intensity values by
358 construction decade. For example, s1 is represented by s2 adjusted to be 10.4% more energy
359 intensive, and s9 is represented by s8 adjusted to be 7.1% less intensive.² Renovation-derived
360 classes are proxied from their parent vintage, with efficiency improvements of 30% (standard) or
361 60% (low).

362 Multi-family homes, which are not directly simulated in RC-BEM, are represented using
363 single-family intensity proxies scaled to multi-family floorspace from STURM. When no direct pool
364 was available, we used energy intensity values from similar buildings in the same climate zone,
365 or, if still unavailable, from national averages. This ensures complete coverage of all residential
366 floorspace in STURM.

367 **4. Model Integration and Validation**

368 To integrate the stock turnover and building energy models, we combine STURM stock
369 projections with RC-BEM intensity estimates in a post-processing framework. For each scenario,
370 STURM provides floor area by region, climate zone, and building vintage, which is matched with
371 RC-BEM heating and cooling intensities (kWh/m²). Total demand is obtained by multiplying
372 intensity by floor area and then aggregating across vintages, zones, and regions.

373 Where prototype pools were incomplete, we applied a hierarchical fallback procedure: donor
374 intensities were drawn first from the same city, then from the climate zone, and finally from the
375 national average. Monthly demand was distributed using donor shares from the best-available
376 pool. The integrated energy outputs include regional and national totals, per-square-meter
377 intensities, and monthly demand profiles.

378 Base-year validation compares modeled housing stock and energy demand against observed
379 statistics. Modeled housing units (129.75 million) closely match 2020 U.S. Census housing
380 counts (~126.8 million occupied units). Simulated heating and cooling energy use is calibrated to
381 national residential end-use consumption reported by the U.S. Energy Information Administration
382 (Supplementary Table S5).

383 **5. Final Energy Demand**

384 We compute site energy demand for heating and cooling by processing the integrated
385 outputs into monthly and annual metrics. Model outputs of “useful” (delivered) energy from RC-
386 BEM are converted to final energy using end-use-specific efficiencies. For heating, we assume
387 90% efficiency for gas furnaces. For cooling, we apply a year-dependent coefficient of
388 performance (COP), reflecting federal standards and projected improvements (e.g., COP = 3.2 in
389 2020, rising to 4.4 in 2050).

390 To ensure consistency with historical statistics, modeled site demand in 2020 is calibrated to
391 the Residential Energy Consumption Survey (RECS 2020). National totals reported in RECS
392 Table CE3.1 are 4,026 trillion Btu (4.25 EJ) for space heating and 866 trillion Btu (0.91 EJ) for
393 space cooling (Supplementary Table S4). Note that we did not include Hawaii and Alaska in our
394 analyses. Calibration factors are computed as the ratio of observed to modeled demand by
395 Census region and end use and applied across all scenario years. This adjustment anchors the
396 model to empirical benchmarks while preserving relative scenario differences.

397 **6. Primary Energy Demand and Emissions**

398 We translate site energy demand into primary energy by applying region- and year-specific
399 primary energy factors (PEFs) for electricity, derived from NREL’s Cambium dataset. For 2020
400 (historical baseline), we use averages from the 2010s: Northeast 2.63, Midwest 2.92, South 2.62,
401 and West 2.37. For 2050, we use business-as-usual projections: Northeast 2.32, Midwest 2.32,
402 South 2.33, and West 2.17. Non-electric fuels (e.g., natural gas, propane) are assigned a PEF of
403 1, treating site energy as equal to primary energy.

404 Each site energy flow is multiplied by its PEF, yielding monthly primary energy demand in
405 exajoules (EJ). Aggregation across fuels and end uses produces regional and national totals. We
406 also compute peak-to-average ratios of monthly demand to characterize temporal demand
407 profiles.

408 We estimated national CO₂e emissions from the residential sector by combining projected
409 site energy demand with fuel- and grid-specific emission factors. Results are reported as the sum
410 of direct and indirect emissions.

411 Direct emissions arise from on-site combustion of fossil fuels (e.g., natural gas, fuel oil,
412 propane). State-level residential demand for these fuels was taken from the STURM–ResStock
413 framework for each scenario (Reference, Efficiency, Structural Sufficiency, and Combined) and
414 comfort assumption (Baseline vs. Sufficient). We applied EPA’s emission factors for CO₂, CH₄,
415 and N₂O (converted to CO₂e using 100-year global warming potentials) to compute fuel-specific
416 emissions, which were then aggregated nationally.

417 Indirect emissions reflect the carbon intensity of electricity supplied to households. For 2020,
418 we used historical average emission rates from EPA’s eGRID dataset. For 2025–2050, we
419 applied average emission rates from NREL’s Cambium dataset, which provides scenario-specific
420 projections under alternative renewable cost and natural gas price assumptions. Three grid
421 pathways were used: HighRECost_LowNGPrice, MidCase, and LowRECost_HighNGPrice.
422 Residential electricity demand by state, scenario, and behavior was multiplied by the
423 corresponding Cambium emission rates and aggregated nationally.

424

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431
432

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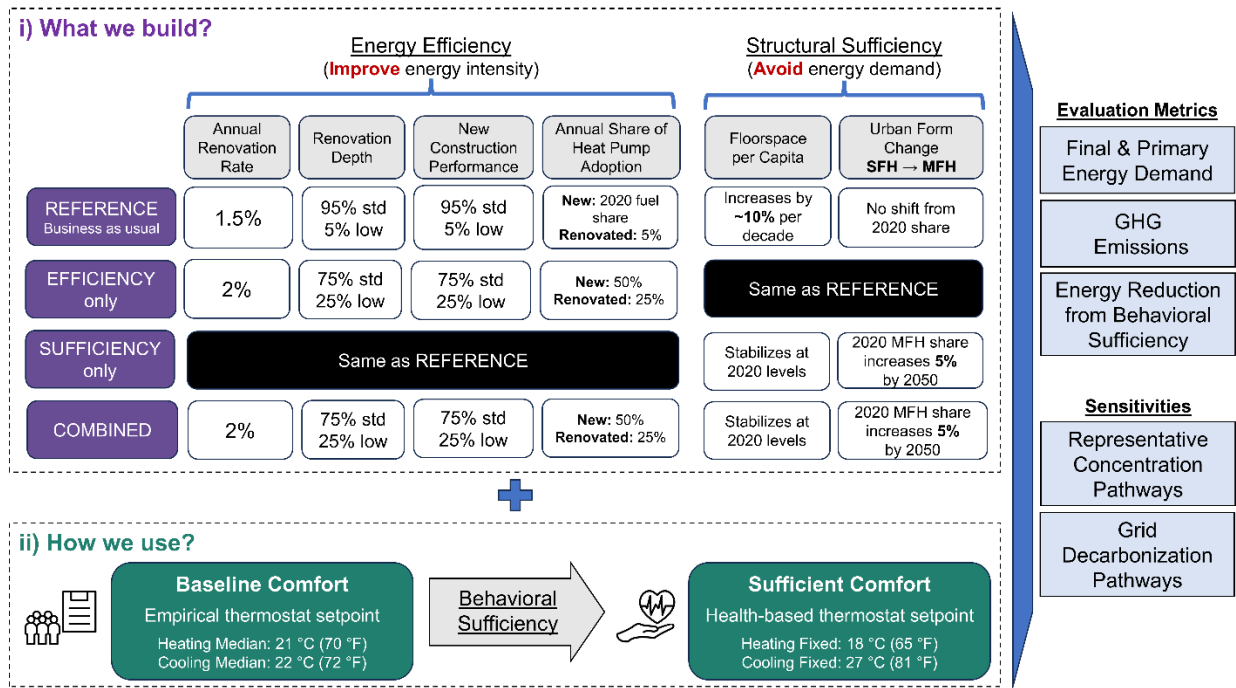
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Figures and Tables



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Figure 1. Scenario framework for the U.S. residential sector.

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Scenarios distinguish between what we build—represented by Reference, Efficiency, Sufficiency,

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and Combined housing stock pathways—and how we use, represented by thermostat setpoint

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assumptions spanning Baseline Comfort and Sufficient Comfort. Outcomes are evaluated for

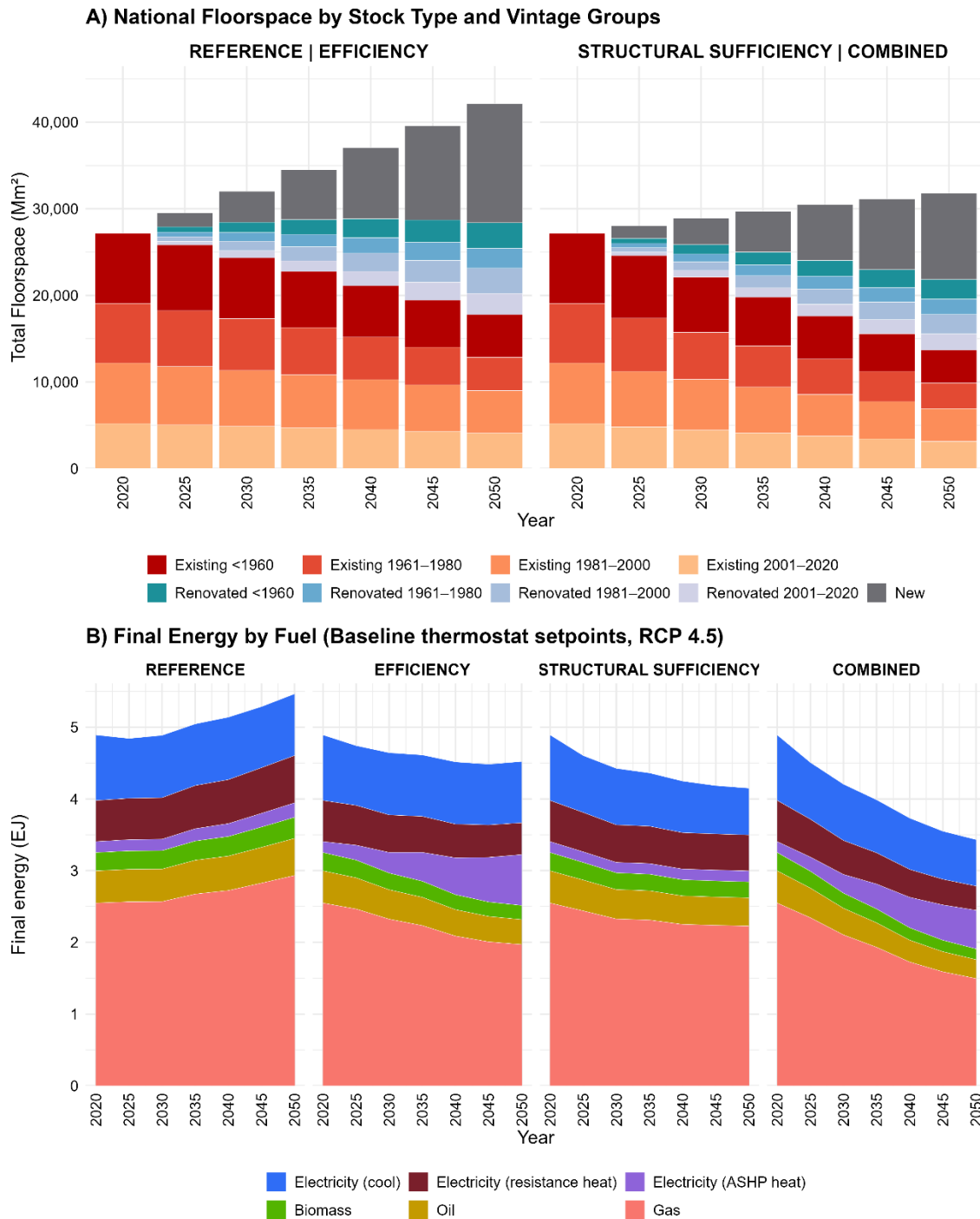
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residential energy demand and greenhouse gas (GHG) emissions, with sensitivities to

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Representative Concentration Pathways (RCPs) and electricity-sector decarbonization pathways.

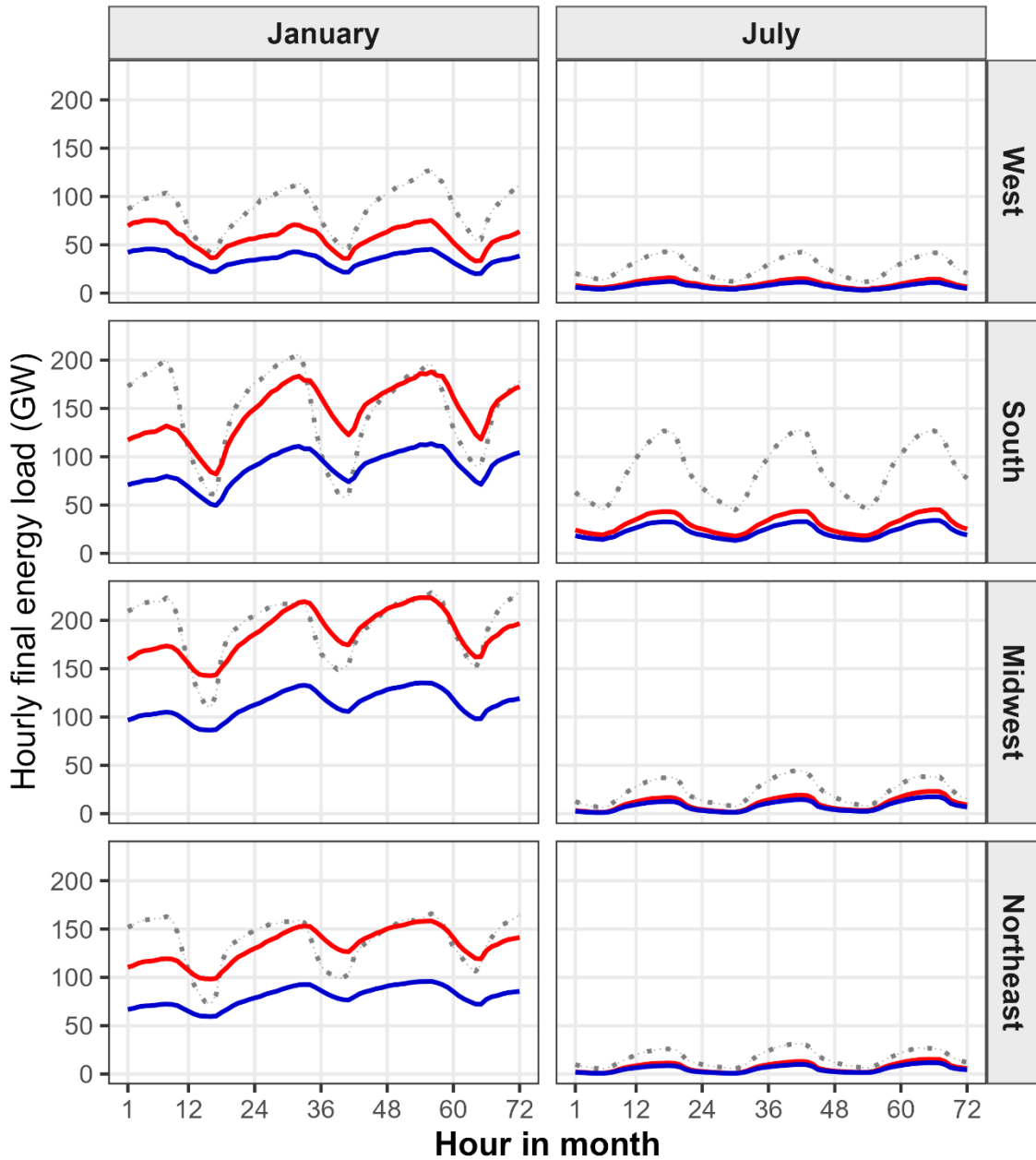
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Figure 2. U.S. residential floorspace and final energy across housing stock pathways. (A) National residential floorspace through 2050 by building cohort (existing, renovated, and new construction). (B) Final energy demand by fuel under the same pathways, assuming Baseline Comfort thermostat setpoints and RCP4.5.

First three days of each month



Scenario: ··· 2020 — 2050 (REFERENCE) — 2050 (COMBINED)

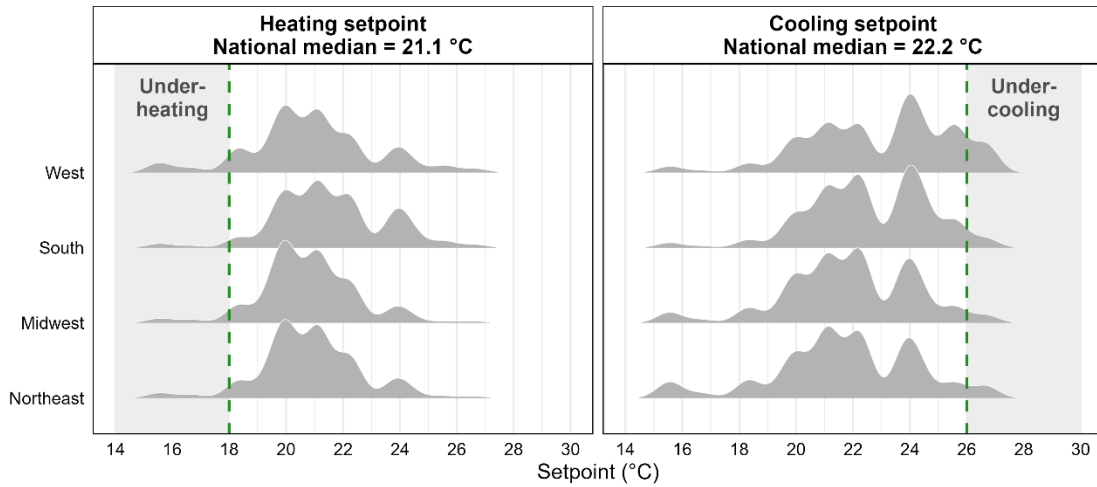
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544 **Figure 3. Hourly residential energy load by Census regions.**

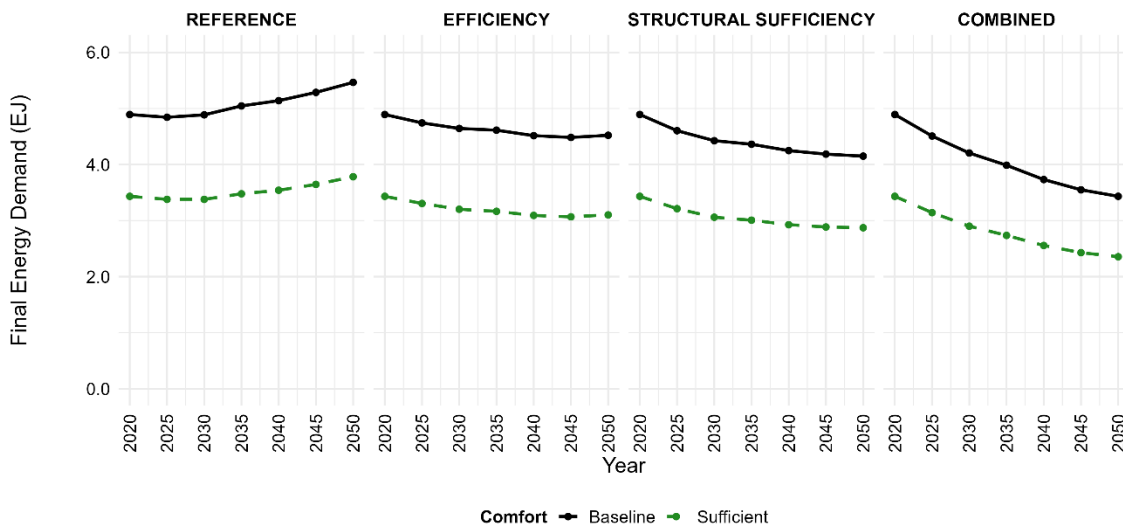
545 Hourly final energy load during the first three days of January and July, shown as illustrative
 546 winter and summer periods across U.S. Census regions, comparing 2020 baseline conditions
 547 with 2050 Reference and Combined scenarios. Results assume Baseline Comfort thermostat
 548 setpoints and RCP4.5 climate conditions.
 549

A) Baseline comfort: Distribution of heating and cooling setpoints by Census region

Note: Green dashed lines indicate health-based setpoints used to define Sufficient Comfort.



B) Final Energy Demand Trajectories (2020–2050, RCP4.5)



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Figure 4. Impact of thermostat setpoints on U.S. residential final energy demand.

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(A) Distribution of heating and cooling thermostat setpoints by U.S. Census region under Baseline

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Comfort; dashed lines indicate health-based thermostat setpoints used to define Sufficient

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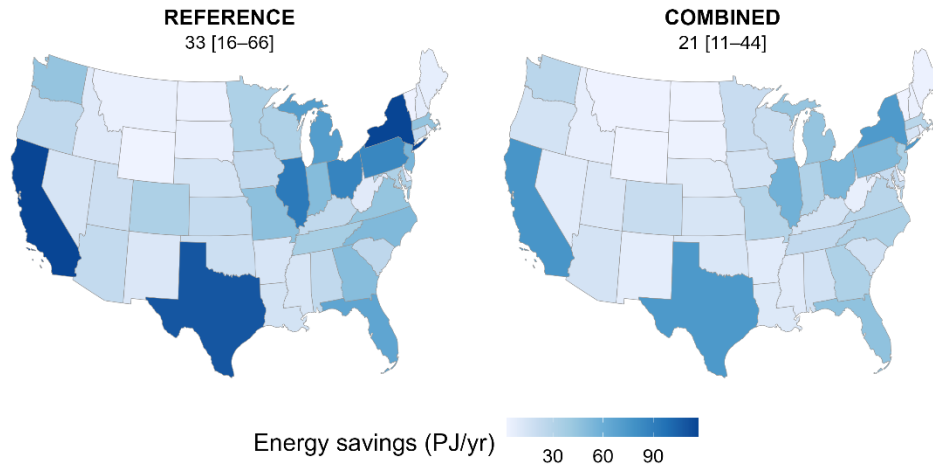
Comfort. (B) National final energy demand trajectories from 2020 to 2050 under Baseline Comfort

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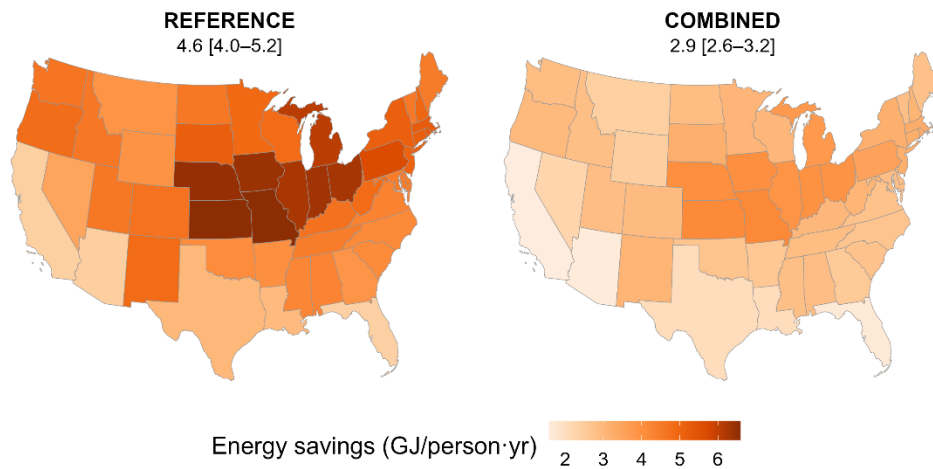
and Sufficient Comfort across housing stock pathways, assuming RCP4.5.

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A) Total energy savings from Baseline Comfort to Sufficient Comfort (2050)

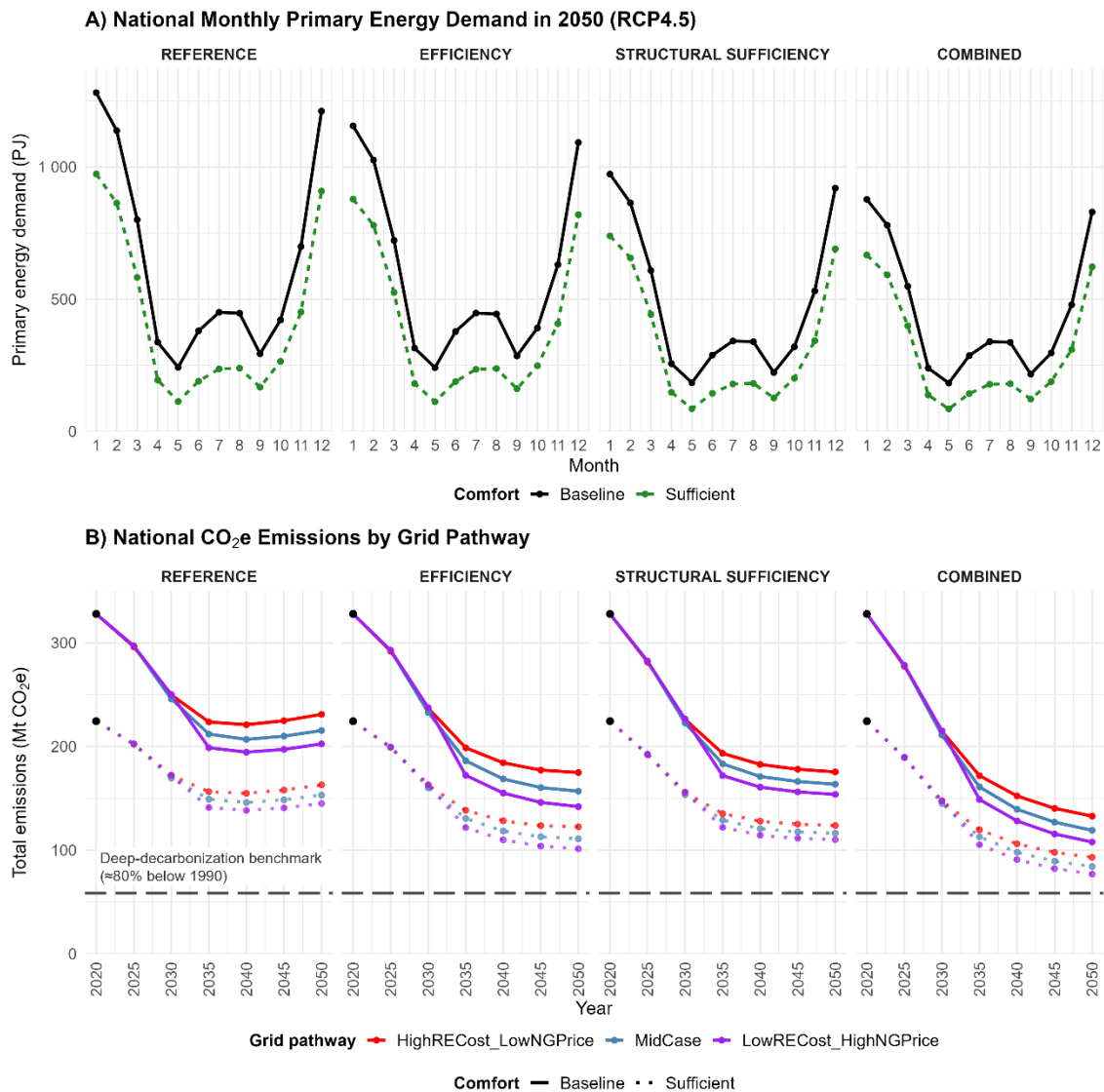


B) Per-capita energy savings from Baseline Comfort to Sufficient Comfort (2050)



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Figure 5. State-level final energy savings from Baseline to Sufficient Comfort in 2050. (A) Absolute annual residential final energy savings by state under the Reference and Combined housing stock pathways. (B) Corresponding per-capita energy savings. Color indicates savings magnitude relative to Baseline comfort; national values above each map indicate the median [IQR] across states.



567 **Figure 6. National primary energy demand and emissions**

568 (A) Monthly national primary energy demand in 2050 under Baseline and Sufficient Comfort
 569 across housing stock pathways, assuming RCP4.5. (B) Annual residential CO₂e emissions
 570 (Scope 1 + 2) under alternative housing stock pathways (facets) and electricity-sector
 571 decarbonization pathways (colors), comparing Baseline and Sufficient Comfort thermostat
 572 setpoints (line types), assuming RCP4.5. The dashed gray line indicates an illustrative deep-
 573 decarbonization benchmark corresponding to an ≈80% reduction in emissions relative to 1990
 574 levels, consistent with Deep Decarbonization Pathways Project targets.

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Supporting Information for

**Beyond efficiency: Sufficiency unlocks deep decarbonization of U.S.
residential sector**

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Figures

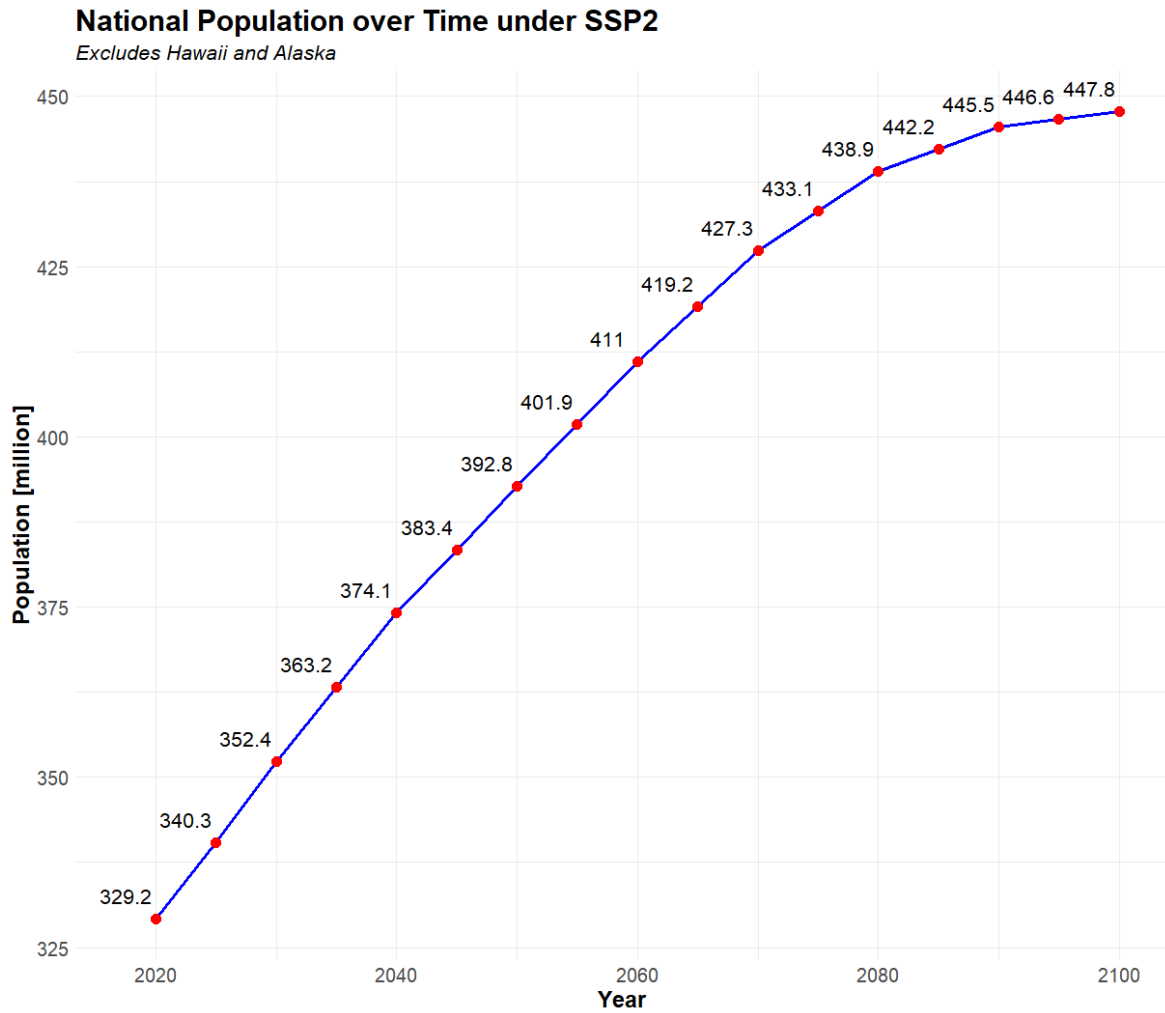


Fig. S1. National population trajectory under SSP2.

U.S. population projections from 2020 to 2100 under Shared Socioeconomic Pathway 2 (“Middle of the Road”), excluding Hawaii and Alaska. This population trajectory is applied uniformly across all policy scenarios. Source: GCAM-USA [1], [2]

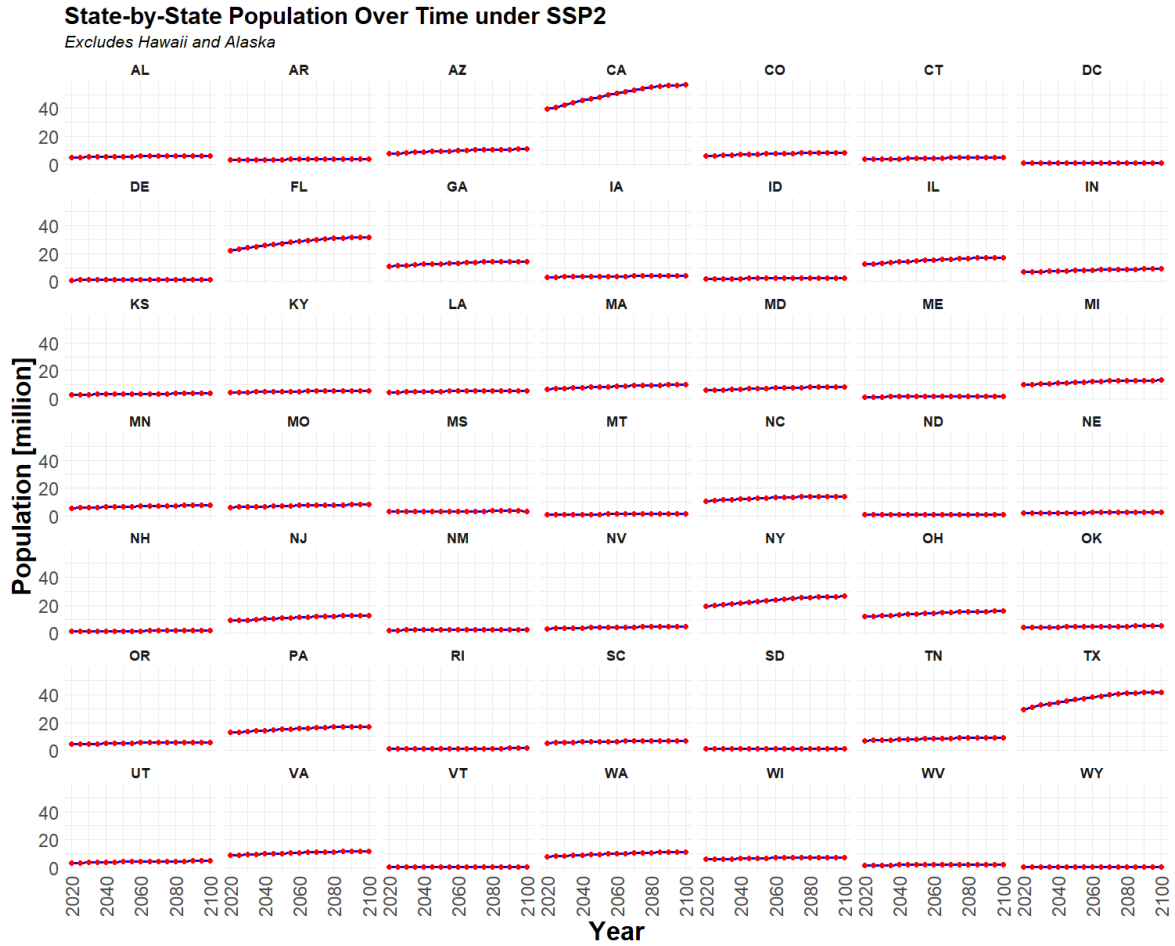


Fig. S2. State-level population trajectories under SSP2.

State-by-state population projections from 2020 to 2100 under Shared Socioeconomic Pathway 2, excluding Hawaii and Alaska. State-level population trajectories are held constant across all policy scenarios to isolate the effects of housing structure, energy efficiency, and comfort assumptions on residential energy demand and emissions. Source: GCAM-USA [1], [2]

National Final Energy Demand; Baseline Comfort

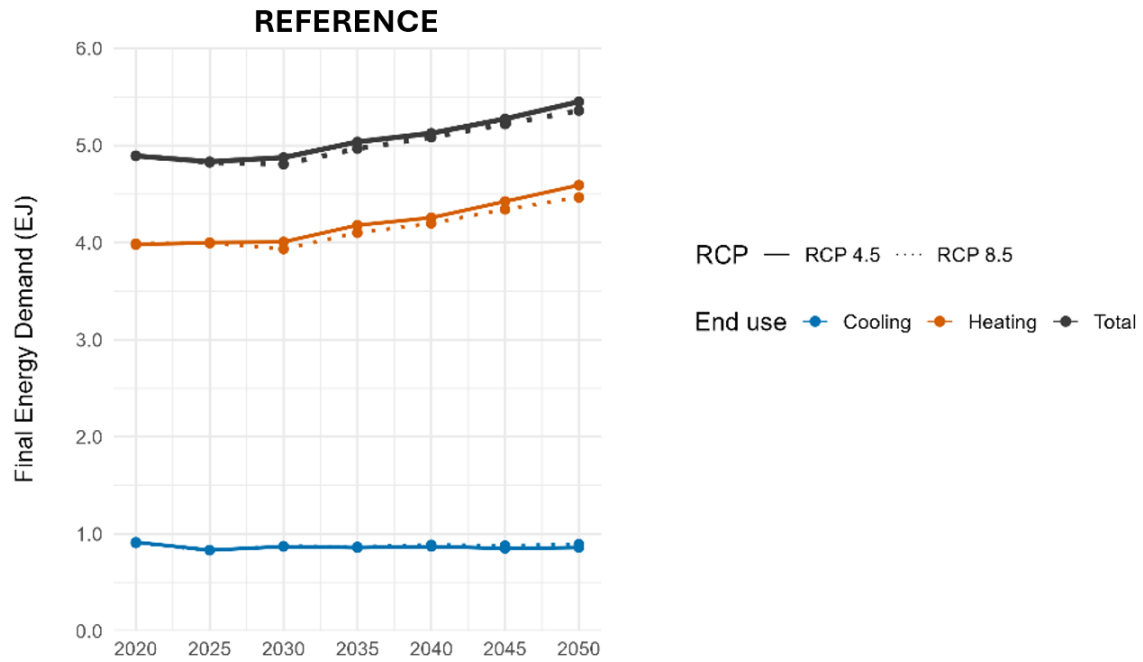


Fig. S3. Sensitivity of national final residential energy demand to climate pathways.

Final residential energy demand for heating, cooling, and total use under RCP4.5 and RCP8.5 using Baseline Comfort assumptions. The Reference scenario is shown as a representative case to illustrate that differences in climate pathways produce relatively small changes in mid-century energy demand compared with differences across policy scenarios.

Tables

Table S1. Comparison of key literatures focusing on U.S. nationwide residential decarbonization.

| | Energy Efficiency | Structural Sufficiency | Behavioral Sufficiency | Stock Turnover | Climate Change | Grid Decarbonization |
|---|--------------------------|-------------------------------|---|-----------------------|--|-----------------------------|
| <i>van Heerden et al (2025)</i> [3] | Yes | Yes | Yes Note: Thermostat adjustments are classified as efficiency improvements | Yes | No | Yes |
| <i>Zhang et al (2025)</i> [4] | Yes | No | No | No | Yes Note: Via Cooling/Heating Degree Days approach | Yes |
| <i>Wilson et al (2024)</i> [5] | Yes | No | No | No | No Note: All simulations use TMY3 data | Yes |
| <i>Berrill et al. (2022)</i> [6] | Yes | Yes | No | Yes | No Note: All simulations use TMY3 data | Yes |
| <i>This paper</i> | Yes | Yes | Yes | Yes | Yes Note: Consider hourly future weather conditions | Yes |

Table S2. Shares of rural residential building archetypes in 2020 by state. [7]

| STATE | MFH_LARGE | MFH_SMALL | SFH_ATTACHED | SFH_DETACHED |
|--------------|------------------|------------------|---------------------|---------------------|
| AL | 5.2% | 5.6% | 1.2% | 88% |
| AR | 5.8% | 6.1% | 1.2% | 87% |
| AZ | 5.0% | 4.2% | 2.6% | 88% |
| CA | 5.1% | 4.2% | 2.9% | 88% |
| CO | 6.8% | 3.9% | 5.3% | 84% |
| CT | 5.4% | 8.1% | 4.2% | 82% |
| DC | 0.0% | 0.0% | 0.0% | 0.0% |
| DE | 0.0% | 0.0% | 0.0% | 0.0% |
| FL | 7.4% | 5.6% | 3.9% | 83% |
| GA | 5.0% | 5.4% | 1.8% | 88% |
| IA | 6.0% | 3.6% | 2.4% | 88% |
| ID | 4.7% | 4.3% | 3.1% | 88% |
| IL | 4.0% | 4.1% | 1.7% | 90% |
| IN | 4.5% | 3.3% | 1.8% | 90% |
| KS | 4.8% | 4.0% | 3.0% | 88% |
| KY | 6.2% | 6.2% | 2.1% | 86% |
| LA | 4.7% | 6.0% | 1.6% | 88% |
| MA | 6.7% | 7.3% | 4.3% | 82% |
| MD | 6.9% | 3.7% | 5.7% | 84% |
| ME | 4.7% | 5.6% | 1.6% | 88% |
| MI | 4.9% | 2.8% | 2.2% | 90% |
| MN | 7.7% | 2.6% | 3.4% | 86% |
| MO | 3.6% | 5.3% | 2.1% | 89% |
| MS | 6.6% | 5.4% | 1.2% | 87% |
| MT | 5.5% | 4.9% | 4.6% | 85% |
| NC | 4.5% | 4.8% | 1.9% | 89% |
| ND | 14% | 4.2% | 6.3% | 75% |
| NE | 5.4% | 3.9% | 2.8% | 88% |
| NH | 8.6% | 6.8% | 5.1% | 79% |
| NJ | 0.0% | 0.0% | 0.0% | 0.0% |
| NM | 4.5% | 4.3% | 2.8% | 88% |
| NV | 5.9% | 5.3% | 3.7% | 85% |
| NY | 5.1% | 6.9% | 2.2% | 86% |
| OH | 4.4% | 4.4% | 2.4% | 89% |
| OK | 4.3% | 3.6% | 1.5% | 91% |
| OR | 5.6% | 5.1% | 3.2% | 86% |
| PA | 3.7% | 3.9% | 6.5% | 86% |
| RI | 0.0% | 0.0% | 0.0% | 0.0% |
| SC | 5.9% | 5.1% | 2.1% | 87% |
| SD | 9.2% | 3.4% | 3.7% | 84% |
| TN | 5.0% | 5.5% | 1.5% | 88% |

| | | | | |
|-----------|------|------|------|-----|
| TX | 4.9% | 4.9% | 1.6% | 89% |
| UT | 5.8% | 4.6% | 5.2% | 84% |
| VA | 5.7% | 3.0% | 3.6% | 88% |
| VT | 5.1% | 7.3% | 1.9% | 86% |
| WA | 6.3% | 3.9% | 2.6% | 87% |
| WI | 6.0% | 4.0% | 3.3% | 87% |
| WV | 5.1% | 3.2% | 2.8% | 89% |
| WY | 6.0% | 4.3% | 6.2% | 83% |

Table S3. Shares of suburban residential building archetypes in 2020 by state. [7]

| STATE | MFH_LARGE | MFH_SMALL | SFH_ATTACHED | SFH_DETACHED |
|-------|-----------|-----------|--------------|--------------|
| AL | 12% | 3.8% | 2.3% | 82% |
| AR | 12% | 5.8% | 2.9% | 79% |
| AZ | 11% | 3.5% | 5.9% | 79% |
| CA | 14% | 5.3% | 8.2% | 73% |
| CO | 13% | 3.1% | 8.3% | 76% |
| CT | 10% | 7.6% | 6.7% | 76% |
| DC | 0.0% | 0.0% | 0.0% | 0.0% |
| DE | 11% | 2.90% | 17% | 69% |
| FL | 20% | 5.5% | 7.8% | 67% |
| GA | 14% | 4.2% | 5.6% | 77% |
| IA | 13% | 3.3% | 8.3% | 76% |
| ID | 6.9% | 4.9% | 4.4% | 84% |
| IL | 11% | 4.7% | 9.0% | 75% |
| IN | 9.6% | 3.9% | 5.0% | 82% |
| KS | 12% | 3.70% | 7.50% | 77% |
| KY | 12% | 6.90% | 4.10% | 77% |
| LA | 9.7% | 5.7% | 2.6% | 82% |
| MA | 12% | 11% | 6.80% | 71% |
| MD | 17% | 1.9% | 20% | 61% |
| ME | 7.7% | 7.7% | 4.7% | 80% |
| MI | 9.7% | 3.1% | 6.3% | 81% |
| MN | 14% | 2.0% | 12% | 72% |
| MO | 8.8% | 4.3% | 5.0% | 82% |
| MS | 11% | 4.5% | 1.9% | 83% |
| MT | 9.3% | 7.3% | 8.1% | 75% |
| NC | 13% | 4.0% | 6.3% | 76% |
| ND | 32% | 4.0% | 11% | 53% |
| NE | 15% | 2.8% | 6.1% | 76% |
| NH | 10% | 5% | 8.40% | 76% |
| NJ | 12% | 8.3% | 11% | 68% |
| NM | 9.3% | 5.0% | 5.2% | 80% |
| NV | 18% | 6.5% | 5.5% | 70% |
| NY | 8.8% | 7.0% | 5.1% | 79% |
| OH | 10% | 4.7% | 6.0% | 79% |
| OK | 12% | 3.8% | 2.8% | 82% |
| OR | 15% | 5.0% | 6.2% | 74% |
| PA | 7.3% | 4.1% | 16% | 72% |
| RI | 10% | 10% | 4.4% | 75% |
| SC | 11% | 4.4% | 4.2% | 80% |
| SD | 21% | 3.1% | 8.8% | 67% |

| | | | | |
|-----------|-------|------|------|------|
| TN | 13% | 4.9% | 4.7% | 77% |
| TX | 18% | 4.1% | 3.1% | 75% |
| UT | 10% | 4.2% | 7.6% | 78% |
| VA | 14% | 2.8% | 17% | 67% |
| VT | 12% | 8.2% | 9.8% | 70% |
| WA | 14% | 4.0% | 4.9% | 77% |
| WI | 14% | 5.9% | 6.1% | 74% |
| WV | 8.70% | 4.9% | 3.1% | 83% |
| WY | 0.0% | 0.0% | 0.0% | 0.0% |

Table S4. Shares of urban residential building archetypes in 2020 by state. [7]

| STATE | MFH_LARGE | MFH_SMALL | SFH_ATTACHED | SFH_DETACHED |
|-------|-----------|-----------|--------------|--------------|
| AL | 18% | 8.1% | 2.5% | 72% |
| AR | 0.0% | 0.0% | 0.0% | 0.0% |
| AZ | 23% | 5.8% | 5.9% | 65% |
| CA | 33% | 8.1% | 9.9% | 49% |
| CO | 23% | 4.0% | 12% | 61% |
| CT | 28% | 30% | 7.6% | 34% |
| DC | 43% | 7.8% | 33% | 16% |
| DE | 0.0% | 0.0% | 0.0% | 0.0% |
| FL | 30% | 7.0% | 6.9% | 56% |
| GA | 44% | 5.3% | 6.9% | 44% |
| IA | 17% | 2.6% | 4.4% | 75% |
| ID | 0.0% | 0.0% | 0.0% | 0.0% |
| IL | 33% | 23% | 5.4% | 39% |
| IN | 22% | 7.0% | 11% | 60% |
| KS | 11% | 5.9% | 4.1% | 79% |
| KY | 19% | 7.0% | 6.6% | 67% |
| LA | 16% | 15% | 8.9% | 60% |
| MA | 34% | 27% | 9.3% | 30% |
| MD | 18% | 7.0% | 57% | 17% |
| ME | 0.0% | 0.0% | 0.0% | 0.0% |
| MI | 12% | 4.2% | 6.7% | 77% |
| MN | 31% | 7.6% | 4.4% | 57% |
| MO | 21% | 14% | 4.0% | 61% |
| MS | 17% | 7.3% | 2.9% | 73% |
| MT | 0.0% | 0.0% | 0.0% | 0.0% |
| NC | 34% | 7.0% | 7.1% | 52% |
| ND | 0.0% | 0.0% | 0.0% | 0.0% |
| NE | 17% | 3.9% | 5.6% | 74% |
| NH | 25% | 16% | 7.6% | 51% |
| NJ | 37% | 35% | 13% | 15% |
| NM | 13% | 6.4% | 4.8% | 76% |
| NV | 32% | 12% | 5.4% | 50% |
| NY | 45% | 21% | 13% | 22% |
| OH | 15% | 12% | 7.8% | 65% |
| OK | 25% | 6.1% | 4.8% | 64% |
| OR | 15% | 6.5% | 4.4% | 74% |
| PA | 14% | 8.7% | 63% | 15% |
| RI | 20% | 34% | 5.2% | 41% |
| SC | 0.0% | 0.0% | 0.0% | 0.0% |
| SD | 0.0% | 0.0% | 0.0% | 0.0% |

| | | | | |
|-----------|------|------|------|------|
| TN | 21% | 6.9% | 6.0% | 67% |
| TX | 27% | 6.0% | 4.2% | 63% |
| UT | 26% | 11% | 11% | 52% |
| VA | 27% | 6.3% | 14% | 52% |
| VT | 0.0% | 0.0% | 0.0% | 0.0% |
| WA | 32% | 5.0% | 6.4% | 57% |
| WI | 13% | 21% | 7.1% | 59% |
| WV | 0.0% | 0.0% | 0.0% | 0.0% |
| WY | 0.0% | 0.0% | 0.0% | 0.0% |

Table S5. 2020 site end-use consumption [trillion BTU] in the U.S. used for model calibration. [7]

| | SPACE HEATING | AIR CONDITIONING |
|-----------------------------------|----------------------|-------------------------|
| ALL HOMES | 4,026 | 866 |
| CENSUS REGION AND DIVISION | | |
| NORTHEAST | 1,030 | 93 |
| NEW ENGLAND | 293 | 19 |
| MIDDLE ATLANTIC | 737 | 74 |
| MIDWEST | 1,484 | 131 |
| EAST NORTH CENTRAL | 1,054 | 84 |
| WEST NORTH CENTRAL | 430 | 47 |
| SOUTH | 906 | 497 |
| SOUTH ATLANTIC | 459 | 248 |
| EAST SOUTH CENTRAL | 184 | 66 |
| WEST SOUTH CENTRAL | 263 | 183 |
| WEST | 605 | 145 |
| MOUNTAIN | 284 | 80 |
| MOUNTAIN NORTH | 208 | 18 |
| MOUNTAIN SOUTH | 75 | 62 |
| PACIFIC | 322 | 65 |

Note: Vacant housing units, seasonal units, second homes, military houses, and group quarters are excluded.

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