Geography, Policy, or Market? New Evidence on the Measurement and Causes of Sprawl (and Infill) in US Metropolitan Regions

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Abstract: This paper proposes four metrics to measure sprawl in metropolitan regions as marginal changes in land use over time. The metrics (change in urban housing unit density, marginal land consumption per new urban household, housing unit density in newly urbanized areas, and percent of new housing units located in previously developed areas) are computed for all 329 metropolitan areas in the continental United States for 1980 and 2000. Regression analysis is used to explain variations in sprawl metrics across metropolitan areas, incorporating variables representing market, geographic, and policy factors. Changes in development patterns reflect interactions of market and geographic structures. States with a substantial state role in planning accommodate a higher percentage of new housing units in previously developed areas and with lower marginal land consumption, suggesting that policy can mitigate sprawl development.
1. Introduction.

Measuring and understanding the nature of urban spatial expansion is the subject of intense academic and popular interest. The literature on “sprawl” has evolved to consider a wide range of measures, descriptions and causes and has utilized different data sources and methods, but has not achieved anything like scientific consensus (Cutsinger and Galster 2006). While this paper does not provide definitive answers to this large and complex literature, it proposes a framework to think about measurement(s) of urban sprawl, offers multiple new data sources to implement these measures, and presents multivariate estimates of the impact of different geographic, policy and market forces in explaining marginal changes in metropolitan land use patterns in the US over 20 years.

Because sprawl is often conceptualized as a multidimensional phenomenon, measurement faces at least two interrelated challenges. First, to achieve analytical tractability, researchers either reduce the dimensionality of measurement through creation of indices, or focus on one dominant feature of land use patterns. The tradeoff in empirical sprawl metrics is that multidimensional indices may conflate process, outcomes and causes, while single-dimension measurements may obscure these important interrelationships. Tradeoffs between conceptualization and measurement may be driven by theory or by data availability. Reducing the dimensionality of a complex phenomenon does not have to be considered “reductionistic” as long as the research clearly describes these tradeoffs and the measurements retain validity.

The second challenge, at least in the US, has been the difficulty in obtaining temporally and spatially consistent land use data, able to be matched with socioeconomic, demographic and policy variables. If measures of sprawl reflect changes in land use patterns within an area, then empirical work requires data sources comparable across time and space and at the scale of the phenomenon being explained. This paper attempts to address these two challenges, both in the conceptual
development of sprawl metrics and with the deployment of new data sources to measure and explain land use changes across US metropolitan areas.

This paper is organized as follows. Section 2 outlines the literature on urban sprawl measurement in the US. This discussion argues for conceptual measurements of sprawl focused on relative and marginal changes in density, and four specific metrics of sprawl are proposed to reflect these concepts. Section 3 presents a discussion of the data sources and methods utilized to implement the metrics and the regression analysis of determinants of sprawl patterns. Section 4 presents the results and discussions testing the impact of various geographic, policy and market forces in explaining differences in sprawl measures across US metropolitan regions. Discussion and conclusions follow.

2. Concepts and measurement of “sprawl.”

The word “sprawl” – in its noun and verb forms – connotes a negative characterization of spatial patterns of land development. However, as a scientific and policy concept it remains contested and poorly defined (Galster, et al. 2001; Ewing, et al. 2002; Cutsinger, et al. 2005; Wolman, et al. 2005). To review the voluminous sprawl literature is beyond the scope of this paper, but this review highlights five components of the measurement debate impacting the development of the proposed metrics. Questions of methodology include whether sprawl should be measured as a multi-dimensional phenomenon or whether it is valid to focus on one element. Second, can sprawl be measured in a cross-sectional framework as variation across places, or should it be measured in a time-series framework as changes within places? Third, is the appropriate measure of the concept of sprawl “average” or “marginal”? Fourth, to what extent should measures of sprawl be integrated into stylized models of urban spatial structure? And, fifth, what spatial scale is appropriate for measurement?
Researchers presenting multi-dimensional indices of sprawl argue that sprawl is a complex phenomenon of both pattern and process such that understanding the interactions of pattern and process are necessary for policy relevance (Galster, et al. 2001; Ewing, et al. 2002; Cutsinger, et al. 2005; Cutsinger and Galster 2006). In the two extant research projects which advance multi-dimensional measures of sprawl (Ewing, et al. 2002; Cutsinger, et al. 2005) a large number of variables for each metropolitan area are collected and data reduction techniques are used to compute statistically valid indices for comparison across places. Ewing et al. (2002) combine multiple data sources for 83 metropolitan areas in 1990 and 2000. Using 22 variables from Census, employment, transportation, and land use databases, they use principal components analysis to identify four factors representing density, neighborhood land-use mix, activity centering, and street-network accessibility. Each of these constructs is operationalized as that principal component which accounts for the maximum variance in the underlying variables. These four factors are scaled and combined into a “Sprawl Index,” to measure differences across metropolitan areas.

Cutsinger, et al. (2005) and Wolman, et al. (2005) update the data and methodology in Galster, et al. (2001) and cover a sample of 50 metropolitan areas for 1990. Combining Census, transportation, employment and land-cover data, they create 14 indices measuring sprawl along the dimensions of density, concentration, centrality, continuity, nuclearity (centers), and mix of uses. They then use principal components analysis to reduce these 14 indices to seven factors, which they label density/continuity, proximity, job distribution, mixed-use, housing centrality, nuclearity, and housing concentration. The seven factors rank and compare metropolitan areas along each dimension of sprawl. The authors argue that different ordinal rankings across metropolitan areas confirm the multidimensional nature of sprawl which would be obscured in a single index.

In contrast, a large group of researchers have measured sprawl primarily in terms of density (Brueckner 2001; Fulton, et al. 2001; Kahn 2001; Malpezzi and Guo 2001; Lopez and Hynes 2003;
Density can be population, housing-unit, or employment density, and measurements include average densities, variations in densities, or density gradients. Density measures focus on the dominant characteristic of sprawl and are more available in terms of data. Even within the literature measuring sprawl multidimensionally, variables representing density often play a strong role. In Ewing, et al’s (2002) index, the density factor accounts for 2/3rds of the variance within the data set. Of the 7 components identified by Cutsinger, et al. (2005) the density/continuity factor accounts for the largest share of variance explained.

A second question of methodology concerns whether sprawl can be measured cross-sectionally or requires measurement over time. Even when theoretical understanding focuses on land use change, data availability has often forced cross-sectional measurements. Cross-section measures pose a challenge in interpretation because places have different histories and policy choices may be endogenous to observed land use differences. For example, density in New York is sometimes compared to Los Angeles, even though these two cities developed at different times in American industrial and settlement history. In Los Angeles, only 25 percent of the housing stock was built before 1950, compared to 50 percent in New York. If comparisons of land use patterns do not reflect historical differences, latter developing cities by definition would be more “sprawled” because they lack historically dense cores.

When sprawl is measured as a process in a time-series framework, metrics should characterize changes in land use patterns within a metropolitan area. Changes across places can still be compared, but only in relative terms. Consider again the New York/Los Angeles example. New York’s housing-unit density (per km² of urban land area, year 2000) is more than twice that of Los Angeles’ (1852 units/km² vs. 897 units/km²). However, in the 20 years from 1980 to 2000, density in the Los Angeles area increased 48 dwelling units/km², while density in the New York metro
declined 13 dwelling units/km². While New York is overall denser, new urban development in previously non-urbanized areas in the New York area occurs at lower densities than new urban development in Los Angeles.

Figure 1 illustrates this concept, showing a nonparametric regression of the natural log of population density on the distance to the metropolitan center for urbanized block-groups in the New York and Los Angeles PMSAs in 2000. Regressing the log of population density on distance is a standard implementation of the “density gradient” but nonparametric regression techniques do not impose any particular functional form on the output. New York fits the model of a traditional “center-dominated” city with a historically dense urban core and lower-density suburban areas. Los Angeles lacks a traditional “center” but is developed at more uniform and higher densities across the region. The suburban areas of Los Angeles beyond 13 miles (21 km) are developed at higher densities than similarly distant areas in New York.

A third area of methodological concern is whether sprawl should be measured in average or marginal terms. In most discussions of sprawl, a common meaning is low density development in suburbs or at the urban-rural interface. However, much research on sprawl measures the density across a metropolitan area, (an average measurement) rather than the (marginal) density of newly developed land areas, or marginal land consumption per household. Marginal land consumption measures reflect a conceptualization of sprawl as a process.

The fourth area of methodological difference is the extent to which researchers understand measurements to be rooted in models of urban spatial structure. Consider the standard model of spatial structure in urban economics, the monocentric or Alonso-Mills-Muth (AMM) model.
Development densities within a region reflect land prices, as different uses compete for accessibility to the central business district. The emergent pattern of land use is the envelope of each use’s bid-rent curve. Land values, and hence development densities, decline with distance to the center. The spatial extent of urban development is increasing in population and income, and decreasing in agricultural land prices and transportation costs (Wheaton 1974; Brueckner and Fansler 1983; McGrath 2005; Spivey 2008; Paulsen 2012).

Despite its overly simplified model structure, the monocentric city model continues to explain a large percentage of variation in development patterns across US areas (Malpezzi and Guo 2001; Wassmer 2006; Spivey 2008; Paulsen 2012). In the context of this stylized model, increases in population and income are expected to lead to outward urban expansion. Should we therefore understand all spatial expansion of the development area of a city as “sprawl?” Urban economist Jan Brueckner argues from this model that we should only label “excessive” spatial expansion of urbanized areas as “sprawl” (Brueckner 2001). Urban spatial expansion is “excessive” when one or more policy or market failures extends land development further out than would otherwise be socially optimal. There is yet no agreement how to measure the welfare losses of habitat loss, traffic congestion or other impacts of sprawl, so there is no agreed-upon method to identify the “optimal” size of urbanized areas (Cheshire and Sheppard 2004). But the intuition behind this argument is that not all expansions of urban areas constitute “sprawl.” When sprawl is measured as marginal changes in land patterns and densities across places, comparisons could show that one place is more “sprawling” than another, while not directly testing or measuring whether such spatial expansion is “excessive.”

The fifth concern of methodology considers the spatial scale at which sprawl is measured. Most of the research cited above uses metropolitan statistical areas. Even when land-use data is derived from non-Census sources, the metropolitan area is used to represent the economic
integration of rural areas with one or more central (principal) cities. Metropolitan areas are based on county boundaries, so a county with some economic integration (commuting) with the central city would be included even if it contains significant amounts of rural and/or undeveloped land. Metropolitan area boundaries are often used because of the wider range of data available and because these boundaries include “rural” or “exurban” sprawl: (very) low-density development at the fringe without urban-type services (water and sewer connections).

Alternatively, a number of researchers within the urban economics tradition use “urbanized area” boundaries of the Census (Spivey 2008; Wassmer 2006, 2008). Their measures focus on sprawl as “urban decentralization” (Wassmer 2008, p.537) but exclude (very) low-density development outside of “urban areas” by definition. Wolman et al. (2005) and Cutsinger et al. (2005) argue that metropolitan area definitions “overbound” sprawl because including rural areas upwardly biases sprawl indicators. Likewise, they argue that urbanized area boundaries “underbound” sprawl because some amount of rural or fringe development is not included. Instead, they propose a measurement called the “extended urban area” (EUA) as the Census-defined “urbanized area” plus “each additional outlying square-mile cell ...[within the] metropolitan statistical area that has 60 or more dwelling units and from which at least 30% of its workers commute to the urbanized area.” (Cutsinger, et al. 2005, p.237).

The research presented here focuses on measuring “urban” sprawl (as incremental urban expansion) and would therefore necessarily exclude or “underbound” what some would call “rural sprawl.” The results of this research should therefore be understood in this context and not as a complete measure of all land use change and development. There is a large and growing literature on measuring and defining rural sprawl, which argues that rural housing development is driven by amenities more than proximity to employment (Deller et al. 2001; Esparza and Carruthers 2000; Radeloff et al 2005). “Urban” sprawl therefore has different drivers and policy responses than
“rural” sprawl. The purpose of this paper is not to diminish or ignore concerns about rural sprawl, but rather to develop conceptually robust and empirically implementable measures of urban sprawl.

As noted by Wassmer (2008), most of the independent variables needed for explanatory models are not available or measureable at the “EUA” level. The approach used in this paper is therefore to present measurement of sprawl as changes in “urban” areas within the spatial extent of each metropolitan statistical area. While based on the Census approach to “urban area” classification, the approach here (described in detail in section 3) does not impose the minimum population size of 50,000 for an area to be considered urban, thus mitigating some (but not all) of the concern of spatial boundaries.

This paper operationalizes four metrics of sprawl to measure relative (not absolute), marginal (not average) changes in density over time (not cross-sectional) in metropolitan areas, embedded within a framework derived from the monocentric model. Because of the need for analytical tractability metrics are computed in terms of housing-unit density.

Theoretical development of the sprawl metrics in this paper requires more detailed consideration of the implications of the monocentric city model, illustrated in Figure 2 with actual data from the 100 largest metropolitan areas for 1980 and 2000. Like Figure 1 above, shown is a nonparametric regression of the natural log of population density on the distance from the center of each metropolitan area. This analysis combines hundreds of thousands of census block-groups to illustrate a “standardized” or “equilibrium” pattern of urban form across US metro areas. The shape of the density gradient is stable over the 20 years of this study. Consistent with theory (increasing population and increasing real income), the density gradient shifted upward and outward from 1980 to 2000. New land on the urban fringe was converted to urban development (outward shift), and additional development density was added to lands already considered urbanized in 1980 (upward shift).
Consider Figure 3 which is a stylized representation of the model, showing the spatial extent of urban development shifting from $X_1$ to $X_2$ ($X_1 > X_2$). The right panel of Figure 3 shows that, from a “bird’s-eye view,” the spatial extent of urban development appears as a circle of radius $X_1$ and, later, $X_2$. In the first time period, the area of urban development, labeled as the dotted area “A”, is given by: $A = \pi X_1^2$. The area newly converted to urban development from time 1 to 2 on the urban fringe is labeled as the lined-area “B.” The new area of the city, $A+B$, is now $A+B = \pi X_2^2$.

To implement metrics of sprawl reflecting marginal land consumption and changes in land use patterns, the metrics presented here capture changes in density gradients, the change in urban area extent from $X_1$ to $X_2$, and the density of development in areas labeled “B” in Figure 3.

1. **Change in housing-unit density.** Housing-unit density in the first period (1980) is computed as $H_1/A$; housing units divided by urbanized area. Housing-unit density in the second period (2000) is given by $H_2/(A+B)$. The change in housing-unit density, representing the change in the slope of the density gradient, is therefore given by $[H_2/(A+B)]-[H_1/A]$.

2. **Marginal land consumed per net new urban housing unit** is given by $(H_2-H_1)/B$. This measures the efficiency with which a metropolitan area accommodates new households. If new housing units are built in areas already considered “urban” (area “A” in Figure 3), the marginal (urban) land consumption for that housing unit is 0. For new housing units constructed in newly urbanized areas (area “B” in Figure 3), marginal land consumption is measured as the area of “B” divided by the...
number of housing units in “B.” As a metric calculated over the entire metropolitan area, this measurement combines the percentage of new housing units added to previously urbanized areas and the density of lands newly urbanized.

(3) **Housing-unit density in newly urbanized areas.** This metric directly measures marginal changes in land use. Referencing Figure 3, this metric is the density of housing units in newly urbanized areas, area “B.” One advantage of the data set used in this paper is that the data are at the block-group level, permitting direct measurement of the actual number of housing units in newly developed areas. Such measures were not previously possible with remote-sensed sources.

(4) **Percent of net new urban housing units in previously urbanized areas.** This measure proxies “infill” (or the inverse of sprawl): areas with higher percentages of new urban housing units accommodated in previously urbanized areas are less sprawling. Because, as above, the data used to develop measures is built-up from block groups, the net increase in housing units in areas previously classified as urbanized in the first time period (1980) can be measured. A block-group might reach the population density threshold to be considered urban in 1980, while still having vacant or undeveloped land on which new housing could be built. This measure captures both what planners traditionally would call “infill” (development or redevelopment on previously developed lands) and what can be called “near-fill” (urban development directly adjacent to or within the same census block-group as already developed land).

3. **Data and methods.**

Previous research has been limited by nationally available data measuring both land use change and population variables in temporally and spatially consistent units. Even when sprawl has been conceptualized in dynamic and marginal terms, data limitations have required empirical implementation with static or average metrics.
For measures of land use patterns, (as the denominator in density calculations) researchers have relied either on Census data or on measures of developed land area from remote-sensing or surveys. While increased sophistication of remote-sensing and image processing opens the possibility of more refined measures of land use patterns in the future, the only consistently interpreted satellite-derived measurement of urbanized land area extent is the National Land Cover Database (NLCD) Change Retrofit project, covering 1992 and 2001 (Fry, et al. 2009). USDA’s National Resources Inventory (NRI) is sample-point data on land uses in 5-year intervals from 1982-2007. NRI data are no longer available publically while their methodology is revised and updated. Because NRI data are derived from samples, confidence intervals on small-area measurements are often quite large and statistically problematic. NRI and NLCD data, while useful in describing changes in land use patterns, are also not easily able to be matched to Census geographies where employment, transportation, household or population data is available.

The alternative approach to measuring land use patterns involves Census data itself. Implementing change analysis has been problematic because Census geographies and definitions change with each decennial census. Matching Census geographies across time and space is necessary to ensure that measures reflect real changes and not changes in areal units. The Census construct most closely representing the spatial extent of urban development (or “urban footprint”) is delineation of areas as “urban” (“urbanized areas” and “urbanized clusters”). The Census classifies areas as “urban” based on population density thresholds and proximity to other densely-developed areas. The difficulty for sprawl metrics is that the definition and classification-method for “urban” has changed every decade and is scheduled to change again in the 2010 release.

This present research overcomes these challenges by utilizing a newly published dataset on US urbanized land area (Paulsen 2012) using historically consistent census block-group boundaries and applying the Census 2000 urban area determination algorithm (described below) retroactively.
and consistently to 1990 and 1980 data. Each census block-group within a metropolitan area (using data in consistent year-2000 boundaries) is classified as whether or not it is “urban” in 1980, 1990 and 2000. In terms of the discussion in Figure 3, these data permit measurement of the exact size of areas “A” and “B” for all 329 metropolitan statistical areas (MSAs and PMSAs) in the continental US for 1980 and 2000, as well as the ability to count the number of housing units in each area for each time period. This 20-year time period is used to measure marginal changes in urban form.

The methodology and algorithm to develop urban-extent measures is described in detail in Paulsen (2012). All block-groups with a population density of 1000 persons/mile² (386 persons/km²) are identified as constituting an urban core. Then, adjacent block groups (allowing for up to half-mile “jumps” as per the Census) with a population density of 500 persons/mile² (193 persons/km²) are additionally classified as urban. This algorithm repeats in half-mile buffers until no adjacent block-groups meet the density threshold of 500 persons/mile². There is no minimum total population requirement to classify an area as urban. This method is repeated for each time period and the resulting product is a measure of the total area considered “urbanized” within each metropolitan area. Within each urbanized area, detailed calculations of changes in housing units and population can be made over time. Metropolitan areas may have more than one urbanized area.

The minimum population density to be considered “urbanized” is therefore 500 persons/mile². Even assuming a smaller-than-average household-size of 2 persons per housing-unit, and expressed in terms of land consumption per-housing unit (the inverse of density: land area divided by housing-units) this measure would capture housing-units developed at 2.56 acres per house (1.04 hectares per house). This measure thus partially addresses the “underbounding” issue in capturing very low-density suburban development.

Sprawl metrics presented in this paper measure housing-unit density in these classified “urbanized” areas, summarized over each metropolitan area. Housing-unit (rather than
population) density is utilized as a durable and tractable measure of urban spatial form. Each of the four sprawl metrics described in Section 2 (change in housing-unit density, marginal land consumption per net new housing-unit, density of housing-units in newly urbanized areas, and percent net new housing-units in previously urbanized areas) are calculated for each of the 329 metropolitan areas used in this analysis. Because of space constraints, rankings of metropolitan areas on these four metrics are not presented, but are available upon request.

Ordinal rankings of metropolitan areas vary significantly across metrics, indicating that combining variables into a singular sprawl metric could obscure significant variation. Although ordinal rankings vary, there is some degree of correlation across the metrics, as expected. The highest correlation (Pearson r = -0.84; Spearman’s rho = -0.76) is between the marginal land consumption metric and the percent of units in previously urbanized areas. This is inherent to how these metrics are computed. Marginal land consumption per household is a direct function of two other metrics (percent in previously urbanized areas and density in newly urbanized areas). There is still value in using each metric separately in the regression analysis because of different significant independent variables. However, interpretation should proceed with understanding that these metrics are not completely independent of each other.

As a test of the informational added-value of the marginal metrics (2, 3 and 4) relative to the average metric (1—change in average density), correlation tests were performed. The most significant correlation with the first metric was with metric 3, density of development in newly urbanized areas (Pearson r = 0.44; Spearman’s rho =0.30). These correlations do show the durability of metropolitan-specific density paths, but also demonstrate that marginal metrics add additional information to characterize changes in metropolitan spatial patterns.

Multivariate analysis consists of regressions where the four metrics of sprawl are dependent variables. The purpose of the regressions is to explain variation in sprawl across areas, utilizing as
controls a number of independent variables which theory and previous literature have shown to significantly effect land development patterns. For purposes of presentation, these independent variables can be categorized as “market,” “geographic” and “policy” factors, even though variables could be considered in more than one category.

To the extent that land development reflects market forces, explanation of changes across places includes variables reflecting prices, income and demographic demand factors. Median household income in 1980 (in 1999$) is included. Land price data is represented by the average value of agricultural land in each metropolitan area, based on Census of Agriculture county-level data. The value of agricultural land in each metro is a weighted average (by county size) of county-year observations of price data on each side of the decennial census population observation. For example, 1977 and 1982 agricultural values are interpolated and weighted by county-size to reflect average metropolitan land values for 1980. Land data are inflation-adjusted to 1999 dollars. Variables representing demographic demand characteristics include percentages of housing units owner-occupied and the percentage of residents over age 65. Also included is a measure of the share of the metropolitan population of racial or ethnic minorities. This variable is calculated as one minus the share of non-Hispanic whites in 1980. Although there is a literature on the effects of sprawl on racial segregation (see Galster and Cutsinger 2007; Rothwell and Massey 2009) this variable is more of an indirect test of the effects of racialized land-use decision making. Do jurisdictions in areas with higher minority populations adopt land use regulations to lower densities in order to exclude minorities? (Pendall 1999, 2000)

In terms of geographic variables, two variables reflect the historical evolution of metropolitan structure: housing-unit density for 1980 and the percentage of housing units built before 1950. Controls for metropolitan land area test whether growth patterns differ by city size. Also included is a measure of land availability for development, the percent of the metropolitan area
considered “undevelopable.” Although some measures of topographic constraints or coastal proximity have been utilized in previous sprawl research, the particular combination here (including “protected lands”) has not. The measure of undevelopable land area consists of three separate layers. First, NLCD 2001 national data is used to identify lands classified as water or wetlands. Second, using the EDNA (Elevation Dataset for National Applications) product, areas with greater than 15 percent slope are identified. Both NLCD and EDNA products are 30-meter resolution and cover the entire US. Third, the Protected Areas Database of the US (PAD-US) is utilized. The USGS GAP analysis program supervises and verifies this data, and federal, state, local and non-profit lands with GAP codes of 1, 2, or 3 (indicating permanent protection from land cover conversion) are coded as “protected” and therefore unavailable for development. Using geospatial overlay analysis to prevent double counting, these three land features (water/wetlands, steep slopes and permanent protection) are added together to measure the percent of land in each metropolitan area undevelopable. Because this measure includes federal lands and not just topographic constraints, it identifies important forces shaping metropolitan growth patterns. Recent work by Saiz (2010), for example, measures geographic constraints on housing supply with topographic features (slopes and water area) but not protected land areas. In the data set presented here, the metropolitan area with the largest percentage of land undevelopable is Las Vegas, driven by federal land holdings.

In terms of policy variables, two concepts frequently included in the literature are fragmentation and growth management. Local government fragmentation is frequently linked to sprawl outcomes (Carruthers and Ulfarsson 2002; Mills 2006). The general hypothesis in the literature is that greater levels of fragmentation are associated with increased sprawl. Following the literature, fragmentation is measured here as the number of general-purpose local governments per 1000 population (year 2000 population).
Three variables measure regional or state planning for growth management to reduce sprawl. First, data on urban containment as collected by Nelson and also utilized by Wassmer (Wassmer 2006; Nelson, et al. 2007) measure whether a metropolitan area had a regional “urban containment” policy in place during the time from 1980-2000. Although this variable has shown mixed results in empirical work (Paulsen 2013), it is one of the most widely available and used metrics of regional growth management. This variable is available for all 329 metropolitan areas and is implemented as a dummy variable equaling 1 for regions with urban containment. Alternative specifications measuring the number of years such containment was in place did not alter results and are not reported.

An alternative measure of growth management planning which has been utilized is to measure the strength of each state’s planning system in supervising and regulating local land use planning. This data was developed by the American Planning Association and the Institute for Business and Home Safety based on analysis of state planning enabling laws and frameworks. Each state’s role in managing local planning is classified as “weak,” “significant,” or “substantial”. A dummy variable for those states with a “substantial” state role (California, Florida, Maryland, New Jersey, Oregon, Rhode Island, and Washington) is used to test whether stronger state planning frameworks alter land development patterns. Because a strong state planning role and regional containment policies may complement each other, they both enter into the regression analysis as policy variables.

Alternatively, the Wharton Restrictions Index (WRI) is a measure of the overall restrictiveness of residential land development within a region, but is only available for the largest 94 metropolitan areas which cross state lines, a sensitivity analysis is conducted to determine whether state assignment affects results. Regressions are run with the primary state (identified in the Census) code. Multi-state metros are then re-assigned to the secondary state and regressions re-run, testing for any significant differences. Regressions are also run with a “multi-state metro” dummy variable, as in Wassmer (2006). Results confirm the validity of using the assigned primary state from the Census.
metropolitan regions (Gyourko, et al. 2008; Saiz 2010). The version of the index reported in Saiz (2010) is used here, and is constructed to represent multiple dimensions of land use restrictions in a consistent scale, higher numbers indicating a more restrictive development environment. In the regression results reported below, separate regressions utilizing the WRI are estimated because the urban containment and state planning roles are included in the WRI index. For each dependent variable, therefore, two models are estimated: one with the containment/state policy variables and a separate one (for only 94 metros) for the WRI.

The regression results for the 8 models are reported in Table 1. The format of each of the regressions is the same: the measure of sprawl or land use change from 1980-2000 is regressed on independent variables measured at the beginning of the time period (1980). Diagnostic tests indicate multicollinearity is not a problem (highest VIF is below 3). Diagnostic tests indicate the presence of heteroskedasticity, so each regression is estimated with robust standard errors. R² measures of fit perform better in the sample restricted to the 94 largest regions with WRI data, suggesting the possibility that these metrics developed from stylized urban models are more reflective of larger cities.

[Table 1 here]

The first two columns report change in housing unit density as the dependent variable. Positive (negative) numbers indicate that average housing-unit density within the metropolitan area increased (decreased) over the time period. In the full sample, positive coefficients on agricultural land prices are in keeping with expected land market effects: areas with higher valued land increased density over the 20 year time period, all else being equal. Metropolitan areas with a greater percentage of their land base undevelopable also saw housing-unit density increases. Consistent
with Figure 2, older center-dominated regions (greater percentage of housing before 1950) saw
density gradients flatten (reduced average density.) Metropolitan spatial patterns clearly respond to
land market signals expressed in prices and supply constraints.

The second set of regressions (columns 3 and 4) involve marginal land consumed per net
new urban household. Recall that this measure incorporates both the density of housing units in
newly developed areas and the percent of housing units as “infill,” and therefore measures the
overall land efficiency of a region in accommodating new housing units. Consistent with Fischel’s
(2001; 2004) “homevoter hypothesis” (areas with a higher percentage of housing units owner-
occupied are more likely to zone to protect property values by restricting density of new
development), regions with a higher percentage of units owner-occupied consumed more land per
net new housing unit.

In terms of policy variables, both the dummy variable for a substantial state planning role
and the WRI are significant and negative, indicating that each net new housing unit consumes less
land than housing units in weaker planning states or lightly regulated regions. For advocates of a
stronger state or regional role in managing local planning and growth, these results are consistent
with the idea that stronger planning institutions can potentially reduce sprawl or promote more
efficient land development patterns (Paulsen 2013). As seen when comparing across the next
regressions, a stronger state role in managing growth reduces marginal land consumption per
household not by affecting the density of new housing development but in (re)directing growth to
already urbanized areas.

In terms of the density of housing in newly urbanized lands (columns 5 and 6), a number of
variables reflecting market conditions and patterns of urban growth are significant. Larger
metropolitan regions and regions with higher initial densities saw denser development in newly
urbanized areas. Conversely, older regions (greater percentage of housing built before 1950) and
more supply-constrained lands (percent undevelopable) saw lower-density development. The negative coefficient on percent undevelopable is perhaps surprising. Comparing across to other metrics, areas with a greater percentage of land unavailable for development see overall densities increase because a greater proportion of new development occurs in already urbanized areas. In one sense, then, land supply restrictions function like a “natural” urban containment or growth management strategy (Nelson et al. 2007). However, land supply constraints result in lower density development in newly urbanized areas. These results are somewhat consistent with Saiz (2010) who finds that regions with more land supply constraints are more likely to adopt more restrictive development policies.

The variable measuring the percent of minority residents shows a negative relationship with marginal densities in newly urbanized areas. Interpretation of this coefficient requires some caution because its significance level is higher than 0.05. This negative coefficient would indicate that, all else being equal, housing development in newly urbanized areas is lower in regions with a larger minority population. The most likely explanation and mechanism for this result is that communities at the edge of urban development (suburbs) zone at lower densities to reduce the potential that minority households might move into their communities. This evidence would be consistent with the long-standing concern that lower-density zoning may have racially exclusionary intentions or effects (Pendall 2000; Galster and Cutsinger 2007; Rothwell and Massey 2009).

The final two columns explain variations in metropolitan “infill” patterns, the percent of new housing units in previously urbanized areas. Higher income areas show higher demand for infill, while a higher percentage of home-voters (owner-occupied units) is negatively related to infill. As expected, a higher percentage of the land base undevelopable (supply limitations) is associated with a higher degree of infill. Although this result is consistent with previous results indicating that physical constraints to development can act as either complements or substitutes for “growth
management” (Nelson et al. 2007; Saiz 2010), expansion of this measure to include protected areas could also suggest that policies promoting land preservation and conservation can lead to more compact urban development (Daniels and Lapping 2005). The positive coefficient on a substantial state planning role indicates that states with greater supervision and regulation of local planning and development accommodate a greater percentage of new urban housing units in already urbanized areas.

5. Discussion and conclusions.

Changes in spatial development patterns over time reflect the interaction of land markets and geographic structures, spatial patterns, and each region’s historical evolution. Worldwide, many cities appear to reflect the forces of dispersion and decentralization (Anas, et al. 1998) particularly as transportation costs decline and the necessity of concentrating employment in downtowns is eroded (Glaeser and Kohlhase 2004). Certainly that has been true in the United States, as shown in Figure 2. Cities in the US expanded in land area during the years studied, 1980-2000. For these 329 metropolitan areas, nearly 51,000 km² of land was converted to urban development and 18.5 million net new urban housing units were constructed. Over 12.1 million of these new urban housing units (65 percent) were accommodated in areas already classified as urban in 1980. The four metrics proposed here try to capture these land use changes as relative, marginal and dynamic phenomena and reveal important structural relationships in metropolitan development across the US, including policy initiatives to manage or reduce sprawl. Before “sprawl” can be remediated or constrained, there needs to be conceptual clarity on how to measure sprawl and its relationship to metropolitan structure. The metrics proposed in this paper are one way to conceptualize sprawl, but certainly not the only way.
Policies to constrain sprawl have generally been labeled “Smart Growth” or “growth management” or “urban containment,” and this research offers some insight concerning these policy instruments. First, the statistical insignificance of the variable measuring urban containment should not be taken to mean that urban containment policies do not affect spatial patterns of land development. Rather, it is possible that this widely used measurement, implemented as a dichotomous dummy variable, does not reflect important variations within containment regimes. Some regions coded as having containment, such as Portland, have strict urban growth boundaries. Others use a more growth-accommodating urban service area approach. The institutional complexity across containment regimes cannot probably be captured by existing measures (Paulsen 2013).

Second, the significance of the variable “substantial state planning role” is consistent with the idea that when local governments are required to prepare comprehensive plans and/or prepare growth management plans consistent with state policy objectives, urban development can be accommodated in a more compact manner. The results suggest that greater land-use efficiency in states with stronger planning is due to more development being directed to already developed areas, rather than more dense development in fringe areas. Likewise, areas which have more restrictive land development regulations (WRI) accommodate new housing units with greater land efficiency. Areas where a greater percentage of the land area is unavailable for development show increases in density and accommodate more of new development in previously urbanized areas.

These three policy measures – strength of state planning role, WRI, and percent of land available for development – all are suggestive of the potential that increased regional or state planning and regulation can reduce, at least in part, some of the land consumptive impacts of urban development. If sprawl is conceptualized as marginal changes in density over time within a metropolitan area, the evidence here is consistent with the idea that some policy interventions can
reduce the marginal land consumption associated with additional urban households. The metrics
and analysis presented here, however, cannot measure the welfare effects or costs of regional growth
management policies, something on which research has yet to achieve any consensus (Anas, et al.
1998; Cheshire and Sheppard 2004; Downs 2004).
References.


Estimated Population Density Gradients
New York and Los Angeles PMSAs 2000

LN Population Density
0  5  10  15  20  25
Distance to Mean Population Center in Miles

New York
Los Angeles
Figure 2.

Estimated Population Density Gradient
Large US Metros, 1980 & 2000

LN Population Density

Distance to Mean Population Center, in miles

--- 2000  --- 1980
Figure 3.

[Diagram showing a graph with the x-axis labeled as Distance and the y-axis labeled as Log of Population Density. The graph includes two lines and a region labeled CBD with points A, B, and X along the lines and the distance axis.]
### Table 1. Regression results.

<table>
<thead>
<tr>
<th></th>
<th>Change in urban housing unit density</th>
<th>Marginal land consumed per net new urban household</th>
<th>Density of housing units in newly urbanized areas</th>
<th>Percent of net new urban housing units in previously urbanized areas</th>
</tr>
</thead>
<tbody>
<tr>
<td>Median household income (1980, in thousand 1999)</td>
<td>-1.4544000</td>
<td>0.2500977</td>
<td>-0.0000207</td>
<td>-0.0000195</td>
</tr>
<tr>
<td></td>
<td>[0.379]</td>
<td>[0.914]</td>
<td>[0.121]</td>
<td>[0.188]</td>
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<tr>
<td>Average metropolitan agricultural land value (1980)</td>
<td>0.0058258</td>
<td>-0.0086855</td>
<td>0.0000000</td>
<td>0.0000000</td>
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<tr>
<td></td>
<td>[0.022]**</td>
<td>[0.199]</td>
<td>[0.652]</td>
<td>[0.138]</td>
</tr>
<tr>
<td>Percent of land undevelopable</td>
<td>80.0593000</td>
<td>101.7516000</td>
<td>-0.0001322</td>
<td>-0.0001678</td>
</tr>
<tr>
<td></td>
<td>[0.014]**</td>
<td>[0.078]*</td>
<td>[0.409]</td>
<td>[0.547]</td>
</tr>
<tr>
<td>Metropolitan land area</td>
<td>0.0038000</td>
<td>0.0016593</td>
<td>0.0000000</td>
<td>0.0001794</td>
</tr>
<tr>
<td></td>
<td>[0.306]</td>
<td>[0.733]</td>
<td>[0.311]</td>
<td>[0.717]</td>
</tr>
<tr>
<td>Metropolitan percent minority (1980)</td>
<td>-45.804000</td>
<td>-105.627000</td>
<td>0.0003225</td>
<td>-0.0003414</td>
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<tr>
<td></td>
<td>[0.471]</td>
<td>[0.339]</td>
<td>[0.464]</td>
<td>[0.541]</td>
</tr>
<tr>
<td>Percent residents over 65 (1980)</td>
<td>40.9785000</td>
<td>353.306000</td>
<td>-0.0012707</td>
<td>-0.0000125</td>
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<tr>
<td></td>
<td>[0.889]</td>
<td>[0.517]</td>
<td>[0.410]</td>
<td>[0.040]**</td>
</tr>
<tr>
<td>Percent housing units owner-occupied (1980)</td>
<td>-166.570000</td>
<td>-546.104000</td>
<td>0.0034560</td>
<td>0.0052900</td>
</tr>
<tr>
<td></td>
<td>(0.234)</td>
<td>(0.018)**</td>
<td>[0.001]**</td>
<td>(0.001)**</td>
</tr>
<tr>
<td>Percent of housing units built before 1950</td>
<td>-237.347000</td>
<td>-243.707000</td>
<td>0.0006284</td>
<td>0.0015014</td>
</tr>
<tr>
<td></td>
<td>(0.002)**</td>
<td>(0.136)</td>
<td>(0.282)</td>
<td>(0.040)**</td>
</tr>
<tr>
<td>General purpose governments per 1000 population</td>
<td>1.2865500</td>
<td>-2.0674800</td>
<td>-0.000109</td>
<td>-0.0000395</td>
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<tr>
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<td>[0.09]**</td>
<td>[0.406]</td>
<td>[0.282]</td>
<td>[0.040]**</td>
</tr>
<tr>
<td>Substantial state planning role (dummy)</td>
<td>25.1778000</td>
<td>-0.002696</td>
<td>77.593291</td>
<td>77.593291</td>
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<tr>
<td></td>
<td>(0.216)</td>
<td>(0.043)**</td>
<td>(0.191)</td>
<td>(0.043)**</td>
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<tr>
<td>Urban containment (dummy)</td>
<td>-10.0744000</td>
<td>0.0005112</td>
<td>-59.5524861</td>
<td>-0.0023233</td>
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<tr>
<td></td>
<td>[0.619]</td>
<td>[0.924]</td>
<td>[0.201]</td>
<td>[0.873]</td>
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<tr>
<td>Wharton Restrictions Index (WRI)</td>
<td>11.0653000</td>
<td>-0.0002317</td>
<td>49.6729743</td>
<td>0.0296547</td>
</tr>
<tr>
<td></td>
<td>(0.549)</td>
<td>(0.029)**</td>
<td>(0.126)</td>
<td>(0.262)</td>
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<tr>
<td>Constant</td>
<td>151.8960000</td>
<td>366.1930000</td>
<td>-0.000099</td>
<td>-0.0021962</td>
</tr>
<tr>
<td></td>
<td>[0.087]*</td>
<td>[0.041]*</td>
<td>[0.041]</td>
<td>[0.041]*</td>
</tr>
<tr>
<td>Observations</td>
<td>329</td>
<td>94</td>
<td>317</td>
<td>94</td>
</tr>
<tr>
<td>R²</td>
<td>0.169</td>
<td>0.371</td>
<td>0.121</td>
<td>0.401</td>
</tr>
</tbody>
</table>

Robust p values in brackets.

* significant at 10%; ** significant at 5%; *** significant at 1%