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APB VALUATION ADVISORY 5: IDENTIFYING COMPARABLE PROPERTIES IN AUTOMATED VALUATION MODELS FOR MASS APPRAISAL

The Appraisal Foundation
1155 15th Street, NW, Suite 1111, Washington, DC 20005
T 202.347.7722

APPRAISAL PRACTICES BOARD



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APB Valuation Advisory #5

Identifying Comparable Properties In Automated Valuation Models for Mass Appraisal

This communication is for the purpose of issuing guidance on recognized valuation methods and techniques. Compliance with such guidance is voluntary, unless mandated through applicable law, regulation, or policy.

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Application: Automated Valuation Models in Mass Appraisal

Issue: As part of its ongoing responsibilities, the APB is tasked with identifying where appraisers and appraisal users believe additional guidance is required. One such issue identified by the APB is identifying comparable properties. Comparability analysis is a fundamental study in determining property value. This analysis involves a side-by-side examination of physical and transaction characteristics of the identified comparable properties relative to the subject. The reliability of this valuation technique relies heavily on the proper selection of suitable comparable properties.

The APB discovered in developing guidance on Identifying Comparable Properties that separate guidance on *Identifying Comparable Properties in Automated Valuation Models for Mass Appraisal* was required.

This guidance discusses the terms and definitions associated with a comparable property, the characteristics generally considered for determining comparability; and the degree of suitability of a property as a comparable when using automated valuation and mass appraisal models.

Additionally, a general overview of mass appraisal's automated collateral comparable selection process is included.

The proposed guidance addresses the sales comparison method - direct market models using sales transactions. Model specification and model calibration are discussed. The guidance

provides sample model structures using the direct market method. Comparable sales selection is discussed in detail.

Lastly, the guidance examines appraisal uniformity based on the equity comparison method.

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Grant W. Austin – New Smyrna Beach, Florida
Anthony M. Graziano – Toms River, New Jersey
Michael Ireland – Bloomington, Illinois
Karen Oberman – Clive, Iowa
Jo Anne Traut – Brookfield, Wisconsin

APB Liaisons: Guy Griscom and John S. Marrazzo

Identifying Comparable Properties for Automated Valuations in Mass Appraisal Models

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NOTE: This Valuation Advisory includes examples that utilize proprietary software applications. Neither the Appraisal Practices Board nor The Appraisal Foundation endorses or recommends utilization of the particular software applications referenced in this document. They have been included to assist in the illustration of the applicability of such applications.

I. Introduction

1 The terms Mass Appraisal Models¹ and Automated Valuation Models (AVM)² involve the use of
2 comparable properties in their market analysis methods. The term AVM is used in this guideline
3 to represent both mass appraisal and automated valuation models keeping the focus on data
4 requirements used in the development of valuation models. “The development of an AVM is an
5 exercise in the application of appraisal principles and techniques, in which data are analyzed
6 from a sample of properties to develop a value prediction model that can be applied to a similar
7 population of properties that represent subject properties. These may be either individual
8 properties of interest or all properties that meet the requirements of the model.”³

9 The Sales Comparison Approach is applied using the AVM via several methods, the two
10 discussed in this guideline are: 1) direct market models; and 2) comparable sales models. A third
11 method used and of interest to appraisers doing ad valorem work is the equity comparable
12 method. The direct market models and the comparable sales models begin with the validation of
13 all transaction data to be used in developing the model. These sales include transactions deemed
14 as arms-length, as they are the most credible. When specific market segments are influenced by
15 “distress” sales (e.g., short sales and foreclosures) these transactions may be used in the model
16 training set. A model training set in this guideline consists of data used to discover predictive
17 relationship in the market place.

18 Income approach models require analyzing leases (rents and occupancy), expense ratios, and
19 capitalization rates in completing the model specification phase. Validation of the comparable
20 includes verification of the property characteristics for each transaction at the time of its
21 consummation. Readers of this document may also want to review the “sister” document,
22 *Identifying Comparable Properties*, as there is some overlap of definitions and concepts.

23 Users and readers of this guidance are encouraged to keep the methods and techniques described
24 in context with their specific scope of work, see STANDARD 6, USPAP⁴. An AVM developed
25 for use in collateral, or portfolio valuation will have a different scope of work than an AVM
26 developed for ad valorem use. Awareness of administrative rules, case law, statutory
27 requirements and client request are required.

II. Sales Comparison Method Direct Market Models using Sales Transactions

28 Primary for developing a direct market model are model specification⁵ and model calibration⁶.
29 Model specification involves indentifying all property characteristics to use in the model. These
30 characteristics include features that represent location, physical, economic, and legal interest.

¹ IAAO. 1997. Glossary for property appraisal and assessment. Chicago: IAAO.

² Ibid

³ IAAO, 2003 Standard on Automated Valuation Models (AVMs) page 5, Chicago

⁴ *UNIFORM STANDARDS OF PROFESSIONAL APPRAISAL PRACTICE* (USPAP) 2012-13 Edition page U49

⁵ IAAO, 1997. Glossary for property appraisal and assessment. Chicago, IAAO.

⁶ Ibid

31 Accurate property characteristics are important in model development as each sale is processed
 32 during calibration via a comparative analysis that relies on measuring price changes as the
 33 property characteristics change. The sales used for calibration of direct market models also are
 34 used in the comparable sales method, as the adjusted comparables.

35 Accurate property characteristics and sales validation provide the cornerstone for achieving
 36 credible results in the development of an AVM. The purpose of automated valuation models is
 37 to estimate a value for many properties by using an iterative process of data and analysis. The
 38 influence of market areas and neighborhoods are captured using proximity variables, be it the
 39 street level, neighborhood, or market area. Market influences from another city, state, or
 40 country, are addressed when specifying the model. When properly specified and calibrated,
 41 these variables will identify and account for the influence of physical location on value.

42 When an AVM is specified using only sales in a specific neighborhood, the model is only
 43 applicable to properties within the specific boundaries. If a subject property were located in a
 44 neighborhood different from where the comparables are located, the credibility of the AVM
 45 suffers due to the fact the AVM may not identify or measure any of the economic influences
 46 specific to the subject. The concept of using locational similarities is helpful in identifying all
 47 local effects in the marketplace that have influence on a given subject’s property value. To draw
 48 a distinction, not all neighborhoods or market areas will provide a sufficient sample of
 49 transactions for each and every subject property. When the pool of market transactions is
 50 insufficient, identification and selection of comparable requires the use of additional physical
 51 property characteristics. When market transactions are insufficient in number to provide credible
 52 results, selecting transactions from a location outside of the subject’s neighborhood, or market
 53 area is often required.

The table below provides a focus regarding example proximities that influence the value direction. These proximities variables when specified in the AVM are useful in selecting properties with similar market influences. Adjusting values when the influence is not present or not entirely similar may be required. Markets are dynamic thus any positive can become negative and vice-versa, Table 1 provides an illustration.

TABLE 1: Examples of Location and Proximity Characteristics with a Value Influence

Sample Proximity Categories										
	Schools	Employment Centers	Shopping	Public Trans	Water Front	Golf Courses	Public Parks	Airports	Rail	Highway
Property Class										
Residential	+	+	+	+	+	+	+	-	-	-
Income Producing				+	+		+	+/-	+/-	+/-
Industrial				+	+			+	+	+

54 The previous discussion on the development of the AVM focused on specifying the model
 55 specifically for variations due to a subject property’s proximity or location. Model specification
 56 includes the use of physical characteristics as well. The AVM must be capable of valuing a
 57 subject that is different in size, age, condition, style, exterior wall type, quality, and utility as

58 examples. When the AVM is to value vacant land, features such as lot size, frontage, depth,
59 view, topography, ingress, egress, lot shape, lot location, (e.g., corner lot, interior lot) land use,
60 zoning, etc. are characteristics to be considered. Each property characteristic specified in the
61 AVM is calibrated using the known transaction prices as the dependent variable upon which all
62 market adjustments are based for each of the specified property characteristics.

63 Property characteristics used in an AVM measure the economic influences, and how each
64 contributes to the value of a subject property. User confidence in the credibility of AVM values
65 will suffer when an important property attribute is either missing, incorrectly specified or
66 calibrated during the modeling process. When working with property data it is important for
67 appraisers to learn about data exploration techniques, to avoid problems with missing data, and
68 models that have been mis-specified, resulting in irrational and unexplainable calibration of a
69 variable. The number of subject properties to which the AVM is applicable is the “hit rate.”⁷ If
70 a subject is outside of acceptable parameter ranges for location or property characteristics, the
71 AVM will not possess sufficient comparability to value the subject property with a high degree
72 of accuracy or credibility.

III. Sample Model Structures - Direct Market Method

73 Direct market models are expressed in one of three math structures: additive, multiplicative, or
74 hybrid. The skill and experience level of the analyst and software availability are also factors
75 involved with choice of the model structure. The additive model is the most common.

76 *Additive* models have the form:

$$77 \text{ MV} = B_0 + (B_1 X_1) + (B_2 X_2) + (B_{xx} X_{xx}) \dots$$

- 78 • MV is the dependent variable (Sale Price, or Sale Price per Unit);
- 79 • B_0 - is a constant dollar amount;
- 80 • X_i - represents the independent variables (property characteristics) in the model: X_i
81 represents X_1, X_2 etc. from above model notation, and
- 82 • B_i are corresponding adjustment rates or “coefficients” applied to X_i , B_i represents B_1, B_2
83 etc. from the above model notation

84 In the model above **MV** also known as the dependent variable and is either “sale price” or “sale
85 price per unit.” When modeling **income** the dependent variable is “income” or “income per
86 unit.”

87 In a *multiplicative* model, the contribution of the variables are multiplied rather than added:

$$88 \text{ MV} = R_0 * (X_1^{B_1}) * (X_2^{B_2}) * (X_{xx}^{B_{xx}}) \dots$$

⁷ IAAO, 2003 Standard on Automated Valuation Models (AVMs) page 28, Chicago

89 Multiplicative models consist of a base rate (R_0) and percentage adjustments. Base rates are
90 expressed as dollars, or a percentage (e.g., vacancy rate, expense ratios, GIMs, OARs). The
91 multiplicative model advantages include the ability to capture curvilinear and interactive
92 relationships and to make adjustments proportionate to the value of the property appraised.
93 Multiplicative models can be calibrated using linear regression software; doing so will require
94 the variables to be converted to a logarithmic format for calibration. Non-linear regression can
95 be used also and does not require the logarithmic transformation.

96 *Hybrid* models are a combination of additive and multiplicative models. Hybrid models allow
97 specifications that separate value into buildings, land, and “other” components (e.g.,
98 outbuildings). The model format is:

$$99 \text{ MV} = \pi\text{GQ} \times [(\pi\text{BQ} \times \Sigma \text{BA}) + (\pi\text{LQ} \times \Sigma \text{LA}) + \Sigma \text{OA}]$$

- 100 • MV is the estimated market value;
- 101 • πGQ is the product of general qualitative terms (e.g., location influences market
102 conditions);
- 103 • πBQ is the product of building qualitative terms: (e.g., quality, condition, wall types, and
104 styles);
- 105 • ΣBA is the sum of building additive terms; (e.g., total living area, total rooms, baths, FP’s);
- 106 • πLQ is the product of land qualitative terms: (e.g., view, traffic, topography, egress);
- 107 • ΣLA is the sum of land additive terms: (e.g., lot frontage, lot area); and
- 108 • ΣOA is the sum of other additive terms: (e.g., pools, out buildings, accessory structures).

109 A full discussion on the topic of model structures is available in the IAAO Standard on AVMs.⁸
110 Calibration of the models, developing coefficients, or rates of adjustment may be completed
111 using applications like linear multiple regression analysis (MRA) and non-linear MRA.

112 Automated valuation models require data representative of the full range of transaction prices
113 and associated property characteristics in a defined area. The sold properties used for market
114 comparison, and unsold subjects to be valued must represent comparable segments of the market
115 population. The AVM relies on market data from standardized property coding methods.
116 Collectively all the validated sold properties represent the potential comparables for any known
117 subject.

118 Model performance and validation is also an important step to having a direct model that will
119 produce credible results. The IAAO Standard on Automated Valuation Models (AVMs) may be
120 referenced for completing this step.⁹

⁸ Ibid pages 10-12

⁹ Ibid page 24

IV. Sales Comparison Method – Comparable Sales Selection

121 As with single property appraisal, Automated Valuation Models (AVMs) rely on arms-length
122 sales primarily to adjust a model to local real estate trends. In market areas where properties
123 have a high degree of homogeneity, the AVM will produce value estimates with a high degree of
124 precision. Areas where the market is heterogeneous the error term of the AVM is usually higher
125 and not as precise. The sales comparison model is a familiar format to appraisers and expressed
126 as follows:

127 $MV = SP_c + ADJ_c$, where:

- 128 • MV = Market Value;
- 129 • SP_c = Sale Price of the comparable; and
- 130 • ADJ_c = adjustment to the comparable \pm .

131 As the integrity of the data used to calibrate an AVM decreases, the precision and accuracy of
132 the value estimate is often deemed less credible. Using automation to select comparable
133 properties that will produce credible and reliable value estimates is the challenge of the AVM.
134 USPAP compliance is attainable using automated selection methods (see the discussion on data
135 verification). The following will provide the steps leading to selecting the most comparable
136 properties for a single subject.

137 Step (1) is **selecting the best comparables**. Numerous techniques are used in practice with the
138 “exception, or sorting techniques” being most popular. The most efficient and objective
139 technique is the use of a distance metric that assigns a dissimilarity index to each sale in the file.
140 The more similar a comparable property is to the subject, the lower its metric value
141 “dissimilarity index.” Two indexes used frequently are the “Squared Euclidean distance
142 metric”¹⁰ and the “Minkowski metric” (differences are not squared). These distance metrics rely
143 on a selection model or algorithm specified by the appraiser to find the most comparable
144 property (e.g., sales, rents, expenses) used for comparison to a subject.

145 The Euclidean distance metric is expressed as:¹¹

$$146 \sum_i^k [W_j(X_{ij} - X_{ij})/\sigma_j]^2$$

147 Where:

- 148 • Σ = the summation of all the weighted squared attribute difference;
- 149 • W_j = a weight assigned by the user to the j -th attribute;

¹⁰ IAAO Glossary for Property Appraisal and Assessment, IAAO.org

¹¹ *Fundamental of Mass Appraisal*, IAAO, 2011 p 144

- 150 • S_{sj} = the value of the j -th attribute for the subject;
- 151 • X_{ij} = the value the j -th attribute of the comparable property;
- 152 • σ_j = the standard deviation of the j -th attribute for the comparables;
- 153 • 2 = squaring creates an absolute scale of difference; and
- 154 • K = the number of attributes over which comparability is defined.

155 Assigning weights to the selection characteristics is the most difficult aspect of using a
156 dissimilarity index. The higher the weight the greater the importance assigned to the
157 characteristic. As the weight on a characteristic is increased, the algorithm will allow less of a
158 difference in that characteristic.¹²

159 Tables 2 and 3 on the following page provide an example demonstrating how the metric
160 calculation works:

¹²Ibid144

Table 2: Example of Euclidean Distance Metric -- Comparable No 1

Attribute	Subject	Comparable #1	Difference	Standard Deviation	Standardized difference	Appraiser-assigned weight	Weighted difference	Squared weight difference
Living Area	1800	1660	140	225.7	.6203	2	1.2406	1.5391
Age	10	4	6	4.1	1.4634	1	1.4634	2.1416
Quality	3.5	5	-1.5	0.9	-1.6667	1	-1.6667	2.7778
The sum of the standardized, weighted squared differences is 6.4585 based on the three selection attributes chosen								6.4585

Table 3: Example of Euclidean Distance Metric -- Comparable No 2

Attribute	Subject	Comparable #2	Difference	Stand Deviation	Standardized difference	Appraiser-assigned weight	Weighted difference	Squared weight difference
Living Area	1800	1700	100	225.7	.4431	2	.8861	.7852
Age	10	4	6	4.1	1.4634	1	1.4634	2.1416
Quality	3.5	5	-1.5	0.9	-1.6667	1	-1.6667	2.7778
The sum of the standardized, weighted squared differences is 5.7046 based on the three selection attributes chosen.								5.7046

161 Comparable No. 2 is closer in size to the subject and has a metric score of 5.7046, lower than the
162 score of 6.4585 for Comparable No. 1. Each sale would require the same number of adjustments
163 but the total adjustment for Comparable No. 2 would be less than for Comparable No. 1.

164 Minkowski metric differs from the Euclidian, in that absolute differences are used rather than
165 squared differences making a smaller metric scale¹³.

166 Division by the standard deviation also provides the algorithm the ability to standardize the
167 magnitude in data differences that are measured or recorded in different scales, similar to the use
168 of Z-scores.

169 Once a distance metric is calculated for all potential comparables, the properties are sorted based
170 on their metric. The comparables most similar to the subject are those with the lowest metric
171 value.

172 The higher the weight on an attribute the more important that property attribute is in the selecting
173 similar properties. Initially weights are assigned to those property attributes identified as the
174 most influential in explaining why prices change from property to property. Indicators helpful
175 with weighting include; regression coefficients (weights), beta coefficient, correlation
176 coefficients, and T-statistics. Assigning weights forces the algorithm to select comparables
177 whose attributes are most similar to the subjects. The appraisers, skill, knowledge, and
178 experience also play a role in knowing how much weight to assign an attribute. Selecting
179 comparables without disclosing the attribute(s) that are most important has been the practice
180 primarily used by appraisers. Using a dissimilarity index exposes the importance of specific
181 attributes in contrast to attributes considered less important having lower weights or no weight
182 assigned.

183

¹³ Ibid145

184 **Comparable Sales Selection Method**

185 Table 4: Provides a Distance Metric, Weighting Example, from the NCSS Statistical Software¹⁴

186 **Table 4**

Adjustment Variable:	Distance Weight:	Amount (\$ or %):
TOTAL_SF	50	0
SIZEADJ	0	-\$65000
FIN_BS_SF	0	\$20
SF_FIN_ATT	0	\$64.15
LV_MV	0	\$1
AC	0	\$25
NO_FP	0	\$2500
NO_BATHS	0	\$4500
PORCH_SF	0	\$5
DECK_SF	0	\$8
GAR_NO	0	\$5500
POOL_SF	0	\$25
WALL_COEFF	0	10%
GRADE_07QUAL	50	100%
DEP1	50	100%
NH_07FINL_QUAL	25	100%
STY_TYPE	100	

Template Id: Sale Grid Adjustment 12-12-07

187 The characteristics (variables) Total Square Feet (Total_SF), Quality (GRADE_07Qual),
 188 Depreciation (Dep1), Neighborhood, or Location (NH_07FINL_QUAL), and Style, or Story
 189 Type (STY_TYPE) are weighted in the selection process. The sales selected for the adjustments
 190 have attributes closest in these defined categories collectively and are placed on the adjustment
 191 grid. Defining the metric selection model is a skill in attribute importance, and magnitude of the
 192 data. Using a distance metric selection will improve the appraiser’s impartiality from personal
 193 judgment with an objective method.

¹⁴Hintze, J. (2012). NCSS 8. NCSS, LLC. Kaysville, Utah, USA. www.ncss.com. Comparable – Sales Price

194 Table 5 below, indicates five (5) property characteristics that are used in the selection process.
 195 Story Type receives the most weight, while Quality, Condition and Location were each given
 196 equal weights. The following Table 6 provides a look at the metric scale that is automatically
 197 calculated using the formula provided.

Table 5: Example of Characteristics used in comparable selection				
Variables	Weight	Mean	Standard Deviation	COV
TOTAL_SF	10	1847.1	760.0	41
Grade_07Qual	25	1.1	0.2	17
Dep1	25	0.8	0.1	14
NH_07FINL_Qual	25	1.0	0.1	9
STY_TYPE	200	6.5	4.5	69

198 *Of the 18917 properties on the database, 1 was a subject property and 16988 were excluded*
 199 *by the selection variable(s), leaving 1928 comparables for consideration.*

200 **Distance Report for Subject = 14-25-103-031**

Table 6: PARCEL_ID	Distance Metric
14-25-103-031	Subject
14-25-103-022 – Comp 1	0.00187
15-31-176-024 – Comp 2	0.01199
14-25-103-037 – Comp 3	0.01741
14-25-102-011 – Comp 4	0.02305
14-25-103-011	0.02519
15-30-155-001	0.02913
15-30-152-013	0.03246
15-31-128-009	0.03501
15-31-128-007	0.03550
20-13-276-007	0.03842

201 The NCSS software allows the user to display the metric report for (n) comparables, in this
 202 example. Of the ten (10), the top four move to the grid for adjustment.¹⁵

203 Step (2) – **adjustment methods for the selected sales.** Ideally, adjustments to the selected
 204 comparable are based on market parameters derived from a direct market model calibration
 205 process used in the AVM development. Using and adjusting comparables from the sales file
 206 provides a level of detail and precision pertaining to the subject that exceeds what is available by
 207 using only the direct market model values. Since the coefficients (adjustment rates) are derived
 208 from the transactions and have been tested using various statistical methods within the modeling
 209 process, the appraiser is presented another level of independence, impartiality, and objectivity.

¹⁵ Ibid

210 In the prior example of the automated comparables, the third column holds the rate of adjustment
 211 for the specific property attribute. The rates of adjustment are the same coefficients used in the
 212 direct market model and were derived from a non-linear calibration method shown in Table 7. If
 213 a direct market model does not exist, other “market intuitive” adjustments are substituted. In an
 214 article on the use of automated comparables method, the author addressed the matter of
 215 adjustments noting, “It appears that appraisers prefer to say that an adjustment is their opinion
 216 confirmed by market research rather than to be responsible for explaining the research itself.”¹⁶
 217 Appraisers have stated that the adjustments on the market grid in the sales comparison approach
 218 are market derived without further explanation. Automating the comparable selection process
 219 does nothing to defuse that claim. Comparables selection must be emphasized over the
 220 adjustment process; the importance of finding comparables that are as similar as possible to the
 221 subject greatly reduces the need for making many (sometimes controversial) adjustments to the
 222 sales prices.

223 Table 7 provides an explanation of the property attribute’s on the grid, and the adjustment
 224 process.

Table 7		
Adjustment Variables	Adjustment Value	Adjustment Explained
YRMO	-0.003	% Market Change Rate based on study of resale’s
TOTAL_SF	\$0	See SIZEADJ
SIZEADJ	\$-65000	Base rate per total sq.ft. using size curve
FIN_BS_SF	\$20	Finished basement sq.ft. rate
SF_FIN_ATT	\$64	Finished attic sq.ft. rate
LV_MV	\$1	Land value established; adjust comparable to equal value.
AC	\$2500	Lump Sum Adjustment-Air Conditioning
NO_FP	\$2500	Lump Sum Adjustment-Fire Place
BATH_COEFF	\$4500	Lump Sum-per 3 fixture bath
PORCH_SF	\$5	Square foot rate for porch
DECK_SF	\$8	Square foot rate for deck
Gar_NO	\$5500	Lump Sum Adjustment-per garage stall
POOL_SF	\$25	Base rate for pool square feet
WALL_COEFF	75%	Percentage adjustment for difference in Exterior Wall Type
Grade_07Qual	\$50000	Percentage adjustment for difference in quality, a 5% difference is a \$2,500 adjustment
Dep1	50%	Percentage adjustment for difference in condition
NH_07FINL_Qual	10%	Percentage adjustment for difference in location
STY_TYPE	\$0	Style or Story type forced to be identical in selection or nearly identical, no adjustment.

¹⁶ Commercial/Industrial Property and Sales Comparison Approach, IAAO 1986

225 The SIZEADJ is used to capture the economy of scale built into the model instead of actual
226 square footage. The adjustment for size is non-linear and transformed on the grid to make the
227 correct adjustment in an additive fashion, and in the proper direction.

228 Step (3) – **Opinion of value from the selected sales after adjustment:** “It is important that the
229 appraiser consider the strengths and weaknesses of each value indication derived, examining the
230 reliability and appropriateness of the market data compiled and the analytical techniques
231 applied.”¹⁷ Appraisers are provided little guidance to help them with what the “**most**
232 **comparable property**” actually means. Words like “reliability” and “appropriateness” are
233 subjective terms and add nothing to the appraiser’s independence, impartiality, or objectivity.

234 As AVMs become more universal, the idea of using statistical measures to evaluate
235 comparability will be an acceptable alternative to the previous manual and subjective method.
236 When comparable selection is based on distance metrics of each attribute compared to the
237 subject, user - assigned weights provide added influence actual differences. In turn, the
238 weighting of an attribute implies greater importance on specific attributes. The first comparable
239 on the adjustment grid is typically the most comparable property available. The remaining
240 comparables are scaled based on similarity to the subject and overall comparability. A
241 comparability index may be assigned giving the appraiser an objective measure for each sale on
242 the grid prior to adjustment. The single point estimate can then be based on:

- 243 • Closest – the comparable property that is closest in proximity to the subject property.
- 244 • Min |\$| Change - a comparable property that had the smallest absolute dollar value
245 change. This property received the smallest dollar adjustment amount.
- 246 • Simple Average - adjusted prices of the M comparables that have the smallest distance.
247 User chooses M as number of comparables in average.
- 248 • Weighted Average. - Use the weighted average of the adjusted prices of the M
249 comparables that have the smallest distance to the subject property. The weights are
250 proportional to the distance from the comparable property to the subject property.

251 These choices are programmed into the reporting function chosen by the appraiser in the final
252 analysis. The use of closest, and Min|\$\$| Change tend to put weight on a single adjusted value,
253 while the simple average and weighted average base the value conclusion on more than a single
254 property as instructed by the appraiser.

255 Disclosure - It is possible to use a spreadsheet in order to customize the comparable selection and
256 display. The guideline presents a comparable selection application already deployed that can
257 also be fully customized to the user’s needs. Many vendors of appraisal reporting software also
258 include various selection methods. The APB does not endorse any specific software; the NCSS
259 used here is simply available to all as are other off-the-shelf applications.

¹⁷ *The Appraisal of Real Estate*, 13th ed, p 428

260 See Table 8: Automated Comparables Report, on the following page for a full example of the
261 output from an automated comparables program.¹⁸

¹⁸ Generated using, NCSS 2007 Ver.07.1.18 Comparable – Sales Price, released June 29, 2009, J. Hintze

Table 8: Automated Comparables Reports

Page/Date/Database	1 4/24/12 /AVMCOMPS V3						
Filter	FLAG=1,2,3;YRMO=>200901;land_use=1;sale_use=3						

Comparative Sales Price Adjustment Report for Subject = 14-25-103-031									
	Subject	Comparable No. 1		Comparable No2		Comparable No 3		Comparable No 4	
PARCEL ID	14-25-103-031	14-25-103-022		15-31-176-024		14-25-103-037		14-25-102-011	
	Value	Value	\$Adj	Value	\$Adj	Value	\$Adj	Value	\$Adj
ADDRESS	2501 KARA CROSSING	30 HODGEHAVEN CIR		38 CONWAY CR		2513 KARA CROSSING		7 HODGEHAVEN CIR	
Neighborhood #	65	65		79		65		65	
SF CRAWL	0	0		0		0		0	
SF SLAB		-						-	
Story Type	2 Story	2 Story		2 Story		2 Story		2 Story	
Wall Type	Brick Front	Brick Front		Alum/Vinyl		Brick Front		Alum/Vinyl	
Quality Name	B+5	B+5		B+5		B+5		B+5	
Year Built	1998	1997		2001		1997		1996	
Lot square feet	17,919	10,583		10,557		10,767		10,920	
Year month sale	199804	200901		201003		201103		201104	
Direct Model Val	\$ 296,951	\$ 295,114		\$ 296,989		\$ 286,463		\$ 270,118	
Comparability		100%		100%		99%		99%	
Sale Price / Market Conditions Adj.		\$260,000	(\$18,720)	\$290,000	(\$8,700)	\$267,500	\$1,605	\$227,500	\$2,048
Total Sf	2,954	2,796	\$0	2,853	\$0	2,424	\$0	2,357	\$0
SIZEADJ	0.8273	0.8402	\$837	0.8354	\$528	0.8746	\$3,073	0.8815	\$3,522
Fin. Basmt	0	610	(\$12,200)	1040	(\$20,800)	1030	(\$20,600)	1113	(\$22,260)
Fin. Attic	0	0	\$0	0	\$0	0	\$0	0	\$0
Land Value	\$ 64,929	\$ 50,799	\$14,130	\$ 53,121	\$11,808	\$ 51,321	\$13,608	\$ 51,759	\$13,170
% AC	1	1	\$0	1	\$0	1	\$0	1	\$0
Fire Place	1	3	(\$5,000)	2	(\$2,500)	1	\$0	1	\$0
# Baths	2.5	3.5	(\$4,500)	2.5	\$0	3.5	(\$4,500)	2.5	\$0
Porch SF	195	0	\$975	129	\$330	0	\$975	142	\$265
DECK SF	0	0	\$0	0	\$0	0	\$0	196	(\$1,568)
Garage SF	3	3	\$0	3	\$0	3	\$0	3	\$0
Pool SF	0	0	\$0	0	\$0	0	\$0	0	\$0
Wall Adj	1.057	1.057	\$0	1	\$11,571	1.057	\$0	1	\$9,519
Quality Adj	1.2824	1.2824	\$0	1.2824	\$0	1.2824	\$0	1.2824	\$0
Cond - % Good	0.9275	0.92	\$883	0.95	(\$3,175)	0.92	\$981	0.9125	\$1,741
NH - Loc Adj	0.9667	0.9667	\$0	0.9945	(\$776)	0.9667	\$0	0.9667	\$0
Story Type /Style	10	10	\$0	10	\$0	10	\$0	10	\$0
Net \$Adj.			(\$23,595)		(\$11,714)		(\$4,858)		\$6,438
Sum \$Adj.			\$38,525		\$51,488		\$43,737		\$52,046

APB Valuation Advisory #5 - Identifying Comparable Properties In Automated Valuation Models for Mass Appraisal

Adj Sales Price	\$	252,817		\$236,405		\$278,286		\$262,642		\$233,938
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262 Table 8, Automated Comparables Report, provides the appraiser information supplemental to the
263 adjustment about each sale. These supplemental attribute items are listed above the “**Sale**
264 **Price/Market Conditions Adj**” on the grid. This display is optional and not all comparables
265 applications provide this option.

266 Special conditions required of the appraiser by the scope of work can limit the available
267 comparable pool within a specific location. The span of time by transaction date can reduce the
268 pool of available comparables to be considered if, for example, transactions six months old or
269 less are prescribed as part of the valuation problem. All sales in the example report are within
270 one mile of the subject property, identical in style, and similar in size, age and condition. The
271 transaction data is limited to within 36 months of the appraisal date; however, sales are adjusted
272 for appropriate changes in market conditions, also known as a time-of-sale adjustment.

273 An automated comparable sales selection method saves time and improves consistency in
274 adjustments and results. The automated selection method provides a sophisticated methodology,
275 with the simplicity of explanation that is common with single property comparison methods.
276 Appraisers in the private sector have been entering comparable data into forms software since the
277 mid-1980’s. The recent addition of the Uniform Appraisal Dataset (UAD) provides a method of
278 standardizing comparable data for selection, thus improving results. The UAD will assist the
279 appraiser in developing the AVM and adding value to their product by including an objective,
280 impartial technique for selecting comparables. The automated comparables selection method
281 saves time and builds efficiency into the appraisal process while providing accuracy and
282 credibility not available in the manual comparables comparison approach.

283 **V. Appraisal Uniformity - Equity Comparison Method**

284 A required component in ad valorem mass appraisal is demonstrating the appraisal uniformity or
285 equitable treatment of identical or similar properties during the valuation process. Many court
286 cases and laws support the notion of equal treatment when developing value opinions for the
287 purpose of ad valorem taxation. The court case carrying significant weight is **Allegheny-**
288 **Pittsburgh Coal Co. v. County Comm'n, 488 U.S. 336 (1989) Allegheny Pittsburgh Coal Co.**
289 **v. County Commission of Webster County, West Virginia.** In part, the United States
290 Supreme Court ruled; “constitution and laws provide that all property of the kind held by
291 petitioners shall be taxed uniformly according to its estimated market value.”¹⁹ Most states
292 require equality of assessment as part of the mass appraisal process. Practitioners should also
293 check for specific jurisdictional laws and/or rules governing uniformity and equal treatment.

294 A comparison method available that will reveal uniformity of values and equitable treatment is
295 the previously described comparable sales method, with the caveat that the *appraised value* of
296 the selected comparables is substituted for variable sale price on the comparison grid. All the
297 adjustments are market-derived coefficients used to value all the property in the class. If the

¹⁹ <http://supreme.justia.com/cases/federal/us/488/336/>

298 mass appraisal is a cost approach model the cost coefficients are used as the rates of adjustment
299 on the grid. This method provides users with an instant view of how the subject property is
300 valued in comparison to other properties and how the adjustment rates affect the value. Relying
301 on uniformity of appraised value as the measure of comparability increases the pool of
302 comparables as no sale is required for this method, the chance of selecting identical or near
303 identical properties is greater than with comparable sales selection. The number of adjustments
304 and their amounts is usually minimized by use of decidedly similar property characteristics.

305 Appraisal uniformity checks are also helpful as a model performance tool to ensure the accuracy
306 of the data, and the model deployment methods produce consistent values on identical property
307 characteristics. In the text, *Mass Appraisal of Real Property*²⁰ the idea of using a basehome
308 model is presented as another method for supporting the AVM and explaining the concept of
309 appraisal uniformity for the market area or class of properties subject to AVM application.

310 **VI. Definitions**

311 **Additive Model**

312 A model in which the dependent variable is estimated by multiplying each independent variable
313 by its coefficient and adding each product to a constant. *Glossary for Property Appraisal and*
314 *Assessment, p 3*

315 **Assessment Equity**

316 The degree to which assessments bears a consistent relationship to market value. *Glossary for*
317 *Property Appraisal and Assessment, p 11*

318 **Automated Valuation Model**

319 An automated valuation model (AVM) is a mathematically based computer software program
320 that produces an estimate of market value based on market analysis of location, market
321 conditions, and real estate characteristics from information that was previously and separately
322 collected. The distinguishing feature of an AVM is that it is a market appraisal produced through
323 mathematical modeling. Credibility of an AVM is dependent on the data used and the skills of
324 the modeler producing the AVM. *Standard on Automated Valuation Models (AVMs) IAAO 2003,*
325 *p 5*

326 **Computer Assisted Mass Appraisal (CAMA)**

327 A system of appraising property, usually only certain types of real property, that incorporates
328 computer-supported statistical analyses such as multiple regression analysis and adaptive
329 estimation procedure to assist the appraiser in estimating value. *Glossary for Property Appraisal*
330 *and Assessment, p 28*

331 **Direct Market Method/Analysis**

²⁰ IAAO 314 W 10th St. Kansas City MO 1999 p207, R.Gludemans, R.Almy,

332 In the direct market method, the market analyst specifies and calibrates a single model used to
333 estimate market value directly using multiple regression analysis or another statistical algorithm.

334 **Euclidean Distance Metric**

335 A measure of distance between two points “as the crow flies.” In property valuation, it is used to
336 find the nearest neighbor or similar property based on an index of dissimilarity between property
337 location and attributes. When using multivariate selection, the squared difference is divided by
338 the standard deviation of the variable so as to normalize the differences. (Also see Minkowski
339 Metric.) *Glossary for Property Appraisal and Assessment, p 52*

340 **Hybrid Model**

341 A model that incorporates both additive and multiplicative components. See also additive model
342 and multiplicative model. *Glossary for Property Appraisal and Assessment, p 67*

343 **Mass Appraisal**

344 The process of valuing a group of properties as of a given date, using standard methods,
345 employing common data, and allowing for statistical testing. *Glossary for Property Appraisal
346 and Assessment, p 85*

347 **Mass Appraisal/AVM Model**

348 A mathematical expression of how supply and demand factors interact in a market. *Glossary for
349 Property Appraisal and Assessment, p 85*

350 **Minkowski Metric**

351 A way of measuring distance. In mass appraisal model building it is the sum of absolute
352 differences in each dimension. *Glossary for Property Appraisal and Assessment, p 87*

353 **Model Calibration**

354 **Calibration:** The process of estimating the coefficients in a mass appraisal/AVM model.
355 *Glossary for Property Appraisal and Assessment, p 19.* **Model Calibration:** The development
356 of adjustments, or coefficients based on market analysis that identifies specific factors with an
357 actual effect on market value. *Glossary for Property Appraisal and Assessment, p 88*

358 **Model Specification**

359 The formal development of a model in a statement or equation, based on data analysis and
360 appraisal theory. *Glossary for Property Appraisal and Assessment, p 88*

361 **Multiplicative Model**

362 A mathematical model in which the coefficients of independent variables serve as powers
363 (exponents) to which the independent variables are raised or in which independent variables
364 themselves serve as exponents; the results are then multiplied to estimate the value of the
365 dependent variable. *Glossary for Property Appraisal and Assessment, p 89.*

366 **Population**

367 All the items of interest, for example, all the properties in a **jurisdiction** or **neighborhood**; all
368 the observations in a data set from which a **sample** may be drawn. *Glossary for Property*
369 *Appraisal and Assessment, p 104*

370 **Regression Coefficient**

371 The coefficient calculated by the regression algorithm for the data supplied that, when multiplied
372 by the value of the variable with which it is associated, will predict (for simple regression) or
373 help to predict (for multiple regression) the value of the dependent variable. For example, in the
374 equation, Value = \$10,000 + \$5,000 x number of rooms, \$5,000 is a regression coefficient.
375 *Glossary for Property Appraisal and Assessment, p 118*

376 **Representative Sample**

377 A sample of observations from a larger population of observations, such that statistics calculated
378 from the sample can be expected to represent the characteristics of the population being studied.
379 *Glossary for Property Appraisal and Assessment, p 120*

380 **VII. Suggested Further Reading**

381 International Association of Assessing Officers, *Glossary for Property Appraisal and*
382 *Assessment.*

383 International Association of Assessing Officers, *Standard on Mass Appraisal of Real Property,*
384 *April 2013.*

385 International Association of Assessing Officers, *Standard on Automated Valuation Models*
386 *(AVM's), September 2003.*

387 Thimgan, G. E., *Property Assessment Valuation*, 3rd ed. (Kansas City, MO: International
388 Association of Assessing Officers, 2010).

389 Gloudemans, Robert J., *Mass Appraisal of Real Property*, (Kansas City, MO: International
390 Association of Assessing Officers, 1999).

391 Gloudemans, Robert J. and Richard Almy, *Fundamentals of Mass Appraisal*, (Kansas City, MO:
392 International Association of Assessing Officers, 2011).

393 NCSS Statistical Software – Data Analysis & Graphics at www.NCSS.com.

394 IBM SPSS Statistical Software at www.IBM.com.

395 XLSTAT Statistical Analysis for Microsoft EXCEL at www.Analyse-it.com.