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Automated Valuation Models:
a brave new world?

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Agenda

- Automated Valuation models - AVMs
- Examples of valuation (appraisal) accuracy
- Challenges
- Quant stuff (time permitting!)
 - analysing prices and valuations
 - analysing errors

Importance of accurate residential property valuations – some examples

- Property purchase is normally an individual's biggest lifetime purchase and an error, say, +/- 5%, in the valuation is a significant component.
- Mortgage providers assessing Loan to Value (LTV) ratios, i.e. the value of collateral on a mortgage.
- Residential values as a proportion income/earnings ratio (credit risk)
- Property values and attendant taxes.
- 'Wealth effect' – macroeconomic impact resulting from spending.
- Borrowing (re-mortgaging).
- Loan impacts on banks' balance sheets.
- Regulators and Central Banks do not want a repeat of the impact of valuation inaccuracies in the boom years preceding a collapse in market values (implementation of macro-prudential policies/regulation so as to avoid financial stress in the economy).
- Accuracy of valuation based residential property indices.
- Mis-valuation implications for investment portfolio asset holdings/asset allocation.

AVMs

- AVMs have their origins in North America, the first commercial application being in 1981, and began to be developed in the UK in the 1990s.
- Despite traditional approaches being extensively employed in the valuation profession, there has been a significant growth in independent residential Automated Valuation Model (AVM) providers, who offer their services routinely on a fee-based basis, to both lenders and the fee-paying public.
- AVMs are widely used by lenders and institutional investors, largely for monitoring purposes, and are seen as complementary to traditional valuations. The widespread use of AVMs is now firmly established.
- These computer-assisted quantitative methods have some advantage in that they are systematic and fast, thereby reducing reliance on labour input in providing an end-to-end valuation.
- By removing the human element, it is claimed by some advocates, it also reduces inaccuracies due to reliance on human judgement. This is an unsubstantiated assertion.
- The overall attitude and degree of acceptance of such automated approaches to valuation varies.

An AVM prediction ?



What is an AVM 1?

- Although different underlying AVM models are employed by vendors, fundamental to the approach are *statistical, data mining and computing technicalities*.
- TEGoVA provide the following, Definition 2.1, in their European Valuation Standards EVIP 6:

‘Automated Valuation Models (AVMs) can be defined as statistic-based computer programmes, which use property information (e.g. comparable sales and property characteristics etc.) to generate property-related values or suggested values.’
- The International Association of Assessing Officers, IAAO (2003), describes an AVM as

‘a mathematically based computer software programme that produces an estimate of market value based on analysis of location, market conditions, and real estate characteristics from information collected. The distinguishing feature of an AVM is that it produces a market valuation through mathematical modelling. The credibility of an AVM is dependent on the data used and the skills of the modeller producing the AVM.’

What is an AVM 2?

- The following definition of an Automated Valuation Model is provided by the RICS AVM Standards Working Group:
‘Automated Valuation Models use one or more mathematical techniques to provide an estimate of value of a specified property at a specified date, accompanied by a measure of confidence in the accuracy of the result, without human intervention post-initiation.’ (RICS 2013).
 - A key component in the RICS definition is the qualification ‘...*accompanied by a measure of confidence in the accuracy of the result...*’.
- ➔ All three definitions of an AVM exclude any appraiser involvement in arriving at a value.

Practitioner attitudes towards AVMs

- An international survey undertaken in 2008 on AVMs and the integration of AVMs within the valuation process provides some interesting findings. There were 473 valuer responses, representing both lending and valuation organisations, and described as senior professional members with ‘much experience of mortgage valuations’. The results of the survey include the following :
 - 71% of the valuers agreed that AVMs were inadequate for loan valuations as a result of no physical inspection.
 - 87% of the valuers agreed that physical valuations were more accurate than AVMs, as a result of local knowledge.
 - 90% of valuers agreed that the ability to evaluate comparables was a major advantage over AVMs.

Information on AVMs

- Debate regarding the role and accuracy of AVM valuations is an ongoing topic of discussion. For there to be a meaningful debate, AVM vendors need to make available access to their models for independent testing and verification of the models' output and accuracy of the results.
- Whilst there are a large number of AVM vendors, the inner functioning of the models and details of their specification are not released, nor are 'accuracy' figures usually disclosed.
- Vendors claim they test their models regularly for accuracy, and some may have the figures independently assessed. However, this non-disclosure puts a constraint on the analysis which can be externally undertaken as regards the assessment of the reliability and accuracy of the models.
- Other than submitting information to rating agencies, European/UK AVM operators are unwilling to have their data/methodologies exposed to independent scrutiny. *Vendors argue that their accuracy figures need to be put into a wider perspective.*
- US AVM market is highly developed and 'accuracy' figures are available on websites.

Measuring AVM's accuracy 1

- A variety of ways to measure AVM valuation error. First, a *benchmark reference* needs to be established:
 - is the AVM forecast measured relative to a valuer's estimate of the property's price or,
 - is the AVM forecast measured relative to the market price achieved in the market?
- Fitch, for example, evaluate model accuracy based on surveyors' assessment of values.
- UK vendors do not provide any details.

Measuring AVM's accuracy 2

- Several US vendors publish accuracy results based on achieved sales prices.
- How can accuracy be evaluated ? Several ways, for example:
 - can look at the percentage of AVMs falling within a specified range of error, for example within:
 - +/- 5%, +/-10%, or +/-20% of the sales prices
 - the 'spread' around the 'average' of all errors (FSD)
 - the median (50% of values less/greater than the median)
 - average of the absolute errors (MAD/MAPE)
- AVM vendors typically qualify their valuation estimates by providing a *prediction range* with a specified degree of confidence.

Margin for error

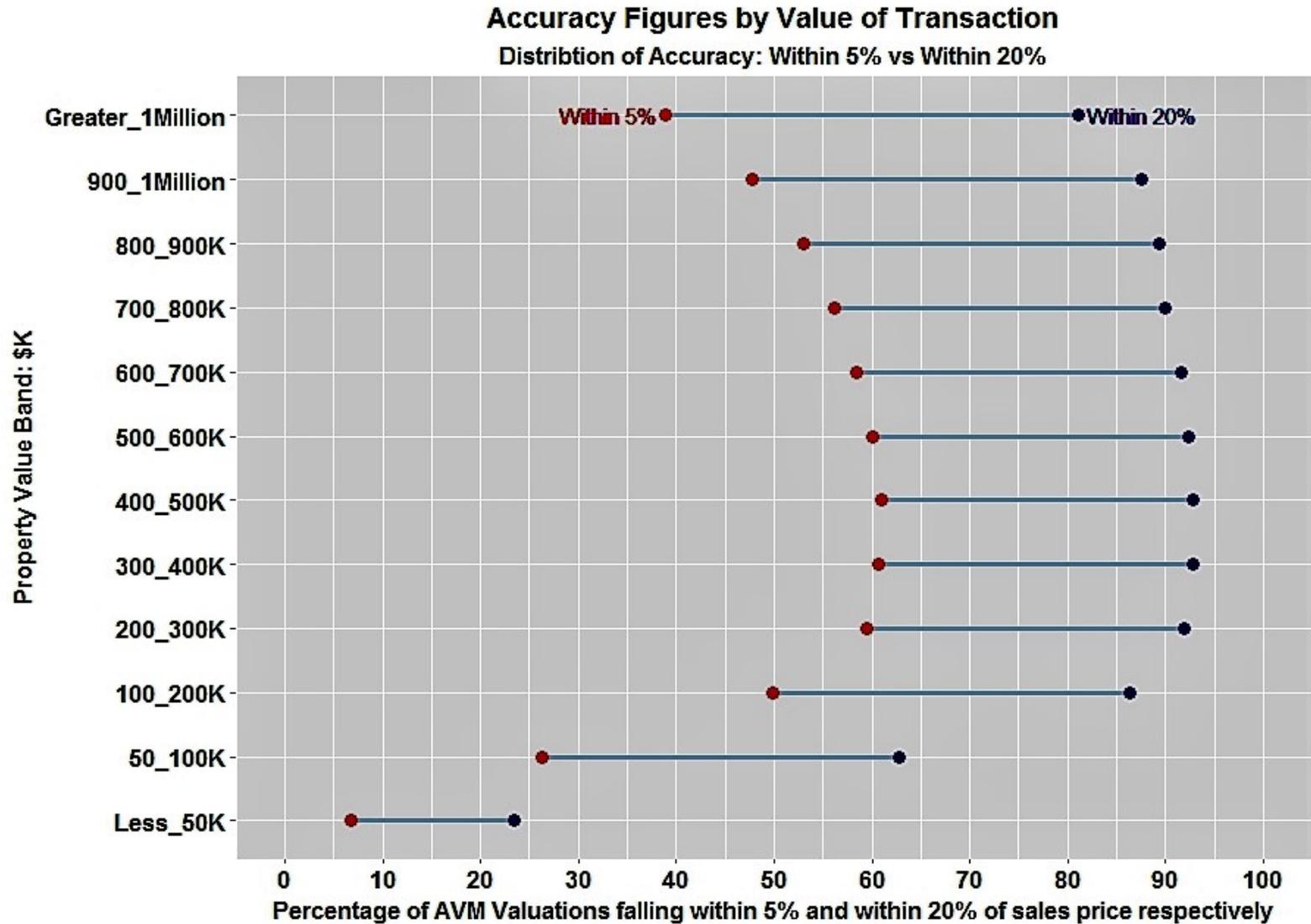
- What is an acceptable margin for error level?
 - a long established concept developed in (UK) case law.
- Very little evidence for residential property markets.
- Market conditions will likely raise or lower the margin.
- Evidence from commercial real estate markets may provide an indication of an upper limit on inaccuracy.
- Examples:
 - valuations compared against achieved market prices.
 - some country comparisons: absolute average errors and the range.

<u>Country</u>	<u>Average Absolute Error (%)</u>	<u>Range</u>
Netherlands	9.1	6.5-13.0
Italy	10.3	6.3-13.8
Germany	11.7	9.1-15.6
UK	10.7	8.5-13.3
France	11.0	7.3-14.5
Sweden	12.5	8.7-19.0
Japan	12.0	7.4-17.2
Canada	11.2	7.8-14.1
USA	9.9	6.7-14.7

*Netherlands, Germany, UK, France, Sweden, Canada, USA: 2000-2016
Japan: 2005-2016
Italy: 2006-2016*

Source MSCI, 2017

Distribution of accuracy - HouseCanary



Source: Figures courtesy of HouseCanary

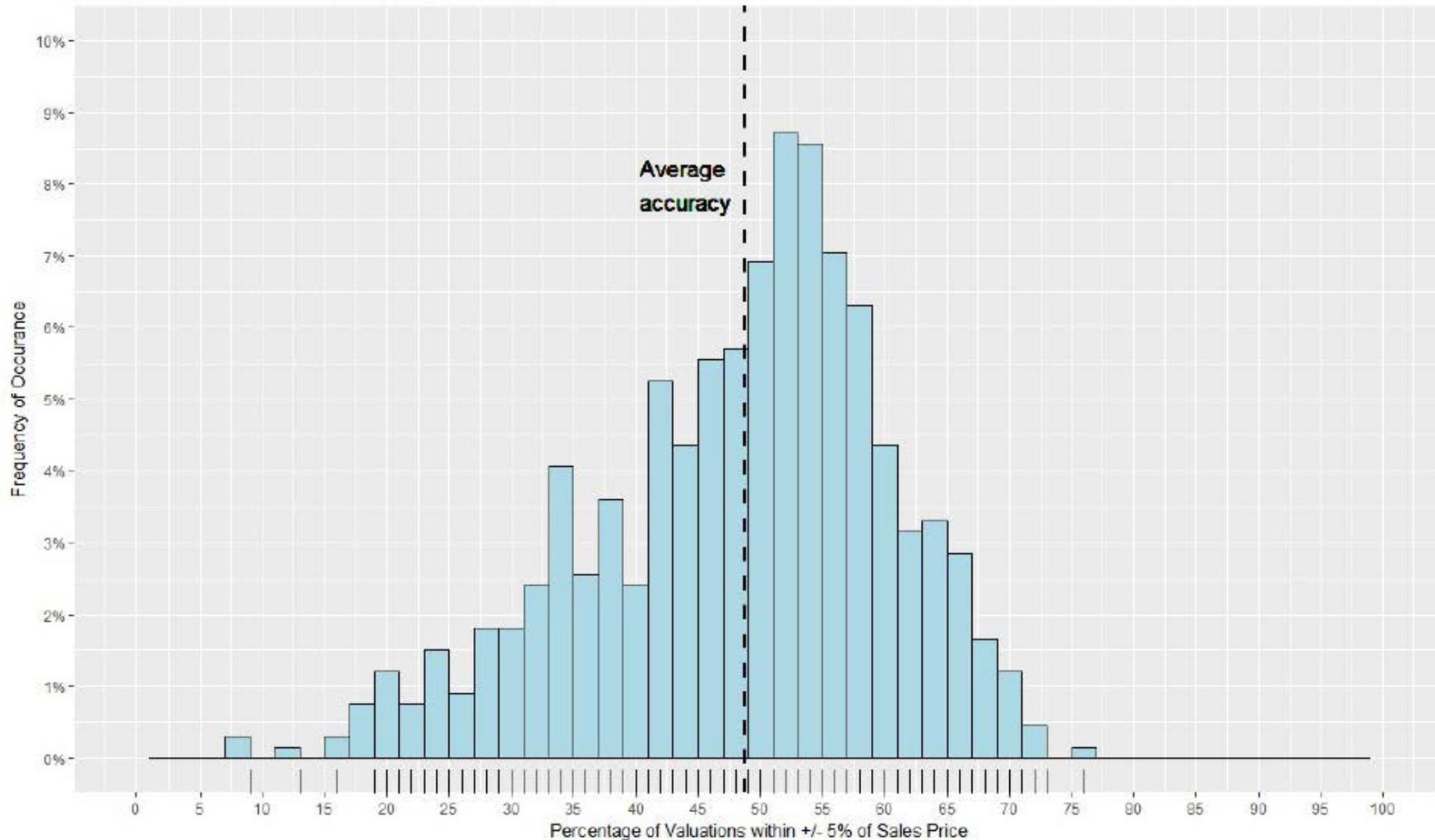
Summarising HouseCanary's accuracy numbers

- The figures show the percentage of properties valued within +/- 5% of the sales price and within +/- 20% of the sales price, across 12 price bands.
- What stands out for the lower valued properties, less than US\$ 50K, only 7% of the valuations were within +/- 5% of the achieved sales prices, rising to 23% being within +/- 20% of the sales price.
- For property values in excess of US\$ 100K, the figures show that in excess of 80% of values are within +/- 20% of the sales price.
- If 10% is seen as an acceptable margin for error, for properties with values in excess of US\$ 100k, some 75% of the valuations are within +/- 10% of the sales price; one-quarter of AVM generated appraisals will have errors greater than +/- 10%.
- The *median absolute percentage error* across all properties (888 metropolitan statistical areas, MSAs, as at June 2017) is 5.8%, i.e. half of the errors nationwide were within 5.8% of the final selling price, and half exceed 5.8 %.

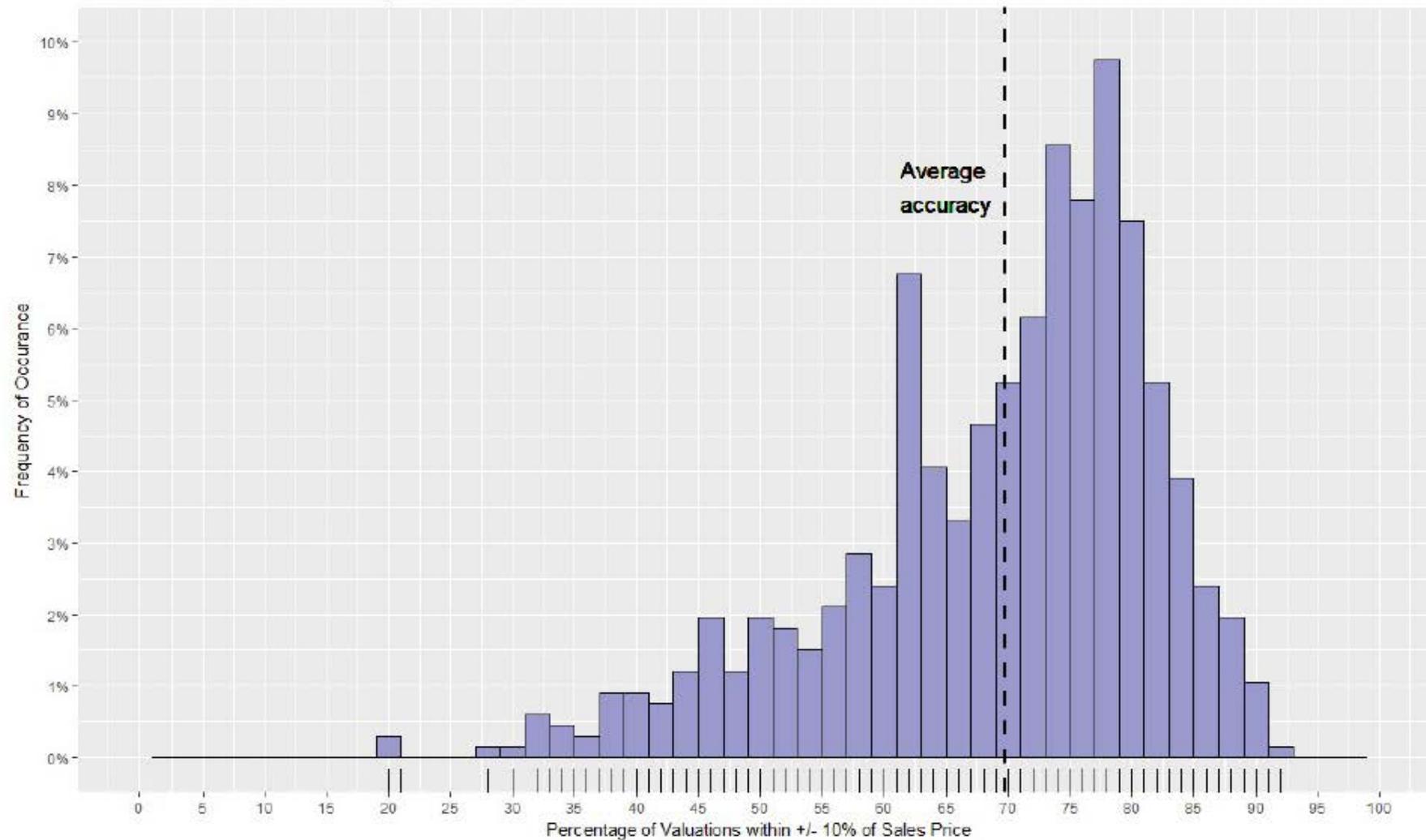
Accuracy numbers – Zillow 1

- Zillow claim to be the largest AVM provider in the US.
 - The following three histograms show the distribution of Zillow's AVM accuracy rates for 666 US Counties.
- shows the distribution of AVM valuation accuracy *within each of the individual 666 Counties.*

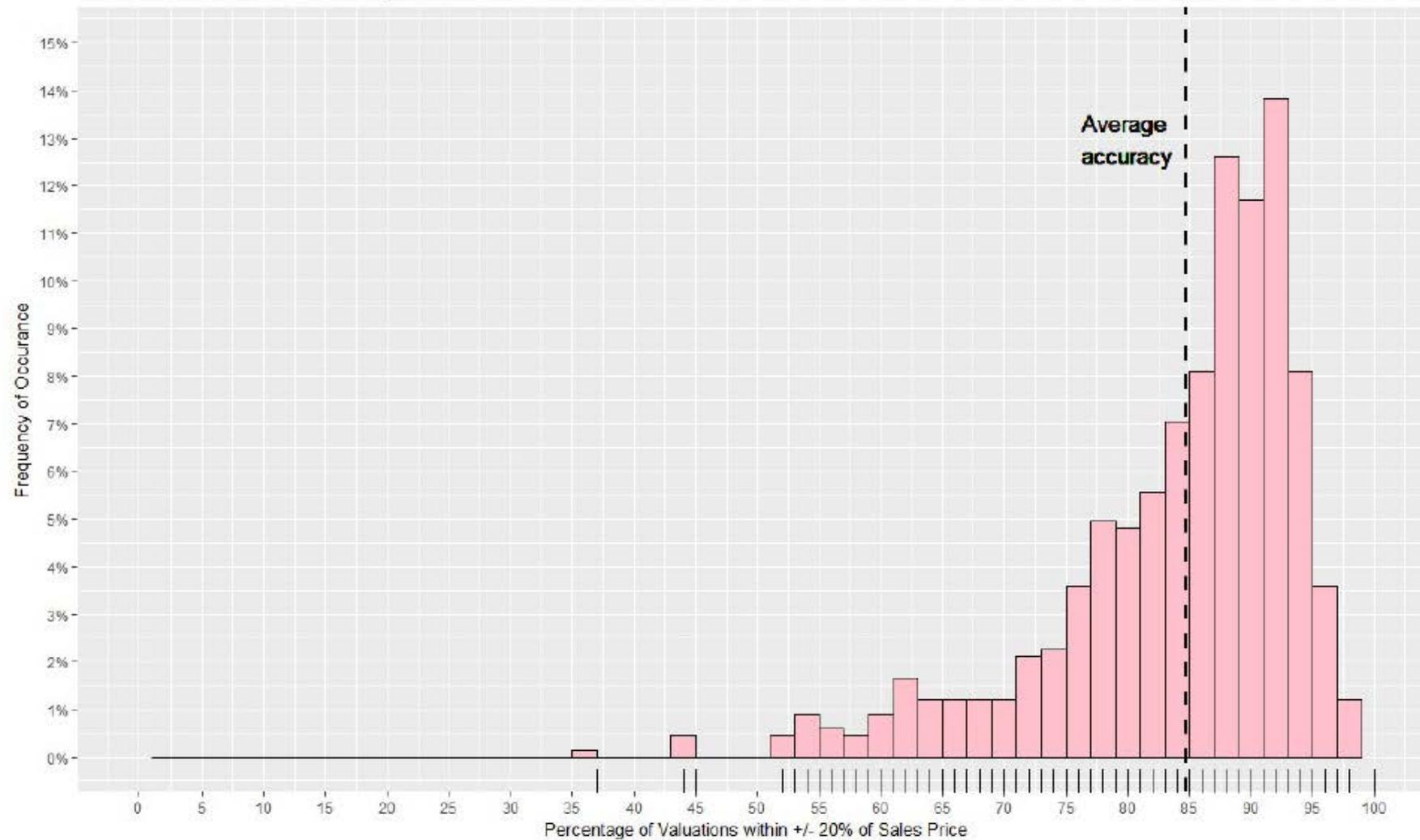
F1: Zillow's distribution of Valuation accuracy within +/-5% of sales price



F2: Zillow's distribution of Valuation accuracy within +/-10% of sales price



F3: Zillow's distribution of Valuation accuracy within +/-20% of sales price



Accuracy numbers - Zillow 2

- The median level of valuation error across 666 Counties in the US is 6.0%, i.e. half of the errors nationwide were within 6% of the final selling price, and half had an error exceeding 6.0%.
- Figure F1: shows +/- 5% valuation accuracy rates; Figure F2: +/- 10% valuation accuracy rates and Figure F3: +/- 20% valuation accuracy rates. The Histograms provide a detailed visual insight across the 666 US Counties.
- The average accuracy across all counties is superimposed in order to provide a reference point. On balance, it appears that some 50% of the valuations are likely be outside the +/- 5% range of achieved sales price, which falls to 30% for the +/- 10% range and 15% for the +/- 20% range.
- *As shown in the histograms, given the skewed nature of the distributions, even at the wider range of +/- 20%, there exist a significant proportion of valuations in many locations which lie outside the specified ranges of accuracy.*

Summarising Zillow's accuracy figures

The figures show the following:

- The median level of valuation error across 666 Counties (as at February 2017) in the US is 6.0%. Meaning, half of the errors nationwide were within 6% of the final selling price, and half exceed 6.0% errors.
- At the *individual County level*, the median ranged from 3% to 25%, which represents a wide range of variation across the different locations.
- On average, almost half of all valuations across all Counties were within +/- 5% of the sales price, half being excess of +/- 5%. However, in one County only 9% of the valuations were within the 5% bracket, with highest recorded accuracy figure being 76%.
- On average, the percentage of valuations across all Counties falling within +/- 10% of the sales price is 70%. This can vary between 20% and 92%, depending on the County.
- On average, the percentage of valuations across all Counties falling within +/- 20% of the sales price is 85%. However, this can vary between 37% and 100%, depending on the County.

Valuation under different conditions

- Is the margin for error unchanging i.e. constant? Will most likely vary under different conditions.
- There is a whole series of circumstances which would need to be taken into consideration when looking to assess what would be an acceptable margin for error in valuing residential properties, including the following:
 - different market environments, such as rising/falling prices
 - different size/value properties
 - quality of property
 - age of property
 - market liquidity e.g. dependent on the volume of transactions
 - geographic location/different neighbourhoods
 - type of property
- All are likely to vary *by* country and *within* each country!

Data considerations

- Current shortcomings of AVMs in the US are much more information based rather than methodology based.
 - current condition of most properties is largely unknown from the typical property level datasets being used today (especially for properties which have not been listed or sold for a long time).
 - secondary data deficiencies include home amenities that are important to consumers, but are not readily measurable in current property level datasets:
Examples:
1: does the home have a desirable open floor plan, 2: is there good natural light coming into the home, 3: is there good usable outdoor space on the lot.
- Above type of data is available today, but only for a very small portion of the entire housing stock, but prohibitively expensive.
- As the data becomes cheaper and more broadly available, the information it can provide will make it into valuation models and one can expect error rates will continue to decline.

Acknowledgement: Insight and observations credited to discussions with HouseCanary.

Data considerations: summary

- Improvement in AVM models will come with increased access to above types of data.
- Data quality improvements at the micro-data level required. The technologies are there but the costs currently inhibit across the board adoption.
- General assumption is that only marginal gains will result from new algorithms used within the final prediction model.
 - a caveat:
developing new algorithms used for non-traditional data feature extraction and/or faster processing times, will remain very important.

Challenges 1

- The independent validation and standards of validation of European AVMs needs to be promoted more vigorously, otherwise the role of AVMs will continue to be contested. How best to proceed?
- In the absence of regulatory/enforceable controls, there has to be a commercial or reputational benefit to the vendor in order to make it worthwhile for them to provide information.
- Independent professional bodies qualified to scrutinise AVMs
 - setting a standard of best practice for AVM vendors
 - access to the underlying models
 - access to database(s) on which the models are calibrated/estimated/tested
 - access to AVM output under different market environments and any adjustment made to pure model generated forecasts
 - AVM accuracy/standardisation of published accuracy measures
 - clear definitions/standards for transparent 'testing' of AVMs procedures

Challenges 2

- Certification, implying positive publicity.
- Alternatively, more voluntary information from AVM vendors, but issues of impartiality arise.
- Collating users' AVM experience.
- Establishing more comprehensive micro-oriented databases.

Observations and questions

- The distribution of the accuracy figures of the US models, *across* both locations and *within* locations, appears to provide tolerable results within what could be considered acceptable levels of statistical confidence.
- However, a purely statistically derived or data-mined valuation (an AVM) risks being widely off the mark, lying well outside the +/- percentage ranges, as shown in the Zillow charts for example.
- Despite the high degree of accuracy reported by some US AVM vendors, there still remains a requirement for professional judgement to augment model-based valuations in arriving at a more broadly considered opinion of value.
- Requirement for more discussion about what is a fitting framework for assessing and evaluating AVMs. Effective independent validation of AVMs is hampered by the lack of industry standardisation across virtually all aspects of the AVM process, including access to underlying data, models and accuracy results.
- Vendors' concerns about commercial sensitivity/intellectual property of their products prevail.
- Does the valuer possess additional information to that contained in the AVM?
- Are AVMs more accurate than physical valuations?

Final thought

- AVMs are extensively used around the world and have become part of the valuation environment. Given the widespread deployment of AVMs, their use is not an either-or question, but a question of *how* can an AVM enhance a valuer's estimate of value.
- The position has been well summarised by a significant provider of valuation technologies to the mortgage banking industry as follows:

“AVMs are going to get more and more mainstream, particularly as data and analytics get more sophisticated. AVMs won't take the place of an appraisal. There will always be a need for local knowledge and expertise, not to mention an on-site evaluation of the physical property.”

Would this take you down the pub for a drink?

Update W , the weights of the output layer

For a particular weight w_{JK} (from units Z_J to Y_k)

$$\begin{aligned}\frac{\partial E}{\partial w_{JK}} &= \frac{\partial}{\partial w_{JK}} \left(0.5 \sum_{k=1}^m (t_k - y_k)^2 \right) = \frac{\partial}{\partial w_{JK}} (0.5 (t_K - y_K)^2) \\ &= (t_K - y_K) \frac{\partial}{\partial w_{JK}} (-y_K) = -(t_K - y_K) \frac{\partial}{\partial w_{JK}} f(y_{in_K}) \\ &= -(t_K - y_K) f'(y_{in_K}) \frac{\partial}{\partial w_{JK}} (y_{in_K}) \quad (\text{by chain rule}) \\ &= -(t_K - y_K) f'(y_{in_K}) z_J\end{aligned}$$



The last equality comes from the fact that only one of the terms in $y_{in_K} = \sum_{j=1}^p w_{jK} z_j$, namely $w_{JK} z_J$ involves w_{JK}

Let $\delta_K = (t_K - y_K) f'(y_{in_K})$. Then $\Delta w_{JK} = \alpha \cdot \left(-\frac{\partial E}{\partial w_{JK}} \right) = \alpha \cdot \delta_K \cdot z_J$

This is the update rule in Step 6 of the algorithm

Or this...

Determining Class Assignment Rules (cont'd)

$$\begin{aligned} p(j^*|t) &= p(j^*, t_L|t) + p(j^*, t_R|t) \\ &= p(j^*|t_L)p(t_L|t) + p(j^*|t_R)p(t_R|t) \\ &= p_L p(j^*|t_L) + p_R p(j^*|t_R) \\ &\leq p_L \max_j p(j|t_L) + p_R \max_j p(j|t_R) \end{aligned}$$

$$\begin{aligned} r(|t) &= 1 - p(j^*|t) \\ &\geq 1 - \left[p_L \max_j p(j|t_L) + p_R \max_j p(j|t_R) \right] \\ &= p_L (1 - \max_j p(j|t_L)) + p_R (1 - \max_j p(j|t_R)) \\ &= p_L r(t_L) + p_R r(t_R) \end{aligned}$$

$$\begin{aligned} R(t) &= p(t)r(t) \\ &\geq p(t)p_L r(t_L) + p(t)p_R r(t_R) \\ &= p(t_L)r(t_L) + p(t_R)r(t_R) \\ &= R(t_L) + R(t_R) \end{aligned}$$



Or this Irish valuer?



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