



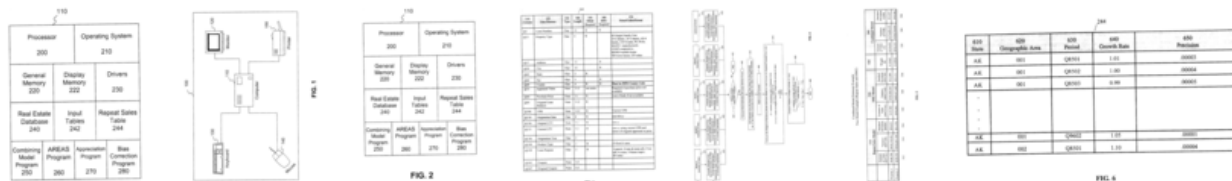
## Method for combining house price forecasts US 6609109 B1

### ABSTRACT

A system computes a plurality of estimates of a real estate entity based on various types and numbers of predictive models. It also selects a limited number of best estimates to be combined. Finally, the system weights each selected estimate according to the calculated precision of that estimate and combines the weighted estimates into a combined weighted estimate.

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<b>Inventors</b>	<a href="#">Michael G. Bradley</a> , <a href="#">James Douglas Gordon</a> , <a href="#">Douglas A. McManus</a>
<b>Original Assignee</b>	<a href="#">Freddie Mac</a>
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### IMAGES (7)



### DESCRIPTION

This application claims the benefit of provisional application Ser. No. 60/005,439 filed Oct. 12, 1995.

#### BACKGROUND OF THE INVENTION

The present invention relates generally to estimating the value of a real estate entity, and more particularly, to combining house price forecasts.

Financial institutions and businesses involved with sales of real estate have long tried to assess the value of real estate entities accurately. For example, financial institutions use estimated value of the real estate entity as one of the key factors in approving loan applications for real estate sales. Relying on the soundness of the estimate, financial institutions accept the risk of lending large sums of money and attach the property as security for the transaction. In this sense, the accuracy of estimated value of the real estate entity is critical.

In addition to the accuracy of the estimate, timeliness is a significant factor. For example, closing on a real estate sales contract may depend on the buyer successfully obtaining a loan within a limited time period. Hence, it is important for lenders to be able to estimate the value of the real estate entity quickly.

Traditionally, real estate personnel performed appraisals manually, but this poses many problems. First, manual appraisals are subjective and vary depending on the appraiser. Second, manual appraisals are expensive. Third, manual appraisals may not be timely due to many unpredictable conditions such as appraiser availability, scheduling conflicts, and weather conditions.

Some have tried to automate the real estate valuation process. For example, Jost et al., U.S. Pat. No. 5,361,201, discloses a neural network-based system for automated real estate valuation. It also discusses other efforts and problems with using statistical models to value real estate entities. In its discussion, Jost

### CLAIMS (22)

What is claimed is:

1. A method of estimating the value of a real estate entity comprising the steps, performed by a data processor, of:

- accessing real estate data and a plurality of predictive models;
- forming a plurality of estimates for the value of the real estate entity based on the predictive models;
- selecting a plurality of best estimates according to a predetermined criteria;
- converting the best estimates into weighted estimates according to the predetermined criteria; and
- allocating the weighted estimates in a combination to obtain a final valuation of the real estate entity.

2. The method of claim 1 including the steps of

- storing historical data corresponding to the plurality of predictive models, and
- bias correcting each estimate using the corresponding historical data.

3. The method of claim 2 wherein the predetermined criteria includes a precision measure.

4. The method of claim 3 wherein the precision measure includes a standard error.

5. The method of claim 1 wherein the selecting step includes the

et al. points out deficiencies of traditional statistical techniques in estimating property values, namely the inability to capture the complexity and the changing trend of the data. It also discusses difficulties involved with selecting a proper sample size for a statistical model to achieve an acceptable stability and reliability of the estimate.

Jost et al., however, did not explore combining predictive models, including statistical models, to predict values of real estate entities. Some studies on general forecasting techniques show combining the results of individual models may produce a significantly better estimate than each individual estimate, but none examined the problems of the real estate market. Moreover, there have been no studies to automate the valuation of real estate by combining the predictive models.

Therefore, it is desirable to increase the accuracy of real estate value forecasts by combining the results of each constituent models.

It is also desirable to provide a timely and reliable estimate of value that is free of human biases and inconsistency.

SUMMARY OF THE INVENTION

The present invention combines house price forecasts to obviate the limitations and disadvantages of the related art.

In accordance with the purpose of the present invention, gas embodied and broadly described, a method of estimating the value of a real estate entity comprises several steps. A data processor accesses real estate data and a plurality of predictive models. The processor forms a plurality of estimates for the value of the real estate entity based on the predictive models and selects a plurality of best estimates according to a predetermined criteria. The processor also converts the best estimates into weighted estimates and combines the weighted estimates into a final estimate.

According to another aspect of the present invention, a system comprises a database, forming means, selecting means, converting means, and a combining means. The database contains real estate data and a plurality of predictive models. The forming means forms a plurality of estimates for the value of the real estate entity based on the predictive models. The selecting means selects a plurality of best estimates. The converting means converts the best estimates into weighted estimates according to the predetermined criteria and, finally, the combining means combines the weighted estimates into a final estimate.

BRIEF DESCRIPTION OF THE DRAWINGS

The accompanying drawings, which are incorporated in and constitute a part of this specification, illustrate the invention and together with the description, serve to explain the principles of the invention.

In the drawings,

FIG. 1 is a diagram of a system according to a preferred embodiment of the present invention;

FIG. 2 is a schematic diagram illustrating in detail components of a computer shown in FIG. 1

FIG. 3 is a table illustrating the input data format for a real estate entity to be estimated;

FIG. 4 is a flowchart illustrating the process of the combined model forecasting according to the preferred embodiment of the present invention;

FIG. 5 is a table showing a combined model estimate example; and

FIG. 6 is a table illustrating the stored data for the repeat sales model.

substeps of

computing the precision of each estimate, and

excluding estimates having an inadequate precision according to a predetermined criteria.

6. The method of claim 5 wherein the predetermined criteria is a threshold level of precision.

7. The method of claim 5 wherein the converting step includes the substeps of

computing a weight for each estimate according to the corresponding level of precision, and

adjusting each estimate with the corresponding weight.

8. The method of claim 1 including the step of

storing a loss function corresponding to a business application.

9. The method of claim 7 wherein the computing step further includes the substep of

using the loss function to compute the weight.

10. The method of claim 1 further including the step of

converting the estimates of the plurality of predictive models into a different format for comparisons.

11. A system for estimating the value of a real estate entity comprising:

a first database containing real estate data and a plurality of predictive models;

means for forming a plurality of estimates for the value of the real estate entity based on the predictive models;

means for selecting a plurality of best estimates according to a predetermined criteria;

means for converting the best estimates into weighted estimates according to the predetermined criteria; and

means for allocating the weighted estimates in a combination to obtain a final valuation of the real estate entity.

12. The system of claim 11 including

a second database storing in the database historical data corresponding to the plurality of predictive models, and

means for bias correcting each estimate using the corresponding historical data.

13. The system of claim 11 wherein the predetermined criteria includes a precision measure.

14. The system of claim 13 wherein the precision measure includes a standard error.

15. The system of claim 10 wherein the selecting means includes

means for computing the precision for each estimate, and

means for excluding estimates having an inadequate precision according to a predetermined criteria.

16. The system of claim 15 wherein the predetermined criteria is a

## DESCRIPTION OF THE PREFERRED EMBODIMENT

Reference will now be made in detail to the present preferred embodiment of the invention, example of which is illustrated in the accompanying drawings.

FIG. 1 is a diagram of a preferred embodiment of the present invention. As shown in FIG. 1, a computer system **100** includes a computer **110** connected to a monitor **120**, a keyboard **130**, a mouse **140**, and a printer **150**. In the preferred embodiment computer **110** is a SUN SPARC **1000**.

FIG. 2 shows a schematic diagram illustrating components of computer **110**. A processor **200** is preferably a SUN SPARC **1000** operating on an operating system **210**, preferably SUN OS 4.06. Computer **110** has RAM composed of a general memory **220** and a display memory **222**. Drivers **230** control various input and output devices connected to computer **110**. Computer **110** also contains real estate database **240**, input tables **242**, and repeat sales table **244**, all of which will be explained below. Additionally, a combining model program **250** controls the overall process of system **100** and oversees an AREAS (Automated Real Estate Analysis System) program **260**, an appreciation program **270**, and a bias correction program **280**; AREAS program **260**, preferably version 4.06, commercially available from HNC, Inc., computes the HNC model.

In the preferred embodiment, computer **110** receives valuation requests and processes the requests. FIG. 3 shows an input table **242** illustrating the format of a sample request. Column **310** indicates the starting position of each field. Column **320** contains data elements about the real estate entity to be valued. As shown in FIG. 3, this data may contain information about loan number, property type, address including city, state, zip, and county, appraisal value, purchase price, original loan balance, origination date of the loan, original loan-to-value, current loan-to-value, loan origination year, loan type, coupon, and original coupon.

The next column, **330**, indicates the type of field, i.e., character or number. Column **340** specifies the length of each field. Columns **350** and **360** indicate whether certain fields are required for weighted repeat sales index (WRSI) and HNC model, respectively. Finally, column **370** contains additional notes, codes, or format information about the fields.

Combining model program **250** may combine any number of predictive models. Additionally, individual forecast models A, B, etc., may be one of any type of predictive models such as statistical models (e.g., repeat sales models, hedonic models), a neural network-based model, or manual evaluation of the real estate entities.

For illustration purposes, combining model program **250** will combine a repeat sales model and the HNC model, but one of ordinary skill in the art can easily modify the number and the types of models to be combined. An example of combining models will be explained below referring to FIGS. 4-6.

FIG. 4 shows a flowchart illustrating the process of the combined model forecasting according to a preferred embodiment of the present invention. Prior to executing each model, combining model program **250** accesses the predictive models and checks whether the models are available for the geographical location of the real estate entity to be valued. Program **250** checks whether real estate database **240** contains data for the geographical location of the real estate entity to be valued. For explanatory purposes, it will be assumed that repeat sales model **510** and the HNC model **520** (see FIG. 5) are available for the real estate entity to be estimated.

The first two stages of combining models are computing the estimate of the real estate entity for each individual model (step **410**) and computing the precision variables (step **420**). These two stages are repeated for each individual model (steps **411-415** and steps **421-425**). For explanatory purposes, Model A is assumed to be repeat sales model **510**, the process of which will now be explained.

To verify that model **510** is available for the geographical location of the real estate entity, program **250** accesses repeat sales table **244** shown in FIG. 6. Table **244** and **620** indicate the state and geographic area of the repeat sales data, respectively. Column **630** shows the quarterly period of the corresponding data. For example, the first entry "Q850" specifies the first quarter in 1985. The next column **640** contains pre-calculated growth rates for the specified state, geographic area, and period. Appreciation program **270** computes this growth rate using historical data from real estate database **240**. The growth rate for the first row, "1.01" indicates an appreciation of one percent. The third row "0.99," on the other hand, indicates a depreciation of one percent. The last column **650** includes a measure of precision to reflect the precision of the computed growth rate.

threshold level of precision.

17. The system of claim 16 wherein the converting means includes

means for computing a weight for each estimate according to the corresponding level of precision, and

means for adjusting each estimate with the corresponding weight.

18. The system of claim 11 wherein the first database contains a loss function corresponding to a business application.

19. The system of claim 18 wherein the computing means further includes

means for using the loss function to compute the weight.

20. The system of claim 11 further includes

means for converting the estimates of the plurality of predictive models into a different format for comparison.

21. The method of claim 1, further including

displaying the final valuation of the real estate entity in a combined model estimate table to a financial institution personnel.

22. The system of claim 11, further including

a display for displaying the final valuation of the real estate entity in a combined model estimate table to a financial institution personnel.

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Prior to executing repeat sales model **510**, combining model program **250** determines whether model **510** is available. To do so, program **250** accesses repeat sales table **244** and verifies that it contains data corresponding to the purchase date of the real estate entity to be valued. If the appropriate data exists, program **250** proceeds to execute the model to compute the estimate. As mentioned above, repeat sales model **250** is assumed to be available in this example.

To obtain the estimate of value for the real estate entity, RSVAL **512** of FIG. 5, repeat sales model **510** accesses repeat sales table **244** for the growth rate of the quarterly period corresponding to the transaction date of the real estate entity and the present date. The estimated value of the real estate entity is then obtained by multiplying the growth rate with the purchase price stored in input table **242**

Thereafter, program **250** computes precision variables for repeat sales model **510** (step **420**). In the preferred embodiment, program **250** uses standard mathematical techniques to compute measures of precision such as the variance, RSVAR **514**, and standard deviation, RSSD **516**, of FIG. 5.

The user also has an option to bias correct the estimate based on historical real estate data. Prior to executing teach model, bias correction program **280** accesses real estate database **240** containing historical data and computes the amount of appropriate bias correction. Specifically, bias correction program **280** computes the average error between past estimates of the real estate entities and actual sale prices of the estimated entity. For example, if program **280** determines that repeat sales model **510** produces estimates that generally tend to be five percent higher than the actual sale price, then the user may invoke the option to bias correct the estimate by five percent. A similar bias correction may be applied to the precision estimates.

If different types of predictive models are combined, program **250** may transform each individual model estimate to facilitate a comparative analysis between the different predictive models. Details on these types of transformations provided in G. S. Maddala, *Econometrics* 314-317 (1977).

HNC model **520** performs a similar process to compute the estimate of the real estate entity. As mentioned above, AREAS program **260**, which is described in detail in documentation accompanying AREAS program **260** version 4.6, executes the HNC model. Referring to FIG. 5, AREAS program **260** obtains estimate of value HNCVAL **522**, variance HNCVAR **524**, and standard deviation HNCSD **526**. If applicable, the HNC model estimate is also bias corrected and transformed for consistency.

Once the individual models complete the estimating process, program **250** checks to see if there are multiple model estimates available (step **430**). If multiple model **432**. If, on the other hand, multiple model estimates are available, program **250** excludes model estimates with low precision (step **440**). In the preferred embodiment of the present invention, a low precision is defined by a comparatively larger standard error than those of other individual models. Other determination factors, such as a threshold standard error, can easily be implemented to determine low precision.

If there are more than one estimate from the same type of model, program **250** can select the best estimate among those model estimates (step **450**). In doing so, program **250** first determines whether any estimates are from the same type of model. If so, program **250** may select from the same model estimates the best estimate. For example, if two hedonic estimates are available, program **250** may select the better of the two hedonic estimates, e.g., the estimate with greater precision.

Next, program **250** selects a loss function (step **460**). Different loss functions may be selected based on the business application and an explicit optimization problem related to the application. In the present example, the loss function is the squared forecast error loss. However, one skilled in the art may easily substitute other loss functions.

Thereafter, program **250** computes the weight for the remaining model estimates based on the selected loss function (step **470**). Implicit in step **470** is the use of a squared forecast error loss. However, one skilled in the art may easily substitute other loss functions. In this example,

$$k = (L' \Omega^{-1}) / (L' \Omega^{-1} * L), \text{ where}$$

$\Omega$  = variance/covariance matrix of forecast errors

$L$  =  $n \times 1$  vector of 1's

$L'$  =  $1 \times n$  vector of 1's (i.e., transpose of  $L$ )

$k$  =  $n \times 1$  vector of weights where  $k_i$  = weight of  $i$ th estimate of value

More specifically, referring to FIG. 5,  $k$  for combining repeat sales model **510** and HNC model **520** may be computed as:  $k = (HNCVAR - W * HNCSD * RSSD) / (HNCVAR + RSVAR - 2 * W * HNCSD * RSSD)$ ,

$$k = \frac{(HNCVAR - W * HNCSD * RSSD)}{(HNCVAR + RSVAR - 2 * W * HNCSD * RSSD)}$$

where

W=correlation of the forecast errors of the two models.

Finally, combining model program **250** computes a combined weighted average estimate of value, C. Program **250** first multiplies k by each remaining model estimate to convert the estimates into weighted estimates. In general, program, **250** combines all the weighted estimates to obtain C as follows (step **480**):

$$C=k_1 f_1 + k_2 f_2 + \dots + k_n f_n, \text{ where}$$

$k_{1,2,3 \dots n}$ =weight of corresponding model estimate

$f_{1,2,3 \dots n}$ =model estimates

Referring again to FIG. 5, in the present example combining repeat sales model **510** and HNC model **520**, the following formulas may be utilized to compute the combined estimate of value, (C **542**), variance (CVAR **544**), and standard deviation (CSD **546**):

$$C=k*RSVAL+(1-k)HNCVAL$$

$$CVAR=k^2 *RSVAR+((1-k)^2)*HNCVAR+2*k*(1-k)*W*RSSD*HNCSD$$

$$CSD=CVAR^{(0.5)}$$

Program **250** completes successfully upon obtaining the combined weighted estimate C (step **490**).

## CONCLUSION

The present invention increases the accuracy of estimates of value for real estate entities by combining the results of each constituent model in an appropriate manner. Additionally, it also enables a timely and reliable estimate that is free of human biases and inconsistency inherent in manual appraisals.

It will be apparent to those skilled in the art that various modifications and variations can be made in the computer network of the present invention and in construction of this computer network without departing from the scope or spirit of the invention.

Other embodiments of the invention will be apparent to those skilled in the art from consideration of the specification and practice of the invention disclosed herein. It is intended that the specification and examples be considered as exemplary only, with the true scope and spirit of the invention indicated by the following claims.

## PATENT CITATIONS

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<a href="#">US5664115</a> *	Jun 7, 1995	Sep 2, 1997	Fraser; Richard	Interactive computer system to match buyers and sellers of real estate, businesses and other property using the internet
<a href="#">US5926792</a> *	Sep 18, 1998	Jul 20, 1999	Bancorp Services, Inc.	Computer system
<a href="#">US5963919</a> *	Dec 23, 1996	Oct 5, 1999	Northern Telecom Limited	Inventory management strategy evaluation system and method
<a href="#">JP40452198A</a> *				Title not available
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<a href="#">US7249146</a> *	Sep 6, 2001	Jul 24, 2007	Shimon Brecher	Computerized system and method for acquiring and analyzing data pertaining to real estate
<a href="#">US7305328</a> *	Oct 29, 2004	Dec 4, 2007	Fannie Mae	Method and apparatus for predicting and reporting a real estate value based on a weighted average of predicted values
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<a href="#">US7653592</a>	Dec 30, 2005	Jan 26, 2010	Fannie Mae	System and method for processing a loan
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<a href="#">US7693765</a>	Nov 30, 2005	Apr 6, 2010	Michael Dell Orfano	System and method for creating electronic real estate registration
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<a href="#">US20120185406 *</a>	Jan 18, 2011	Jul 19, 2012	International Business Machines Corporation	FAST AND ACCURATE METHOD FOR ESTIMATING PORTFOLIO CVaR RISK
<a href="#">US20130080340 *</a>	Sep 23, 2011	Mar 28, 2013	Elif Onmus-Baykal	Indexing and adjusting for property condition in an automated valuation model
<a href="#">US20130103457 *</a>	Oct 16, 2012	Apr 25, 2013	Kevin Leon Marshall	Method and system for providing a home data index model
<a href="#">US20130103459 *</a>	Oct 16, 2012	Apr 25, 2013	Kevin Leon Marshall	Method and system for providing a home data index model
<a href="#">WO2008055226A2 *</a>	Oct 31, 2007	May 8, 2008	Kathy Coon	Appraisal evaluation and scoring system and method

\* Cited by examiner

## CLASSIFICATIONS

U.S. Classification	<a href="#">705/35</a>
International Classification	<a href="#">G06Q40/00</a>
Cooperative Classification	<a href="#">G06Q40/02</a> , <a href="#">G06Q40/00</a>
European Classification	G06Q40/02, G06Q40/00

## LEGAL EVENTS

Date	Code	Event	Description
Dec 12, 2014	FPAY	Fee payment	<b>Year of fee payment:</b> 12
Jan 10, 2011	FPAY	Fee payment	<b>Year of fee payment:</b> 8
Feb 20, 2007	FPAY	Fee payment	<b>Year of fee payment:</b> 4
Dec 23, 2003	CC	Certificate of correction	
Oct 11, 1996	AS	Assignment	<b>Owner name:</b> MAC, FREDDIE, VIRGINIA <b>Free format text:</b> ASSIGNMENT OF ASSIGNORS INTEREST;ASSIGNORS:BRADLEY, MICHAEL G.;MCMANUS, DOUGLAS A.;GORDON, J. DOUGLAS;REEL/FRAME:008269/0925 <b>Effective date:</b> 19961010