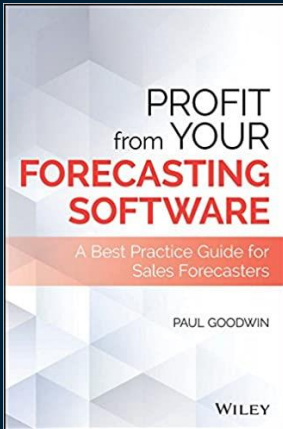


Forecasting Book Club



**Paul Goodwin's**  
***Profit From Your Forecasting Software***  
***Chapters 10-11***

Michael Gilliland  
SAS

1

Forecasting Book Club

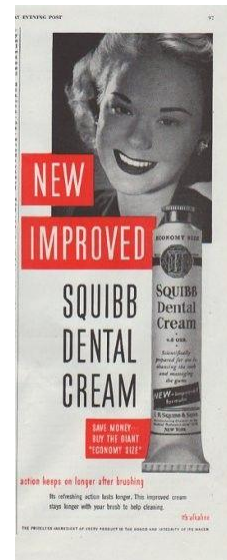
**Chapter 10**  
**New Product Forecasting**

2

## Forecasting Book Club

## ■ Many NPF Situations

- Refinement of existing products
- New markets for existing products
- New to the world products



3

## Forecasting Book Club

## ■ Many NPF Approaches

- Executive Opinion/Decree (evangelical forecasting)
- Sale Force Rollup (bottoms-up forecast)



**Gilliland, M. (2014). Role of the sales force in forecasting. *Foresight: The International Journal of Applied Forecasting* 35: 8-13.**

4

## Forecasting Book Club



## ■ Many NPF Approaches

- Executive Opinion/Decree (evangelical forecasting)
- Sale Force Rollup (bottoms-up forecast)
- Delphi Method (anonymous structured process)
- Prediction Market (anonymous wagering)
- Analogy (utilize history of “like” product)
- Diffusion Modeling (spread of product adoption)

*Goodwin focuses on just these last two approaches*

5

## Forecasting Book Club

## ■ Use of Judgment in Forecasting: Disadvantages

- Cognitive biases
  - Overinfluence of recent events
  - Optimism bias
  - Advocacy bias
  - *Groupthink*
- Biased feedback

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## Forecasting Book Club

**■ Use of Judgment in Forecasting: Advantages**

- Compensates for lack of historical information
  - Data unavailable (new products)
  - Too difficult/costly/time consuming to collect (e.g. test market)
- Compensates for the lack of future information
  - Too difficult/costly/time consuming to collect (market research)
  - Difficult to quantify (fads and fashions)

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## Forecasting Book Club

**Forecasting by Structured Analogy**

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## Forecasting Book Club

**■ Use of analogies is a common practice**

- House Pricing (real estate “comps”)
- New Product Forecasting (look-alike products / similar characteristics)

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## Forecasting Book Club

**■ Structured Analogy Approach**

- Combines analogies with structured judgment
  - Guided statistical analysis that incorporates human judgment
  - Attempts to remove judgmental bias by providing a historical context for each decision
  - Attempts to validate and test the decisions
  - Choice of analogy is driven by a statistical process

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## Forecasting Book Club

## ■ Structured Analogy Data Requirements

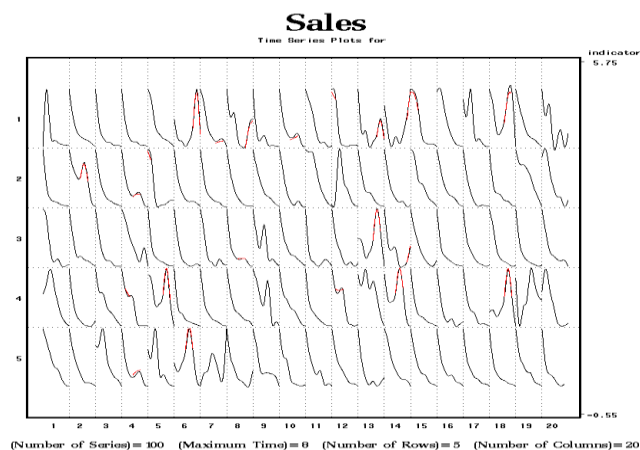
- Associated with each product (new and historical) are **attributes** that categorize the product.
  - Product type (toy, music, clothing, shirts, etc.)
  - Season of introduction (summer-item, winter-item, etc.)
  - Financial (price, competitor-price, etc.)
  - Demographic (gender, age, ethnic, etc.)
  - Physical characteristics (style, color, size, etc.)
  - ... and many others
- Some product attributes cannot be easily quantified.
  - Aesthetic appeal, Beauty, Fads, etc.

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## Forecasting Book Club

## ■ Structured Analogy data requirements

- Historical data on all past new product introductions



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## Forecasting Book Club

## Structured Analogy – NPF Process

- **Query Step** - finds a set of *candidate products* that have similar *attributes* to the *new product*.
- **Filter Step** - removes *inappropriate* or *outlier products* from the set of *candidate products* to form the set *surrogate products*.
  - Select most appropriate surrogates given several choices of filters and/or clusters of candidate products
- **Model Step** - extracts statistical model features from the set of *surrogate products*.
  - Select most appropriate model given several choices
- **Forecast & Override Step** - uses the extracted statistical model features to forecast the *new product* and compensate for timing.
  - Manually override if necessary

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## Forecasting Book Club

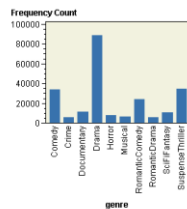
Step 1: Select dataset and attribute variables

Dataset  
DVD\_TITLES

Attribute 1

Box\_Office  
genre  
imdb\_title\_Cd  
mpaa  
runs  
title\_description

Obs	genre	Count	Percent
1	Drama	89093	33.9330
2	SuspenseThriller	34913	13.2674
3	Comedy	34247	13.0437
4	RomanticComedy	24209	9.2510
5	Documentary	11721	4.4642
6	SciFFantasy	11378	4.3336
7	Horror	8201	3.1540
8	Musical	7040	2.6813
9	RomanticDrama	6430	2.4490
10	Crime	6402	2.4383



GENRE  
ActionAdventure  
Animation  
ChildrenFamily  
Comedy  
Crime  
DarkComedy  
Documentary  
Drama  
Horror  
Musical  
Mystery  
RomanticComedy  
RomanticDrama  
SciFFantasy

## Attribute Selection

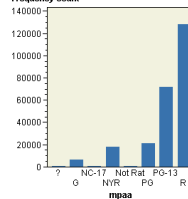
- DVD title to be released
- Example:
  - Select genre (horror) and mpaa rating (R) as screening attributes

Attribute 2

Box\_Office  
genre  
imdb\_title\_Cd  
mpaa  
runs  
title\_description

Obs	mpaa	Count	Percent
1	R	129544	51.9599
2	PG-13	72016	29.1102
3	PG	21119	8.5367
4	NYR	10309	7.4008
5		15185	
6	G	6410	2.5943
7	Not Rat	800	0.3206
8	NC-17	176	0.0711
9	?	1	0.0004

Frequency Count



MPAA

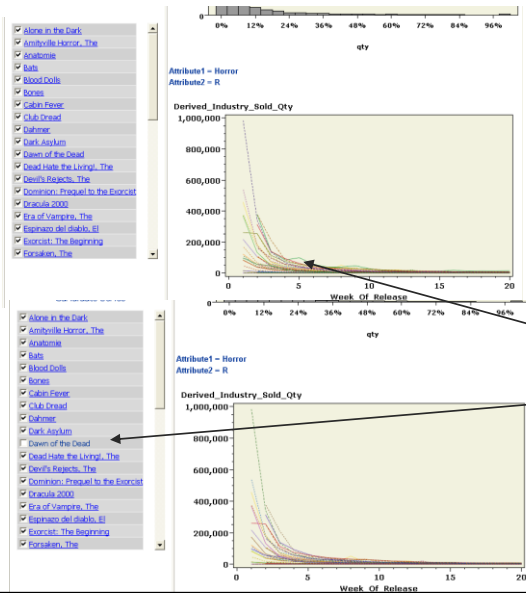
?  
G  
NC-17  
NYR  
Not Rat  
PG  
PG-13  
R

Note: The selected attributes can coincide with marketing plans, sales feedback, etc...

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## Forecasting Book Club

## Step 2: Query step

Query Step

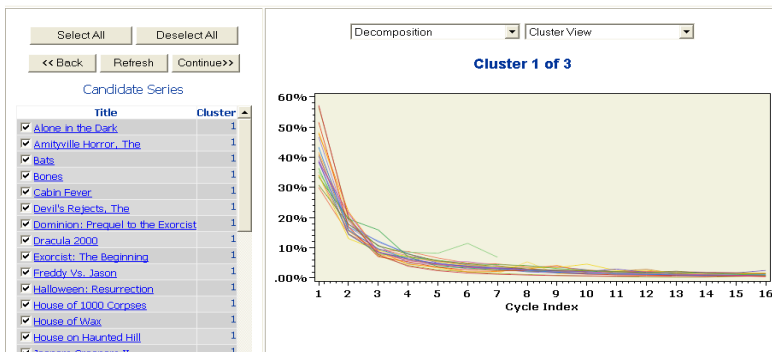
- Explores candidate titles with similar attributes
- Allows for outlier filtration
- Forms a set of surrogate titles

Decided that "Dawn of the Dead" was an outlier  
– removed from candidate list

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## Forecasting Book Club

## Step 3: Filter step



Decided that "Cluster 1" reflected desired pattern

Filter Step

- Explores clustered titles
- Select surrogate cluster based on anticipated pattern
- Prepares for statistical modeling

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## Forecasting Book Club

## Candidate Series

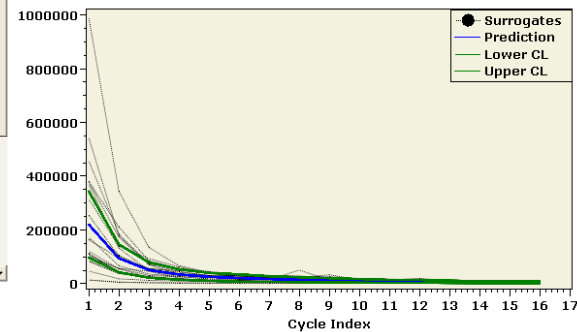
Title	MAPE
<input checked="" type="checkbox"/> Alone in the Dark	90.23
<input checked="" type="checkbox"/> Amityville Horror, The	32.29
<input checked="" type="checkbox"/> Bats	905.09
<input checked="" type="checkbox"/> Bones	173.51
<input checked="" type="checkbox"/> Cabin Fever	24.01
<input checked="" type="checkbox"/> Devil's Rejects, The	48.04
<input checked="" type="checkbox"/> Dominion: Prequel to the Exorcist	1598.29
<input checked="" type="checkbox"/> Dracula 2000	88.02
<input checked="" type="checkbox"/> Exorcist: The Beginning	121.30
<input checked="" type="checkbox"/> Freddy Vs. Jason	19.82
<input checked="" type="checkbox"/> Halloween: Resurrection	99.96
<input checked="" type="checkbox"/> House of 1000 Corpses	30.08
<input checked="" type="checkbox"/> House of Wax	28.87

## Step 4: Model step

Loess Model

Drop down menu has access to variety of models:  
Diffusion, Mixed, Smoothing, Bayesian, etc...

## Cycle Series Model Predictions



## Model Step

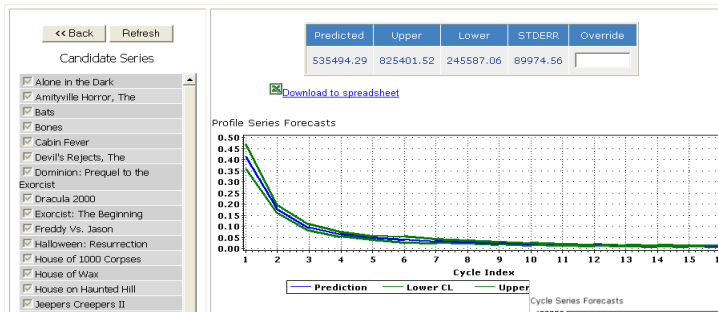
- Generate best statistical model that fits the surrogates
  - Multiple models, weighting criteria, etc...
- Evaluate error for each surrogate
- Prepare for forecast modification

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## Forecasting Book Club

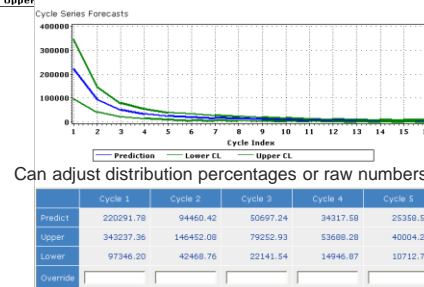


## Step 5: Override step



## Forecast &amp; Override Step

- Generate the profile model for the new product
- Provide cycle view, percentage view and absolute quantity view
- Allow for overrides
- Export to other applications



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## Forecasting Book Club

**■ Structured Analogy approach -- Summary**

- Appropriate in many situations
- Statistical analysis and data visualization are incorporated with human judgment
  - Construction of the NPF model is not fully automated
- Can be used to evaluate the risk / uncertainty in forecasts from other sources (e.g. product manager)

*The software does the grunt work, making the NPF process as automated, efficient, and objective as possible*

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## Forecasting Book Club

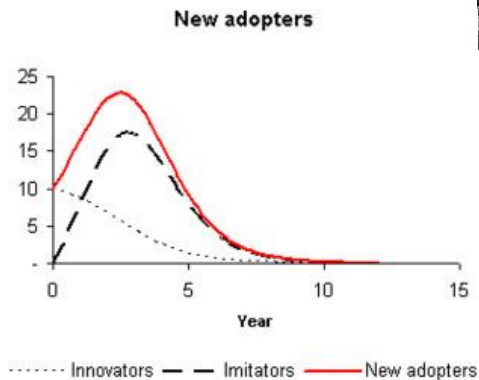
**Diffusion Modeling**

20

## Forecasting Book Club

## ■ Bass Diffusion Model

- Innovators
- Imitators



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## Forecasting Book Club

## ■ Bass Diffusion Model - Parameters

- Market size (number of potential adopters)
- Coefficient of innovation  $p$  = tendency of people to buy because of advertising/media exposure
- Coefficient of imitation  $q$  = tendency of people to buy based on word-of-mouth from earlier buyers

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## Forecasting Book Club

## ■ Bass Diffusion Model - Formula

Number of adopters in a period =

$[p \times \text{number of potential customers who have yet to adopt at start of period}]$

+  $[q \times \text{proportion of market who have already adopted at start of period}]$

$\times \text{number of potential customers who have yet to adopt at start of period}]$

23

## Forecasting Book Club

## ■ Bass Diffusion Model – Applying the model

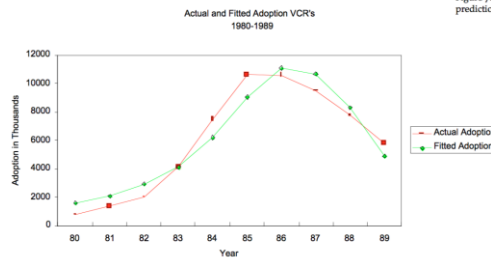
- Estimate market size,  $p$  and  $q$  from available periods of history
  - If no history, estimate  $p$  and  $q$  from analogous products
  - Can estimate market size from consumer intention surveys or other means
- Forecast future periods using these parameters
- Check forecast for reasonableness

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## Forecasting Book Club

## ■ Bass Diffusion Model - Limitations

- Forecasts adoption (new buyers) – not total sales
- Estimate of  $p$  and  $q$  is critical
- Assumes market size and  $p$  and  $q$  remain unchanged
- Other types of S-curves may better fit cumulative adopters
- Not all new products have a single peak of adoption



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## Forecasting Book Club

## Worst Practices in New Product Forecasting

Gilliland, M. Worst Practices in New Product Forecasting.  
*Journal of Business Forecasting* (Winter 2012-13).

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## Forecasting Book Club

- **Unrealistic accuracy expectations**
- **Reverse engineering the forecast**
- **Cherry-picking analogies**
- **Insisting on a wiggle**
- **The Hold-and-Roll**
- **Ignoring the product portfolio**

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## Forecasting Book Club

- **New Product Forecasting - Summary**
  - Have low expectations for accuracy
  - Utilize multiple methods (benefits of combining)
  - Use structured judgment
  - Use Bass model when appropriate – but with caution

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Forecasting Book Club

## Chapter 11

### Summary: A Best Practice Blueprint for Using Your Software

29

Forecasting Book Club

- **Desirable characteristics of forecasting software**
  - Obviously:
    - Ease of use
    - Speed (of interface and computations)
    - Accuracy of calculation
    - Wide range of methods
    - Able to interface with other systems
    - Training and support

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## Forecasting Book Club

## ■ Desirable characteristics of forecasting software

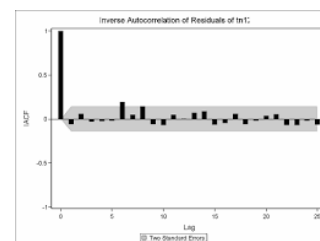
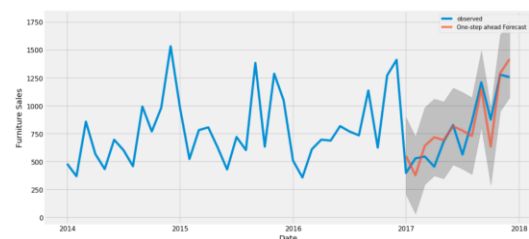
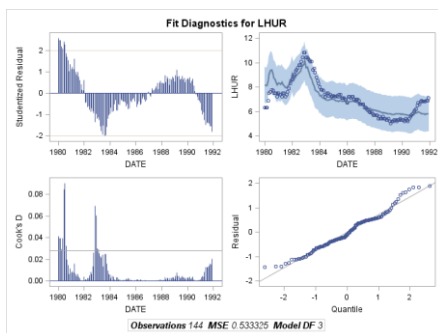
- Data preparation and management capabilities
  - Data quality / completeness / master files
  - Missing values / outliers
  - Data transformations

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## Forecasting Book Club

## ■ Desirable characteristics of forecasting software

- Data preparation
- Graphical displays



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## Forecasting Book Club

## ■ Desirable characteristics of forecasting software

- Data preparation
- Graphical displays
- Automatic modeling / method selection
  - Penalize more complex methods using information criteria
  - Clear demarcation of fit (training) period and hold-out (test)
  - Multiple accuracy measures available
  - Rolling origin evaluation

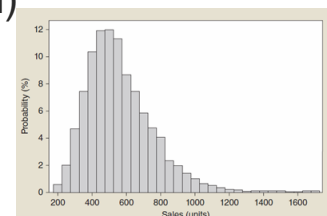
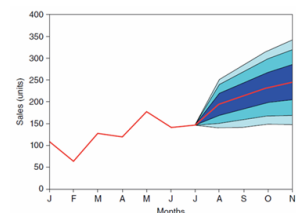
33

## Forecasting Book Club

## ■ Desirable characteristics of forecasting software

- Data preparation
- Graphical displays
- Automatic modeling / method selection
- Combinations / events / predictor variables
- Hierarchies (organizational and temporal)
- Probabilistic forecasting

**Goodwin, P. (2014). *Getting Real About Uncertainty. Foresight* 33: 4-7.**



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## Forecasting Book Club

## ■ Desirable characteristics of forecasting software

- Data preparation
- Graphical displays
- Automatic modeling / method selection
- Combinations / events / predictor variables
- Hierarchies (organizational and temporal)
- Probabilistic forecasting
- Support for judgment

Baker, J. (2021). Maximizing FVA Through Machine Learning and Nudges. *Foresight* 60 (forthcoming).

35

## Forecasting Book Club

## ■ Desirable characteristics of forecasting software

- Data preparation
- Graphical displays
- Automatic modeling / method selection
- Combinations / events / predictor variables
- Hierarchies (organizational and temporal)
- Probabilistic forecasting
- Support for judgment
- Presentation of forecasts

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## Implementing a Forecasting Solution

Gilliland, M. (2010). *The Business Forecasting Deal*.

37

### ■ Why do implementations fail?

- Perceived failure
  - Doesn't deliver what was promised /expected
  - Due to poorly managed expectations
- True failure
  - Software never installed
  - Software installed but never used
  - Software installed and used – but terrible at forecasting!

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## Forecasting Book Club

## ■ Preproject Assessment

- Realistic evaluation of IT infrastructure and capabilities
  - Basic data elements available?
  - Sufficient history available?
  - Data management practices in place?
  - Skillset of IT resources vs. contractors

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## Forecasting Book Club

## ■ Request for Information / Proposal

- Long list of “requirements”
  - Focused on minute and irrelevant details
  - Exposes customer’s lack of understanding
    - E.g. “Software uses artificial intelligence? Y/N”
- Selection process often “fixed” for favored vendor
- Alternative RFI/RFP:
  - What is your understanding of our business problem?
  - How do you propose to solve it?

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## Forecasting Book Club

**■ Evaluating Software Vendors**

- Beware of the dirty tricks of selling
- Beware of claims about ROI
- Be skeptical of references

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## Forecasting Book Club

**■ Warning Signs of Failure**

- Assessment questions:
  - Are sufficient IT and business resources engaged?
  - Is upper management committed?
- Is the project staying on plan (time and budget)?
  - Set early hurdles for go/no-go decisions

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## Forecasting Book Club

## Paul Goodwin's Blueprint for Best Practice

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## Forecasting Book Club

### ■ The Blueprint

1. Restrict judgmental interventions
2. Automate the forecasting process
3. Feed the computer with data
4. Acknowledge uncertainty / use Pis
5. Judge a method by quality of its forecasts, not fit to history

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## Forecasting Book Club

**■ The Blueprint**

6. Simplicity may beat complexity
7. Measure the bias
8. Forecast at appropriate level of aggregation
9. Combine forecasts
10. Correlation does not prove causation

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## Forecasting Book Club

**Questions**

Contact:

Mike Gilliland  
Product Marketing Manager  
SAS  
[mike.gilliland@sas.com](mailto:mike.gilliland@sas.com)

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