

A Structural Model of Salesforce Compensation Dynamics

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Introduction

- Compensation is a fundamental driver of productive activity
- The literature on compensation spans all aspects of business including accounting, finance, strategy, HR, operations, organizational behavior and even *marketing*.
- The fundamental question has always been to ask how might we best compensate agents to maximize output.

Introduction

- There is a long and storied literature on compensation in each of these areas.
- The fundamental building blocks are however essentially the same. . .

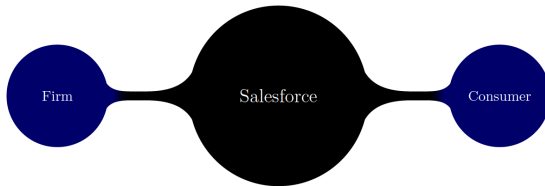
Effort \rightarrow *Output* \rightarrow *Compensation*

- The idea is then to design compensation so that it maximizes the incentives for effort provision.

Introduction

Salesforce Management

- Personal selling via salesforce important component of economy
 - US economy spent \$800 billion on salesforces, $\sim 3\times$ amount spent on advertising, up-to 40% of sales
 - 12% of workforce (Zoltners et al. 2008)
 - Dominant form of marketing in B2B channels
- Salesforce management relates to understanding and improving the interplay between three key actors.



Introduction

The Structural Approach

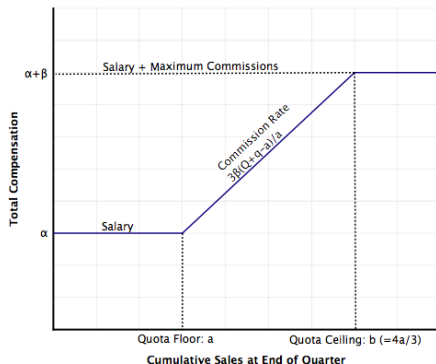
- (My own) Definition of structure:
 - $\text{Structure} = \text{Theory} + \text{Data}$
- Patterns in the data should be explained via models of agent behavior
 - i.e. we need micro-foundations for aggregate, macro patterns (Lucas)
- The basic idea is to use structural methods to investigate salesforce management issues.
- In essence, this brings together the two facets of of the saleforce literature...
 - Lodish, Zoltners, Weinberg, Mantrala, ...
 - Farley, Srinivasan, Lal, Coughlan, Rao, ...

Agenda

- Introduction
- Model Framework
- Data and Model-Free Evidence
- Econometric Implementation
- Results
- Implementing a New Plan
- Conclusions

Model Framework

Compensation Scheme in Data



- $\text{Compensation} = \text{Salary} + \text{Commission} \times \mathbb{I}(\text{Quota} < \text{Sales} < \text{Ceiling})$
 - No bonus, Ceiling is a fixed fraction of quota
 - Quota is reset quarterly and is adjusted based on past

Model Framework

Compensation Scheme, States, Payoffs

■ Compensation Scheme

$$W_t = \alpha + \beta \mathbf{I}(I_t = N) \left[\begin{aligned} &\left(\frac{Q_t + q_t - a_t}{b_t - a_t} \right) \mathbf{I}(a_t \leq Q_t + q_t \leq b_t) \\ &+ \mathbf{I}(Q_t + q_t > b_t) \end{aligned} \right]$$

■ States

- Q_t , cumulative sales achieved in quarter
- a_t , current quota
- I_t , months since the beginning of the quarter

Model Framework

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■ States

- Q_t , cumulative sales achieved in quarter
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 - I_t , months since the beginning of the quarter
- Sales are a stochastic function of effort, which is a function of the agent's state

$$q_t = g(e_t(\mathbf{s}_t), z) + \varepsilon_t$$

Model Framework

Compensation Scheme, States, Payoffs

■ Compensation Scheme

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■ Current Payoff

$$u_t = \left[\begin{array}{ll} E[W_t] - r \text{var}[W_t] - C(e_t; d) & \text{if } \chi_t = 1 \\ 0 & \text{if } \chi_t = 0 \end{array} \right]$$

Model Framework

State Transitions

■ Cumulative Sales

$$Q_{t+1} = \begin{cases} Q_t + q_t & \text{if } I_t < N \\ 0 & \text{if } I_t = N \end{cases}$$

Model Framework

State Transitions

■ Cumulative Sales

$$Q_{t+1} = \begin{cases} Q_t + q_t & \text{if } I_t < N \\ 0 & \text{if } I_t = N \end{cases}$$

■ Quotas (“ratcheting”)

$$a_{t+1} = \begin{cases} a_t & \text{if } I_t < N \\ \sum_{k=1}^K \theta_k \Gamma(a_t, Q_t + q_t) + v_{t+1} & \text{if } I_t = N \end{cases}$$

Model Framework

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■ Months of the quarter

$$I_{t+1} = \begin{cases} I_t + 1 & \text{if } I_t < N \\ 1 & \text{if } I_t = N \end{cases}$$

Value Function

Early in the quota cycle

$$V(Q_t, a_t, N, \chi_t) = \max_{\chi_{t+1}, e} \left\{ u(Q_t, a_t, l_t, \chi_t, e) + \rho \int_{\varepsilon} V(Q_{t+1} = Q(Q_t, q(\varepsilon_t, e)), a_{t+1} = a_t, l_t + 1, \chi_{t+1}) \times f(\varepsilon_t) d\varepsilon_t \right\}$$

Value Function

End of the quota cycle

$$V(Q_t, a_t, N, \chi_t) =$$

$$\max_{\chi_{t+1}, e} \left\{ \begin{aligned} &u(Q_t, a_t, N, \chi_t, e) + \\ &+ \rho \int_v \int_\varepsilon V(Q_{t+1} = 0, a_{t+1} = a(Q_t, q(\varepsilon_t, e), a_t, v_{t+1}), 1, \chi_{t+1}) \\ &\quad \times f(\varepsilon_t) \phi(v_{t+1}) d\varepsilon_t dv_{t+1} \end{aligned} \right.$$

- Optimal effort solves

$$e(\mathbf{s}_t; \Omega, \Psi) = \arg \max_{e > 0} \{V(\mathbf{s}_t; \Omega, \Psi)\}$$

- Empirical Approach

- Estimate $\hat{\Omega}$ given Ψ and current DGP
- Simulate $e(\mathbf{s}_t; \hat{\Omega}, \Psi = \Psi_{new})$ under counterfactual

Our Data are Unusually Rich

Cross-sectional and Temporal Variation for Each Agent

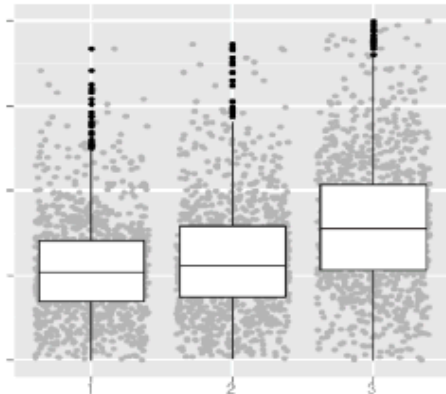
- Data come from a salesforce/division of a Fortune 500 firm
- Sales of contact lens prescribed by physician
- Spans 38 months (2005-2008)
- Sales for each salesperson at month/client level
 - Salesforce has 87 salespeople
 - on average ~160 clients per salesperson!
 - Gives us ~6080 obs **per salesperson** and ~528,960 obs total
- Complete compensation details for each salesperson
 - Quotas for each quarter
 - Commissions and salaries paid

Descriptive Statistics of Data

Variable	Mean	SD
Salary	\$67,632	\$8,585
Incentive Proportion at Quota	0.23	0.02
Age	43.23	10.03
Tenure	9.08	8.42
Num_Clients	162.20	19.09
Quota	\$397,020	\$95,680
Cum:Sales (end of quarter)	\$374,755	\$89,947
% Δ Quota (when +)	10.01%	12.48%
% Δ Quota (when -)	-5.53%	10.15%
Monthly Sales	\$138,149	\$38,319
Cum:Sales (beg: of month)	\$114,344	\$98,594
Distance to Quota (beg: of month)	\$278,858	\$121,594

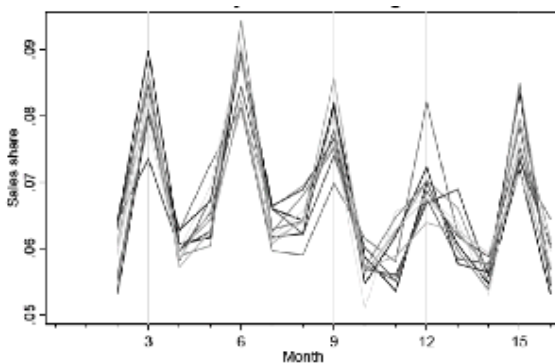
Effort Timing by Agents

Model free evidence - Sales within the quarter



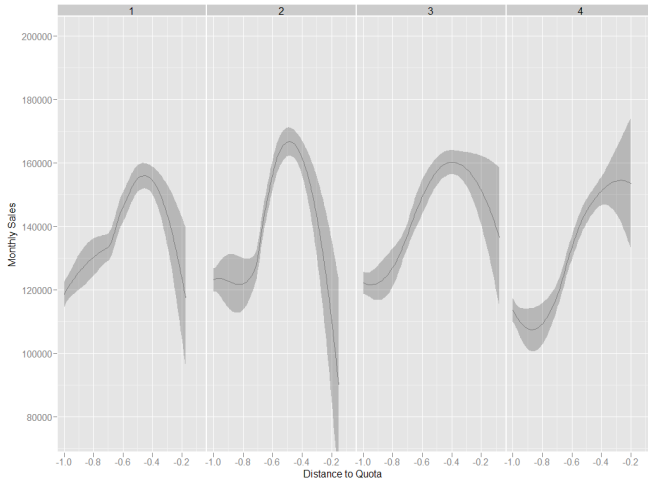
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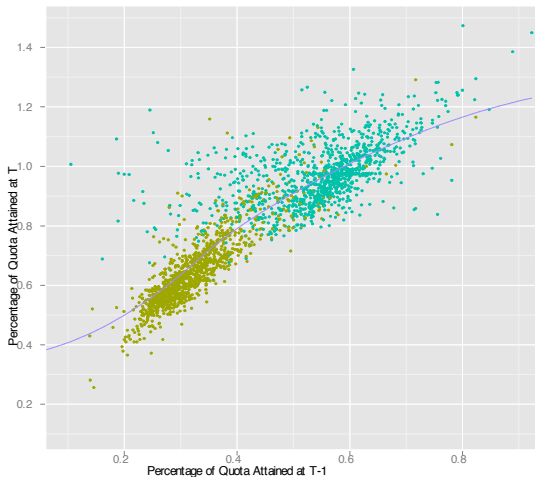
Effort Timing by Agents

Model free evidence - Sales as a function of distance to quota



Effort Timing by Agents

Model Free Evidence - Near Quota Effort



Effort Timing by Agents

Model Free Evidence - Individual Salespeople

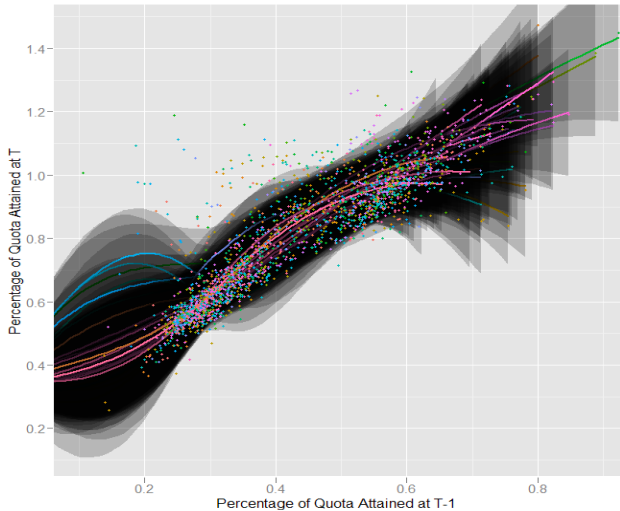
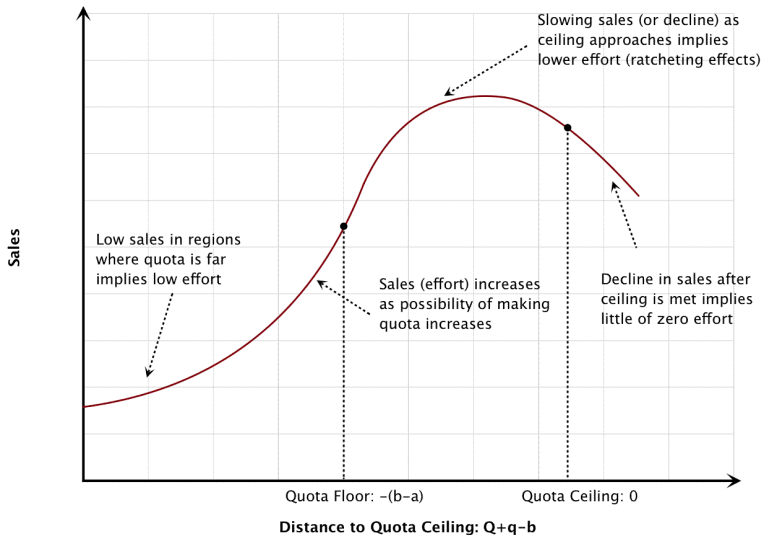


Figure:

Misra

Structural Compensation Dynamics

Identification of Effort Policy



Ruling out alternative explanations

Seasonality and buyer stockpiling

- The identification of effort relies critically on mapping patterns in sales to state variables
 - We need to rule out alternate explanations for the patterns in sales
- No underlying seasonality in product demand from consumers
 - Underlying demand is generated by health outcomes
 - Consumers (or physicians) have no particular reasons to delay or accelerate demand
- Physician stockpiling limited due to profusion of products (15,000 SKU-s in product line)
 - Demand cannot be stored
 - No systematic pattern of sales being lower after months of high demand (at physician level)
 - Data based evidence in paper
- Finally, sales under new plan (with constant incentives) are flat

Econometric Implementation

Estimation Approach Details

- Our estimation approach uses a two-step approach (Bajari, Benkard and Levin 2007)
 - Non-parametrically estimate policy functions in first-stage
 - Estimate parameters by minimizing violations of dynamic optimality
- Agent-level data of unusually long duration and cross-section
 - Accommodate non-parametrically, unobserved agent heterogeneity
- Estimate policy functions & parameters agent by agent
- Important Econometric Challenge
 - Unobservability of effort (pervasive in principal-agent settings)

Econometric Implementation

Nonparametric Estimation of the Effort Policy Function

- Index clients by j
- Recall that the sales production function is

$$q_{jt} = h_j(z_j) + e(\mathbf{s}_t) + \varepsilon_{jt}$$

and z_j are time invariant client characteristics

- Project effort policy on flexible orthogonal polynomial basis functions of state variables, $\boldsymbol{\vartheta}(\mathbf{s}_t)$,

$$q_{jt} = \delta' \mathbf{z}_j + \lambda' \boldsymbol{\vartheta}(\mathbf{s}_t) + \varepsilon_{jt}$$

- Non-Linear Least Squares estimation provides, for **each** agent,
 - Effort policy function, $\hat{e}_t = \hat{\lambda}' \boldsymbol{\vartheta}(\mathbf{s}_t)$, and,
 - Empirical distribution of month-specific errors,
$$\hat{\varepsilon}_t = \sum_j \left(q_{jt} - \left(\hat{\delta}' \mathbf{z}_j + \hat{e}(\mathbf{s}_t) \right) \right)$$

Intuition for Identification of Effort

Two steps

- Step 1: Estimate period specific productivity of sales-calls

$$q_{jt} = \delta' \mathbf{z}_j + \gamma_t + \varepsilon_{jt}$$

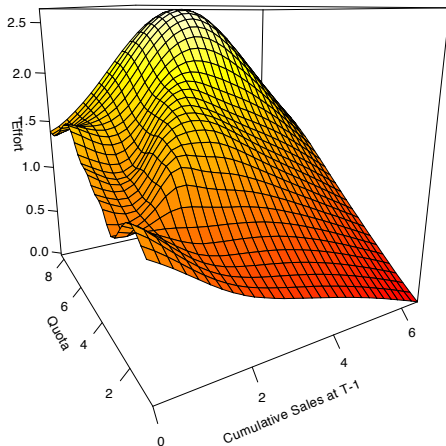
- Step 2: Project productivity on flexible function of the state

$$\hat{\gamma}_t = \lambda' \boldsymbol{\vartheta}(\mathbf{s}_t)$$

- Identification restrictions
 - Effort is not client (j) specific
 - Optimal effort policy is a deterministic function of the observed state variables

Estimation Results

Estimated Effort Policy



Estimation Results

Examples of Individual Effort Policy Estimates

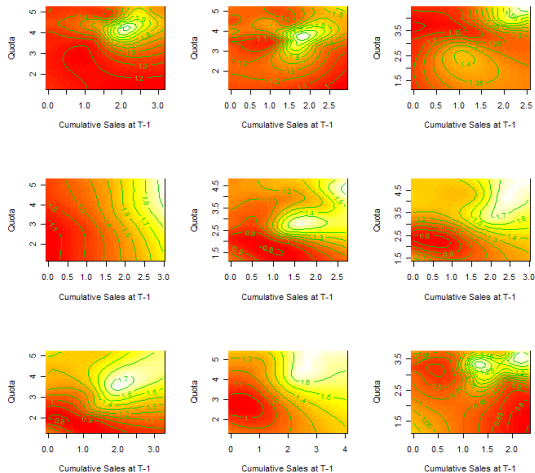


Figure:

Estimation of risk aversion and cost of effort

Minimize violations of agent optimality

Let s_0 be an initial state for an agent, (r^*, d^*) be the true parameters, and e^* the optimal effort policy at the true parameters. Then, (r^*, d^*) must satisfy,

1 Individual Rationality (IR): $V(s_0; e^*, r^*, d^*) \geq 0$

2 Incentive Compatibility (IC):

$$V(s_0; e^*, r^*, d^*) \geq V(s_0; e', r^*, d^*)$$

Estimate (r, d) by minimizing violations of IR and IC

Let $e'(s) = e^*(s) + \varepsilon$.

Then can compute $V(s_0; e^*, \theta) = E_{e^*|s_0} \sum_{t=0}^{\infty} \beta^t E_{\varepsilon} [W(s, e^*(s))]$
and $V(s_0; e', \theta)$.

- Define the following two moment conditions

$$g_1(s_0; \theta) = \min(V(s_0; e^*, \theta), 0)$$

$$g_2(s_0, e'; \theta) = \min(V(s_0; e^*, \theta) - V(s_0; e', \theta), 0)$$

$$\text{and } g(s_0, e'; \theta) = [g_1 \ g_2]'$$

Estimation

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and $g(s_0, e'; \theta) = [g_1 \ g_2]'$

- Define an objective function

$$Q(\theta) = \int [g(s_0, e'; \theta)]' \Lambda [g(s_0, e'; \theta)] dH(s_0, e')$$

Estimation

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- True parameter vector ($\theta = \theta^*$) must satisfy

$$Q(\theta^*) = \min_{\theta} (Q(\theta)) = 0$$

Estimation

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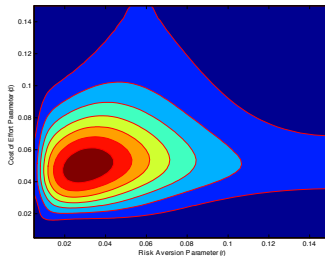
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- True parameter vector ($\theta = \theta^*$) must satisfy

$$Q(\theta^*) = \min_{\theta} (Q(\theta)) = 0$$

- Solve for (r, d) agent by agent

Parameter Estimates



- Mean monthly average risk premium ($\mathcal{E}_{\text{agents}} \left[\frac{r}{T} \mathbb{V}(W) \right]$)
 $\approx \$340$
- Mean monthly average cost of effort ($\mathcal{E}_{\text{agents}} \left[\frac{1}{T} d\mathbb{C}(e) \right]$) \approx
\$1500
- Compare to monthly (max) payout of $\approx \$7,250$

Evaluating the compensation scheme

Comparisons with counterfactual schemes

- For a given compensation plan $\wp(q(e))$ the firms discounted payoffs are

$$\hat{\Pi}_{\wp} = \frac{1}{T \times NS} \sum_{s=1}^{NS} \sum_{\tau=0}^T \beta^{\tau} [q(e_{\wp}; \Psi^s) - \wp(q(e_{\wp}); \Psi^s)]$$

- where Ψ^s is a draw from $\mathcal{G}_{\varepsilon}(\varepsilon_{\tau}) \times \mathcal{F}(\mu, r, d)$
- and (e_{\wp}) is the effort policy when faced with compensation policy $\wp(q(e))$,

$$e_{\wp} = \arg \max_{e > 0} V(s; e | \{\mu, r, d\}, \wp(.))$$

- Approach will be to simulate effort and sales, under the counterfactual plans
- A comparison to current policy quantities $\{\Pi^*, q^*, e^*\}$ allows a relative evaluation

Evaluating Counterfactual Plans

Finding Profit Enhancing plans

- Evaluate aspects of plan by simulating changes to contract
- Simulations reveal profit enhancing changes
 - Elimination of Quotas (no effort shading) and Ceiling (not enough demand variance)
 - Incentive horizon reduced to one month (eliminates early shirking)
 - Linear contract (no intertemporal shifting of effort)
 - Heterogenous plans

A New Compensation Plan

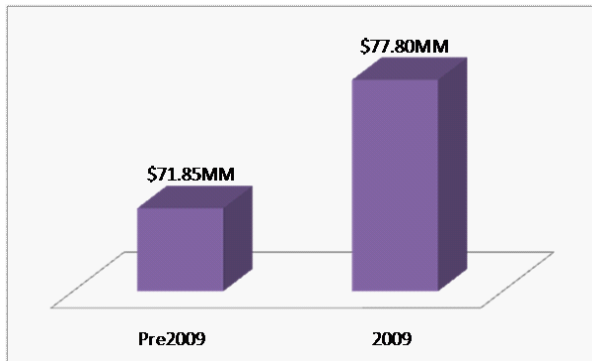
Design & Implementation

- A “Feasible” set defined by cultural, legal and infrastructural constraints at firm recommended
- Firm chose one of recommended plans
- Chosen feasible plan predicts sales would increase by about 8.6% and profits by 5.2%
- New plan introduced introduced in January, 2009
- Most changes deemed profit enhancing are incorporated (cannot reveal exact detail)
- Salespeople were informed and educated about the changes in the last quarter of 2008
- Provide evaluation using data from the first two quarters of 2009

A New Compensation Plan

Overall Results

- Significant improvement in revenues!

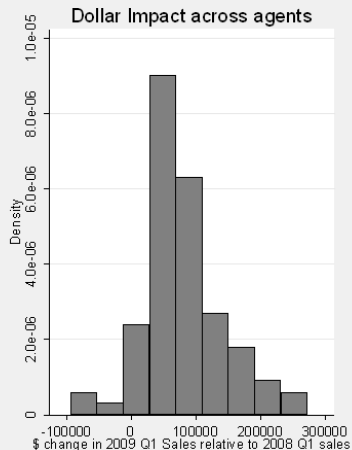
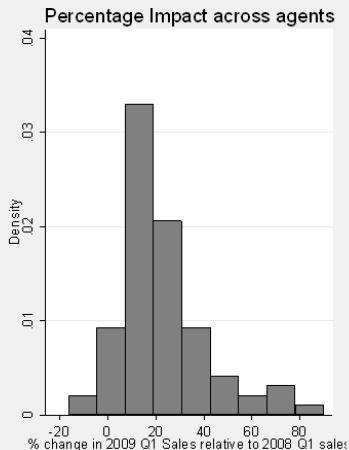


- Year over year the increase in revenues is over 9%
- Profits (*) are also up significantly (over 6%)

A New Compensation Plan

Impact Distribution across agents

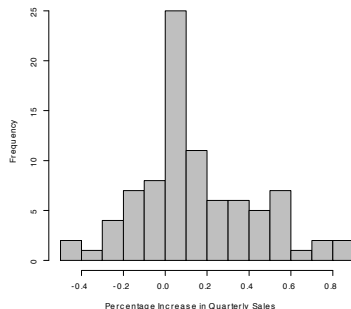
Impact of New Plan on Quarterly Revenues



A New Compensation Plan

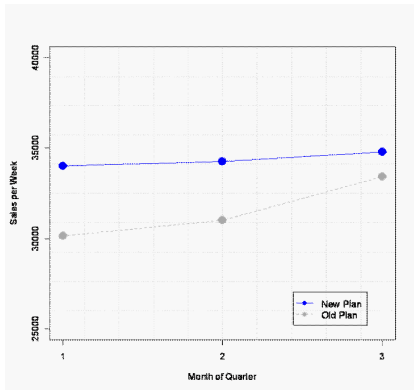
Predictions versus realized sales

- Estimated risk aversion parameter and cost of effort were significant predictors of the change in sales (p-values = 0.0414; 0.0002)
- Salespeople with lower cost of effort and risk aversion parameters exhibited larger improvements in sales



Validating the Model

Are changes in sales patterns as predicted?



- Significant changes in the pattern within quarter
- Sales under new plan are essentially flat!
- While all months are higher months 1 & 2 are significantly so

Conclusions

- Developed a comprehensive framework to measure the net effects of sales-force incentive schemes and evaluate counterfactual compensation plans
- Demonstrates value of using structural approaches and numerical dynamic programming based solutions to address real business problems
- Field implementation provides external validity and supports assumptions employed

- Understanding composition and compensations
 - The role and scope of contract externalities
- Optimal Dynamic Contracting
 - Are real world contracts dynamically optimal?
- The Contract Economics of Startups
 - Thinking of startups as a evolution of contracts and agents