

Managing Learning and Turnover in Employee Staffing*

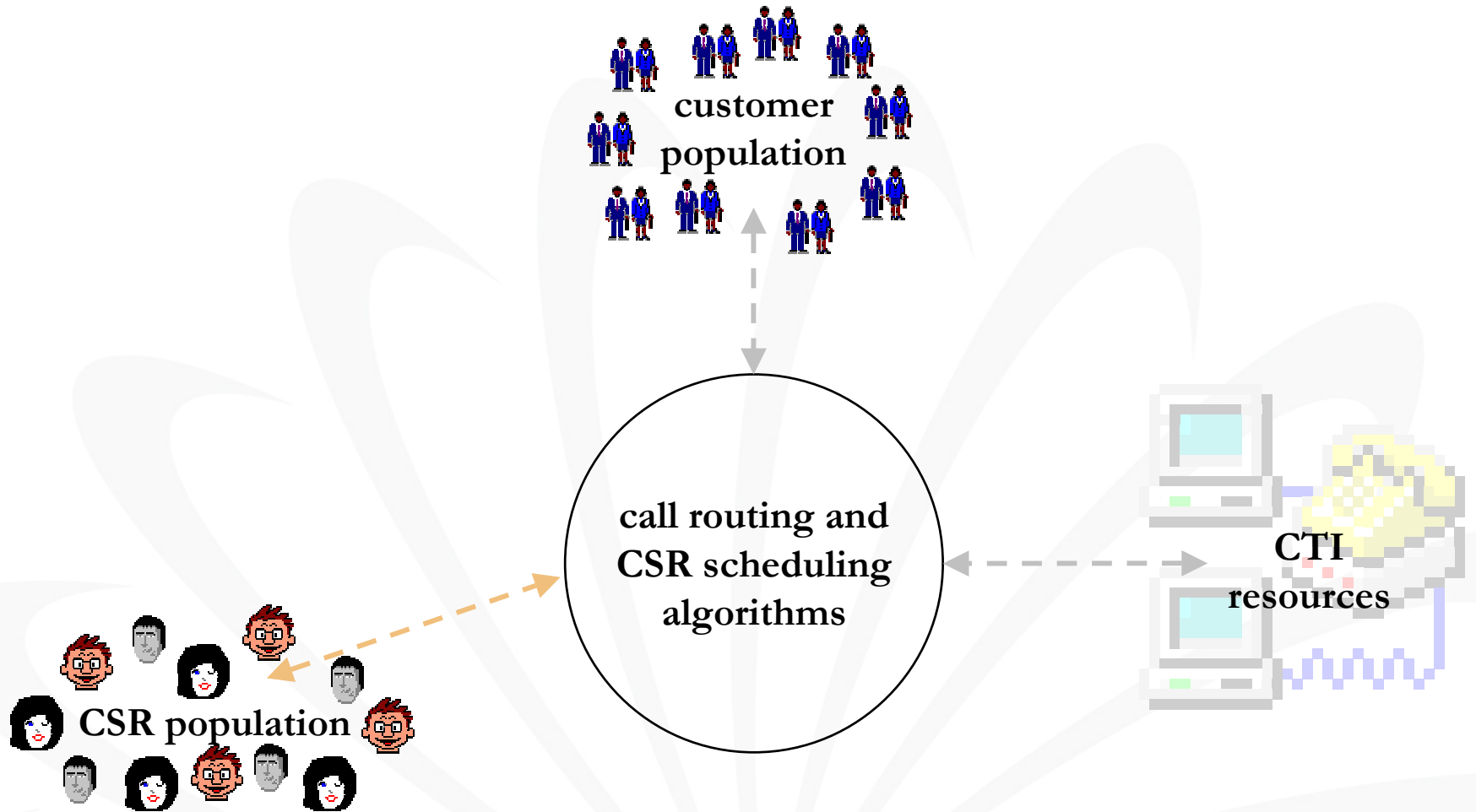
Yong-Pin Zhou

University of Washington Business School

Coauthor:

Noah Gans, Wharton School, UPenn

Call Center Operations



CSR Plans Are Made at 3 Levels

- Moment-by-moment
 - on-the-fly adjustments of call routing, overtime, lunch, breaks
- Week-by-week
 - assignment of CSRs to schedules
- Monthly, quarterly
 - target staffing levels set and hiring plans made

Our Hierarchical Approach

- Focus on high level staffing problem
 - formulate as discrete-time dynamic program
- Embed lower level scheduling and call routing problems
 - scheduling software

Related Literature

- Aggregate Planning
 - HMMS Model (1960)
 - Orrbeck, Schuette, and Thompson (1968)
 - Grinold and Stanford (1974)
 - Ebert (1976)
 - Khoshnevis and Wolfe (1986)
 - Gerchak, Parlar, and Sengupta (1990)
- Human Resources
 - Bartholomew, Forbes, and McClean (1991)

Related Literature

- Inventory Theory
 - Arrow, Karlin and Scarf (1958)
 - Veinott (1965)
 - Iglehart and Jacquett (1969)
 - Karlin (1960) and Zipkin (1989)
 - Gerchak and Wang (1994)
- Call Center
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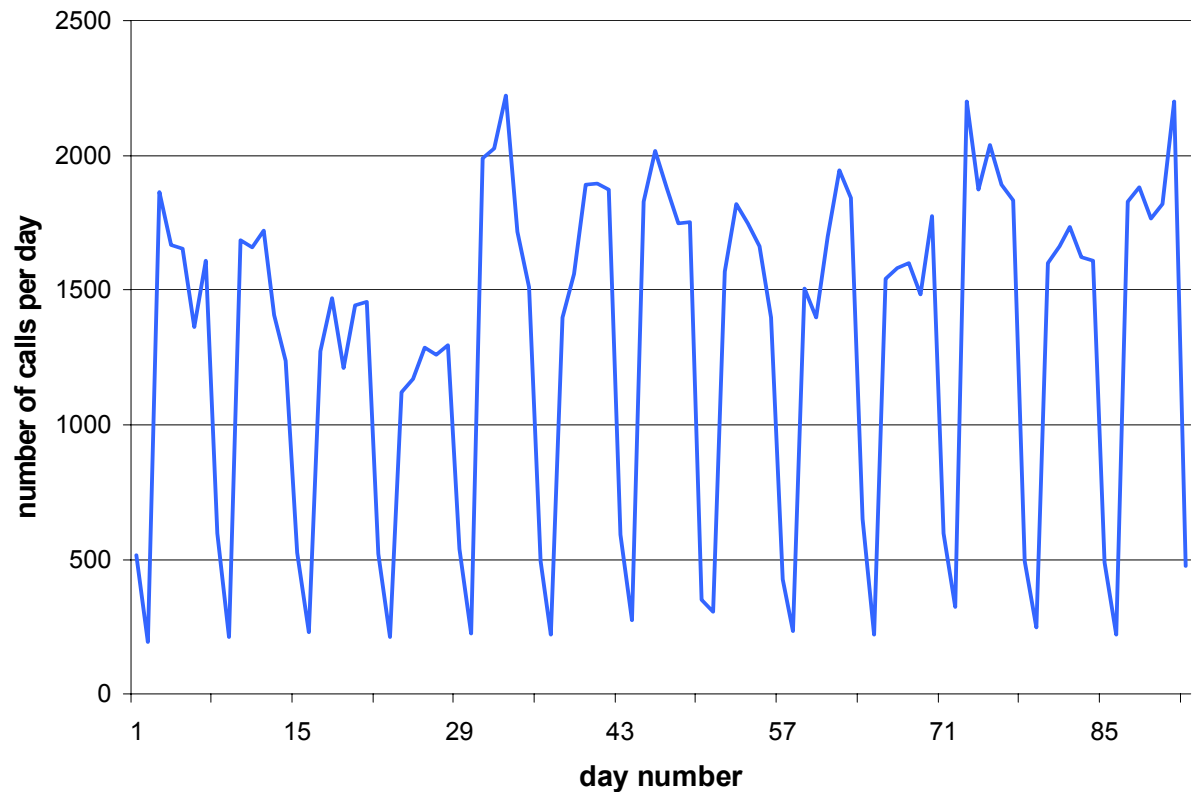
Staffing Problem Description

- Stochastic non-stationary service requirement
- Uncertain service capacity: people's service capacities change
 - Random learning
 - Random turnover
- Long hiring/training leadtime

Service Requirement ...

... Varies by

- Time of day
- Day of week / month
- Month of year
- Promotion
- Special events



Learning

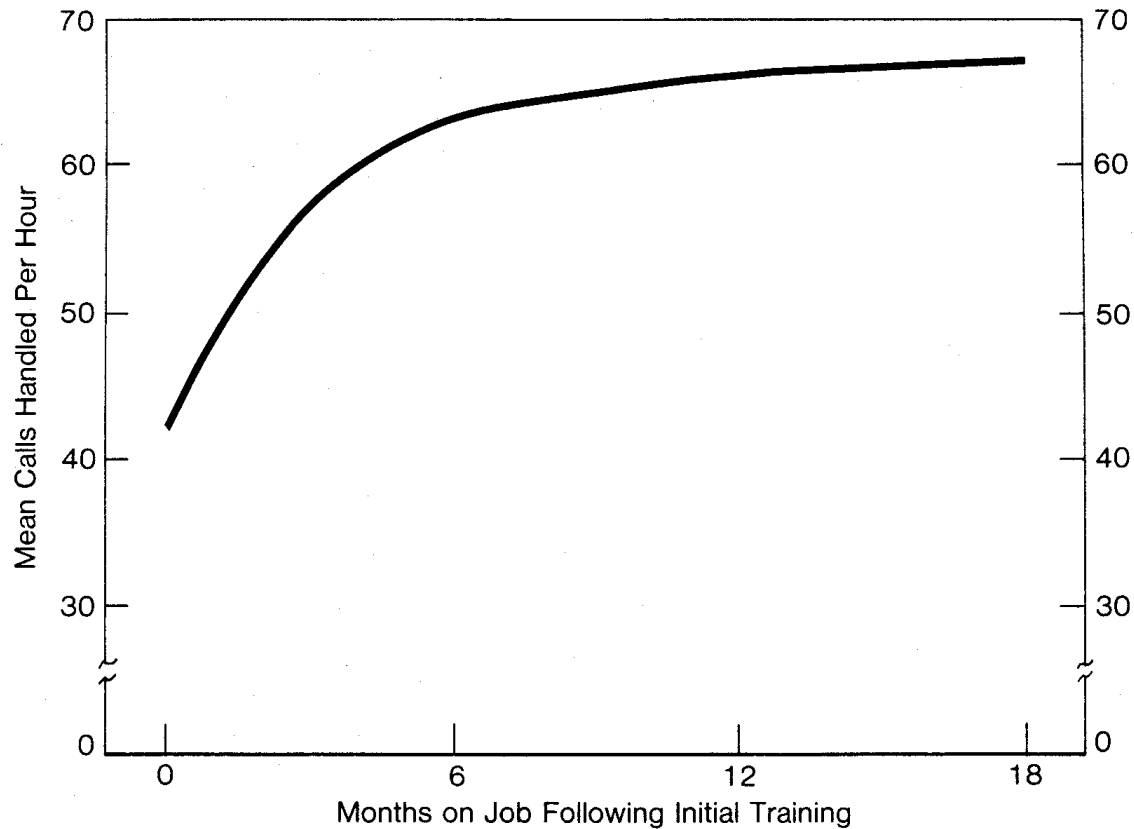


Fig. 3. *Proficiency acquisition curves for Directory Assistance Operators*

Source: Gustafson, 1982

Turnover

- Call centers have high turnover rates
- Turnover decreases with job tenure
 - Empirical evidence
 - National Longitudinal surveys (Light and Ureda 1990)
 - AT&T service representatives (Gustafson 1982)
 - Theoretical explanations
 - longer tenure implies better "fit" (Jovanovic 1979)
 - repeated searches for a better job (Parsons 1973)

Outline

- Model
- Structural Properties of The Optimal Policy
- Model Details
- Numerical Analysis

Model: Dynamics

- Indices
 - $i = 1, 2, \dots, m$: employee type
 - $t = 0, 1, \dots, T$: discrete time periods
- State variables
 - $n_{i,t}$: type i employee, time period t
(before hiring)
 - x_t : type 1 employee hired, time period t
(decision variable)
 - y_t : type 1 employee, time period t
(after hiring)

$$y_t = n_{1,t} + x_t$$

Model: Dynamics (contd)

- Transition variables
 - $\tilde{q}_{i,t}(n_{i,t})$: type- i employees who turn over, out of $n_{i,t}$ **random variable**

$$\tilde{q}_{i,t}(n_{i,t}) = n_{i,t} \cdot \tilde{q}_{i,t}$$

- $\tilde{l}_{i,t}(n_{i,t})$: type- i employees who learn, out of $n_{i,t}$ **random variable**

$$\tilde{l}_{i,t}(n_{i,t}) = n_{i,t} \cdot (1 - \tilde{q}_{i,t}) \cdot \tilde{l}_{i,t}$$

Model: Costs

- α : discount factor
- h : unit hiring cost
- W_i : per stage wage cost, type i employee
- $O_t(y_t, n_{2,t}, \dots, n_{m,t}; D_t)$: operating cost including overtime and outsourcing costs
 - jointly convex in $(y_t, n_{2,t}, \dots, n_{m,t})$

Dynamic Programming Model

$$V_t(n_{1,t}, \dots, n_{m,t})$$

$$= \min_{x_t \geq 0} \left\{ \begin{aligned} &hx_t + W_1(n_{1,t} + x_t) + \sum_{i=2}^m W_i n_{i,t} \\ &+ O_t(n_{1,t} + x_t, n_{2,t}, \dots, n_{m,t}) \\ &+ \alpha E_{\{\tilde{q}_{1,t}, \dots, \tilde{q}_{m,t}, \tilde{l}_{1,t}, \dots, \tilde{l}_{m-1,t}\}} \left\{ V_{t+1}(n_{1,t+1}, \dots, n_{m,t+1}) \right\} \end{aligned} \right\}$$

Constraints

- Subject to the system dynamics:

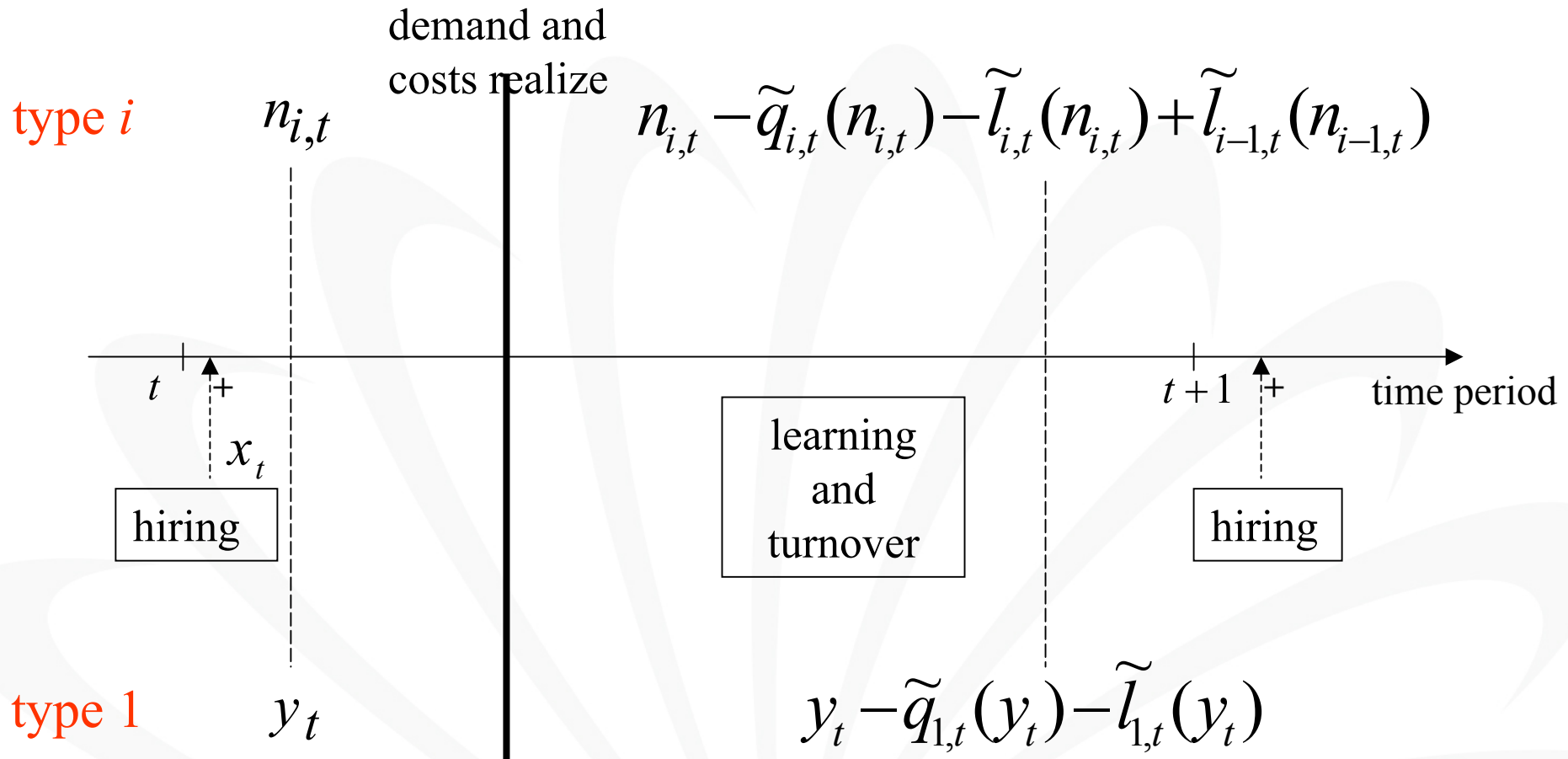
$$y_t = n_{1,t} + x_t$$

$$n_{1,t+1} = y_t - \tilde{q}_{1,t}(y_t) - \tilde{l}_{1,t}(y_t)$$

$$n_{i,t+1} = n_{i,t} - \tilde{q}_{i,t}(n_{i,t}) - \tilde{l}_{i,t}(n_{i,t}) + \tilde{l}_{i-1,t}(n_{i-1,t})$$

$$\forall 1 < i \leq m.$$

System Dynamics Illustrated



Optimality of “Base-Level” Policy

- Base-Level Policy: there exist state-dependent base-levels $y_t^*(n_{2,t}, \dots, n_{m,t})$ such that the optimal number to hire is

$$x_t^* = \begin{cases} y_t^*(n_{2,t}, \dots, n_{m,t}) - n_{1,t} & \text{if } n_{1,t} < y_t^*(n_{2,t}, \dots, n_{m,t}) \\ 0 & \text{otherwise} \end{cases}$$

Proof Idea

Convexity of one-period cost function propagates

$$V_t(n_{1,t}, \dots, n_{m,t}) \\ = \min_{y_t \geq n_{1,t}} \left\{ \begin{array}{l} (h + W_1)y_t + O_t(y_t, \dots, n_{m,t}) \\ + \alpha E \left\{ V_{t+1}(n_{1,t+1}, \dots, n_{m,t+1}) \right\} \end{array} \right\} - h n_{1,t} + \sum_{i=2}^m W_i n_{i,t}$$

where

$$n_{1,t+1} = (1 - \tilde{q}_{1,t})(1 - \tilde{l}_{1,t})y_t$$

...

$$n_{i,t+1} = (1 - \tilde{q}_{i,t})(1 - \tilde{l}_{i,t})n_{i,t} + \tilde{l}_{i-1,t}n_{i-1,t}$$

...

$$n_{m,t+1} = (1 - \tilde{q}_{m,t})n_{m,t} + \tilde{l}_{m-1,t}n_{m-1,t}$$

Proof Idea

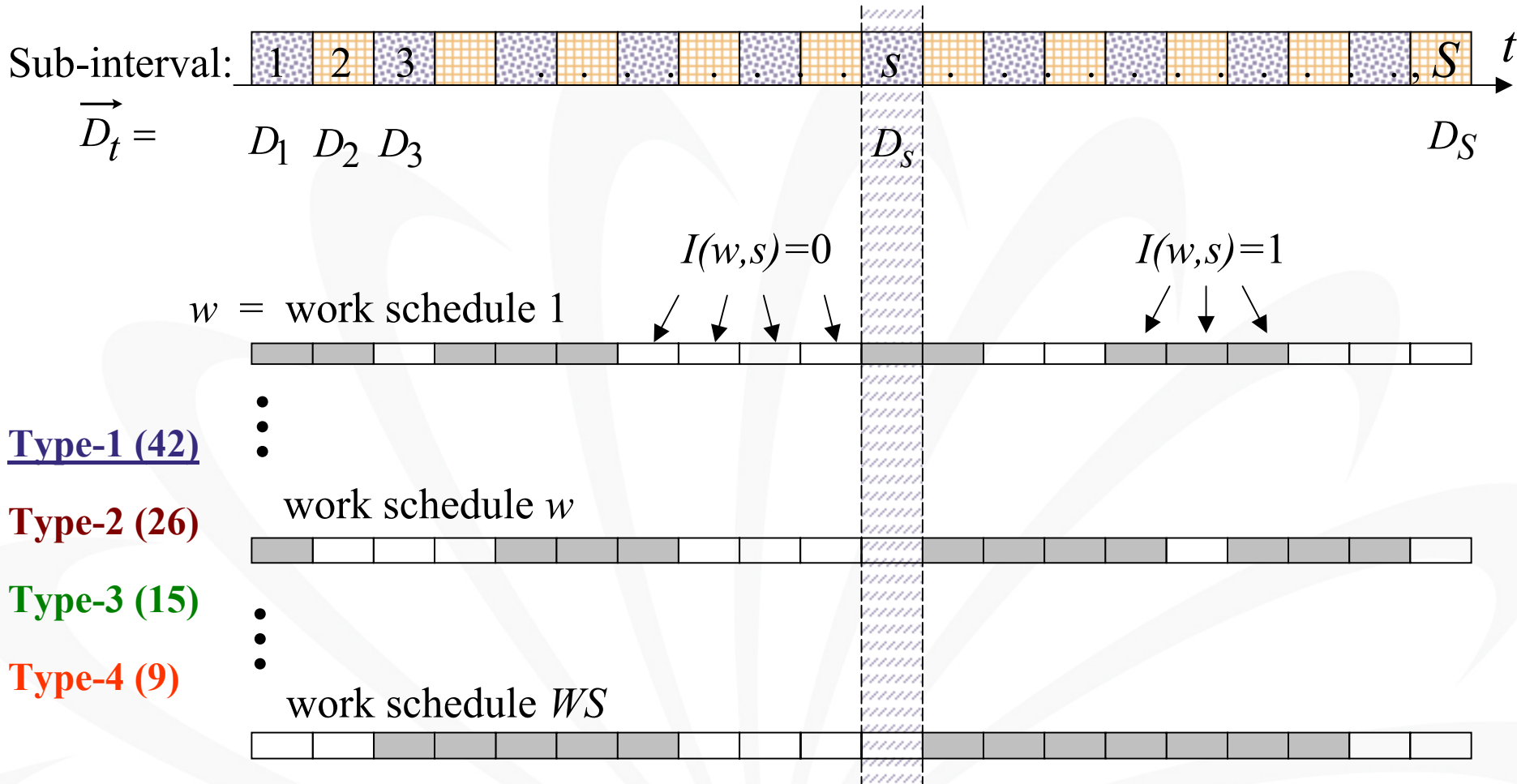
Lemma 1

If $V_{t+1}(n_{1,t+1}, \dots, n_{m,t+1})$ is jointly convex in $(n_{1,t+1}, \dots, n_{m,t+1})$ and $n_{i,t+1}$ ($i = 1, \dots, m$) are linear functions of $(y_t, \dots, n_{m,t})$, then $V_{t+1}(n_{1,t+1}, \dots, n_{m,t+1})$ is jointly convex in $(y_t, \dots, n_{m,t})$.

Lemma 2

Let $f(n_1, n_2, \dots, n_m) = \inf_{y \geq n_1} \{h(y, n_2, \dots, n_m)\}$. If $h(y, n_2, \dots, n_m)$ is jointly convex in (y, n_2, \dots, n_m) , then $f(n_1, n_2, \dots, n_m)$ is jointly convex in (n_1, n_2, \dots, n_m) .

Model Details (I): Operating Cost Function

$$O_t(y_t \dots n_{m,t}; D_t)$$


Operating Cost Function $O(t)$

- $s = 1, \dots, S$: sub-intervals in any hiring period t
- $w = 1, \dots, WS$: work schedules ($I(w,s) = 1$ if w requires working in sub-interval s , 0 otherwise).
- $x_{i,w}$ = the number of type- i employees assigned to work schedule w
- z_s : the amount of work to be outsourced in sub-interval s

Operating Cost Function $O(t)$

$$O_t(y_t, n_{2,t}, \dots, n_{m,t}; \vec{D}_t) = \min_{x_{iw}, z_s} \sum_{i=1}^m \sum_{w=1}^{WS} C_{iw} x_{iw} + \sum_{s=1}^S OS_s z_s$$

$$\text{s.t.} \quad \sum_{i=1}^m \sum_{w: I(w,s)=1} \mu_i x_{iw} + z_s \geq D_s \quad \forall s \quad (1)$$

$$\sum_{w=1}^{WS} x_{1w} \leq y_t$$

$$\sum_{w=1}^{WS} x_{iw} \leq n_{i,t} \quad \forall 1 < i \leq m$$

$$x_{iw}, z_s \geq 0 \quad \forall i, w, s$$

- c.f. Berman, Larson, Pinker (1997)

More General $O(t)$

$$O_t(y_t, n_{2,t}, \dots, n_{m,t}; \vec{D}_t) = \min_{x_{iw}, z_{sj}} \sum_{i=1}^m \sum_{w=1}^{WS} C_{iw}(x_{iw}) + \sum_{s=1}^S \sum_{j=1}^J OS_{sj}(z_{sj})$$

$$\text{s.t. } f(x_{11}, \dots, x_{mWS}, z_{11}, \dots, z_{SJ}; \vec{D}_t) \geq 0 \quad (2)$$

$$\sum_{w=1}^{WS} x_{1w} \leq y_t$$

$$\sum_{w=1}^{WS} x_{iw} \leq n_{i,t} \quad \forall 1 < i \leq m$$

$$x_{iw}, z_{sj} \geq 0 \quad \forall i, w, s, j$$

Example of (2): $P\{\text{waiting in queue} \geq U\} \leq a\%$.

Model Details (II): Hiring/Training Lead Time

- Hiring/training lead time $\lambda > 0$ modeled by adding employee types
 - $m > \lambda$
 - $i = 1, \dots, \lambda$ indicate hiring/training stages
- Hiring lead time
 - $\mu_i = 0$
 - W_i may be 0
- Training lead time
 - μ_i may be 0
 - $W_i > 0$

Numerical Analysis

- Modeling of different CSR classes
- Modeling of the stochastic elements

Compare

- Optimal policy: state-dependent
 - recognize different employees have different capacities
 - take into account the stochastic elements
- Heuristic: headcount policy
 - check only the total number of employees
 - assume everyone has the same (average) capacity
 - take into account the stochastic elements
- Heuristic: LP-based
 - recognize different employees have different capacities
 - use average learning and turnover rates

Example Problem: Slow and Fast CSRs

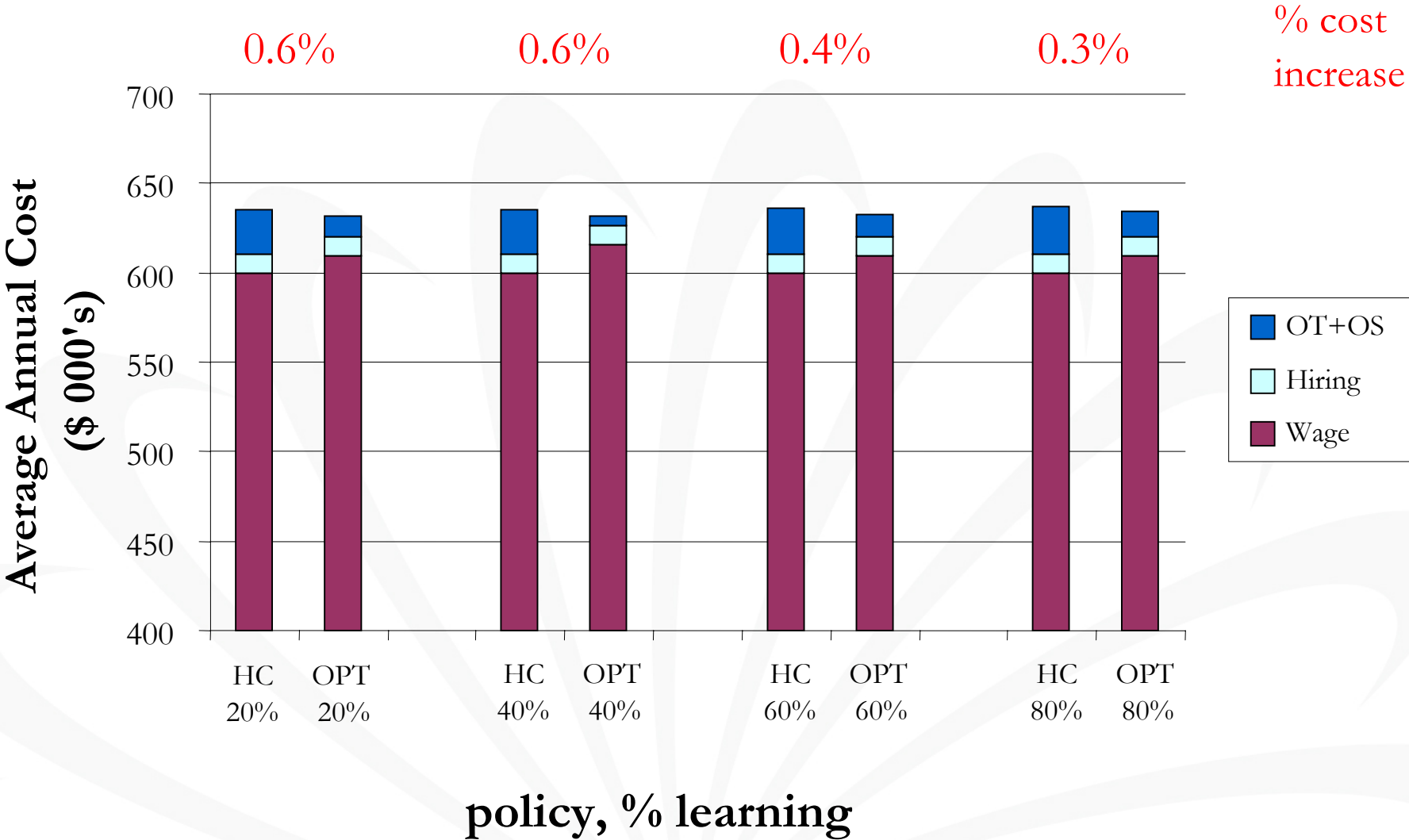
- CSR turnover
 - slow CSRs: 48% yearly (15% per quarter)
 - fast CSRs: 34% yearly (10% per quarter)
- CSR learning: slow and fast CSRs
 - CSR becomes fast after one quarter of experience
 - % speedup with learning: 20%, 40%, 60%, 80%
- Call volume
 - Demand: 250,000 per quarter
 - CSR capacity: 10,000 per quarter on average
 - Overtime as % of regular time: 10%, 20%, 30%

Costs Used for Examples

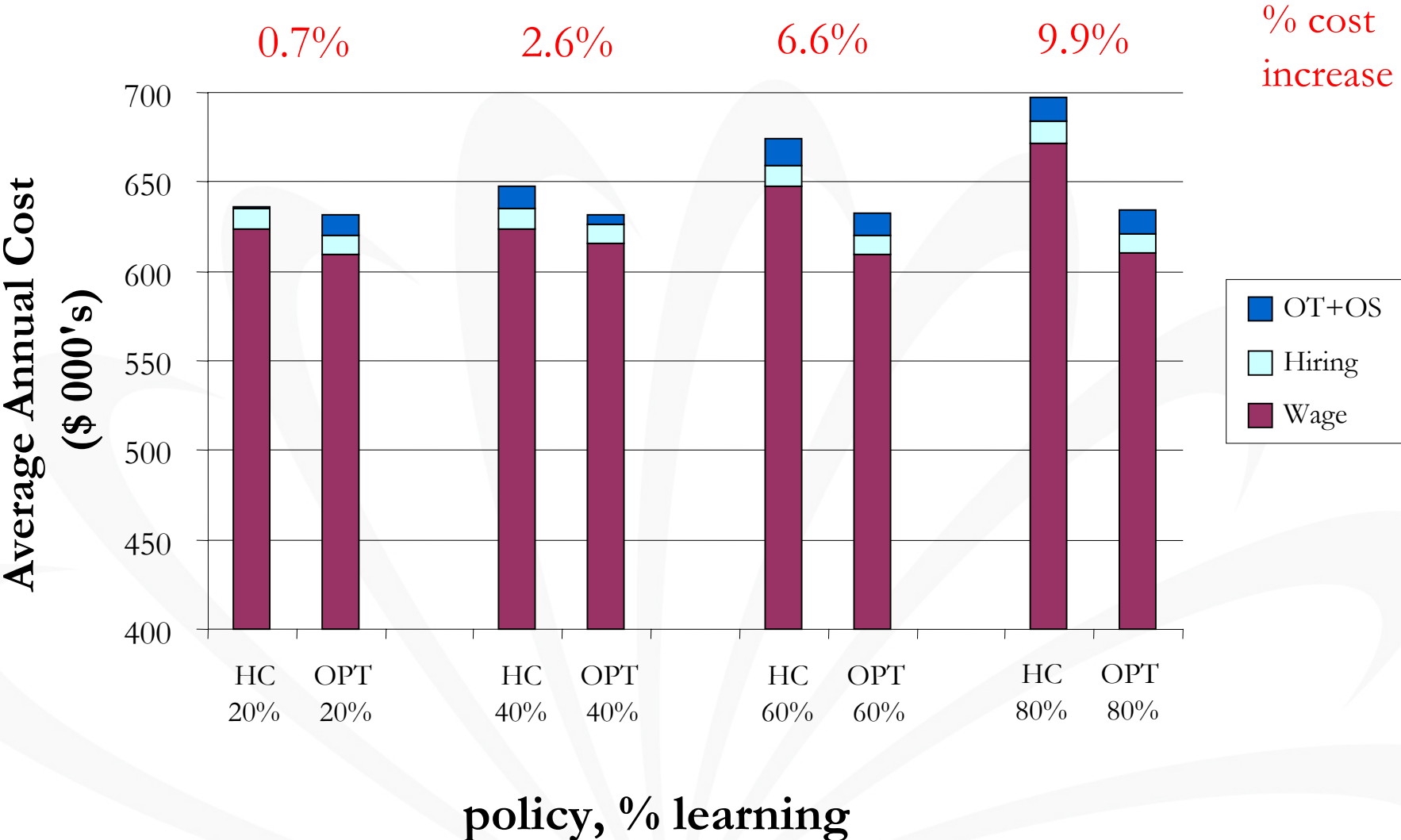
Wage is capacity neutral, for an avg. capacity of 10,000 per quarter:

- Base wage: \$4,500, benefits: \$1,000 (22.2%)
- Overtime: 1.5 times base wage rate
- Outsource cost: 1.8, 18, 36, 91, 182 times base wage rate

Optimal v.s. Headcount: Ample Overtime

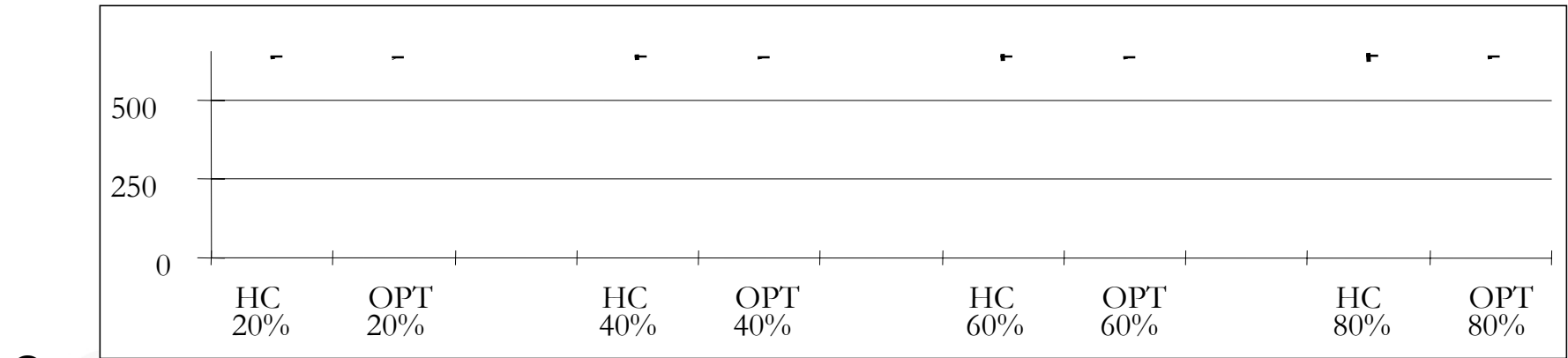


Optimal vs Headcount: Limited Overtime

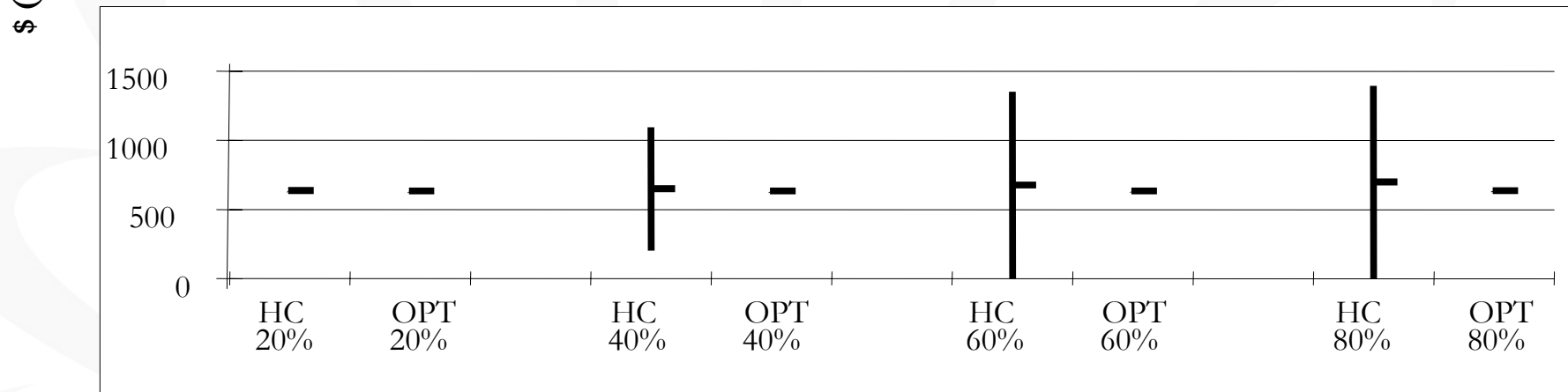


Cost Variability

Average Cost +/- 1 Standard Deviation



Ample Overtime

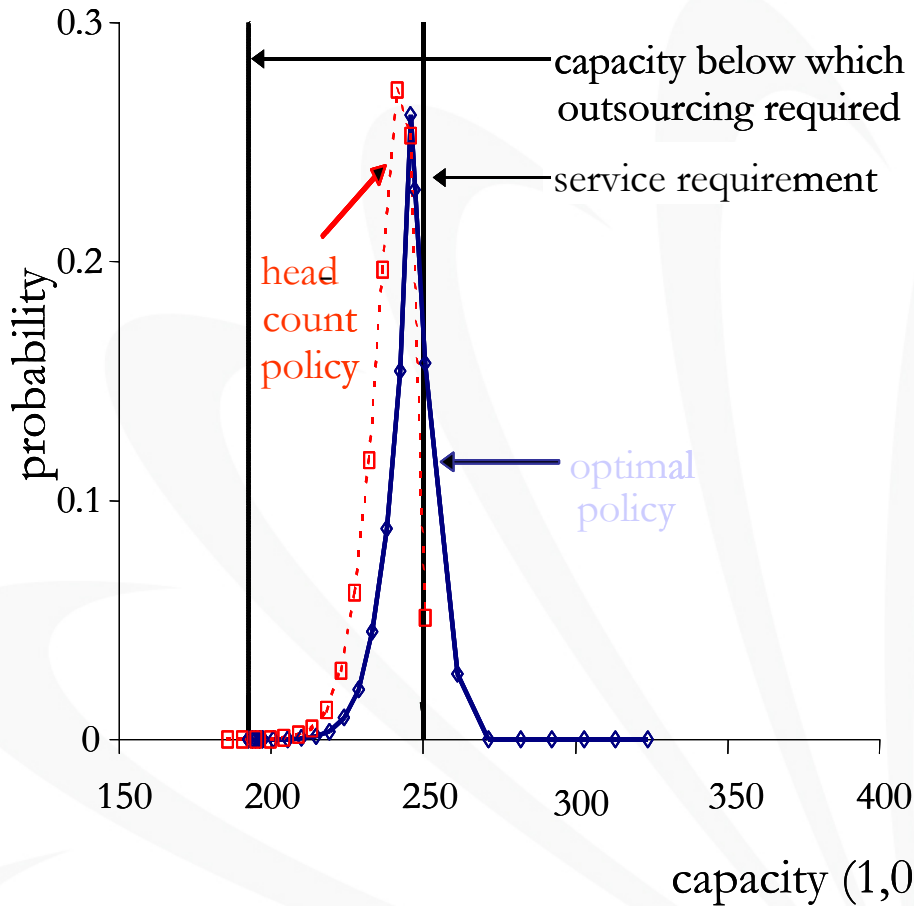


Limited Overtime

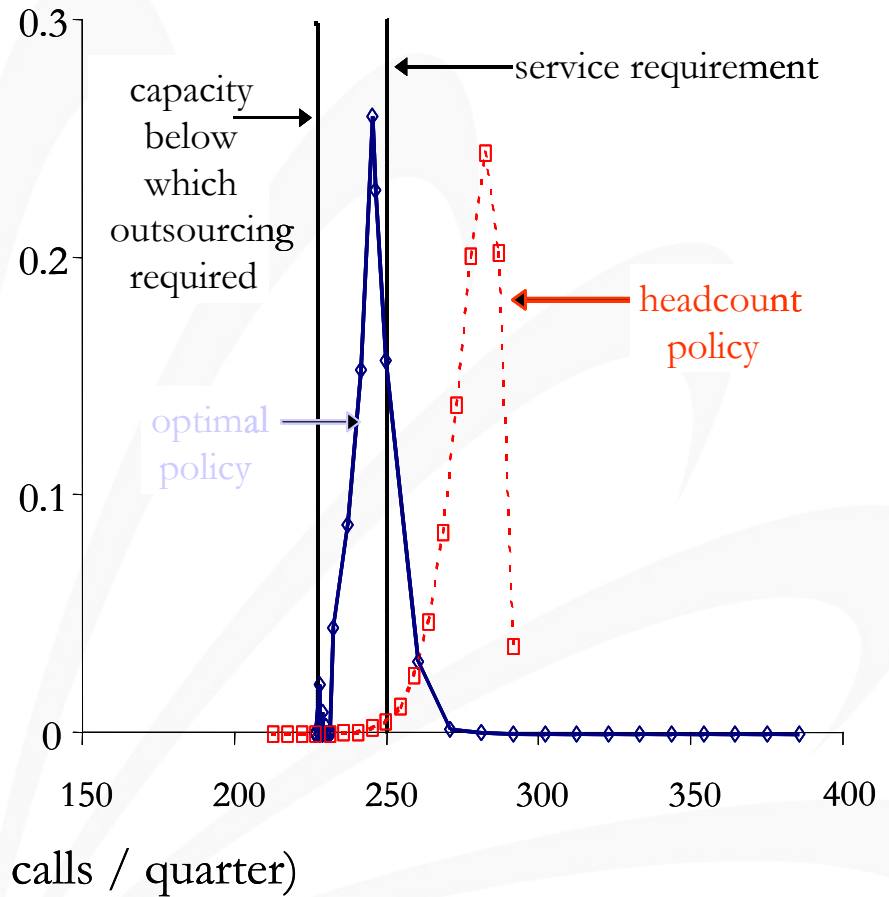
Optimal vs. Headcount

Distribution of Regular Time Capacity

30% overtime



10% overtime



Our Findings

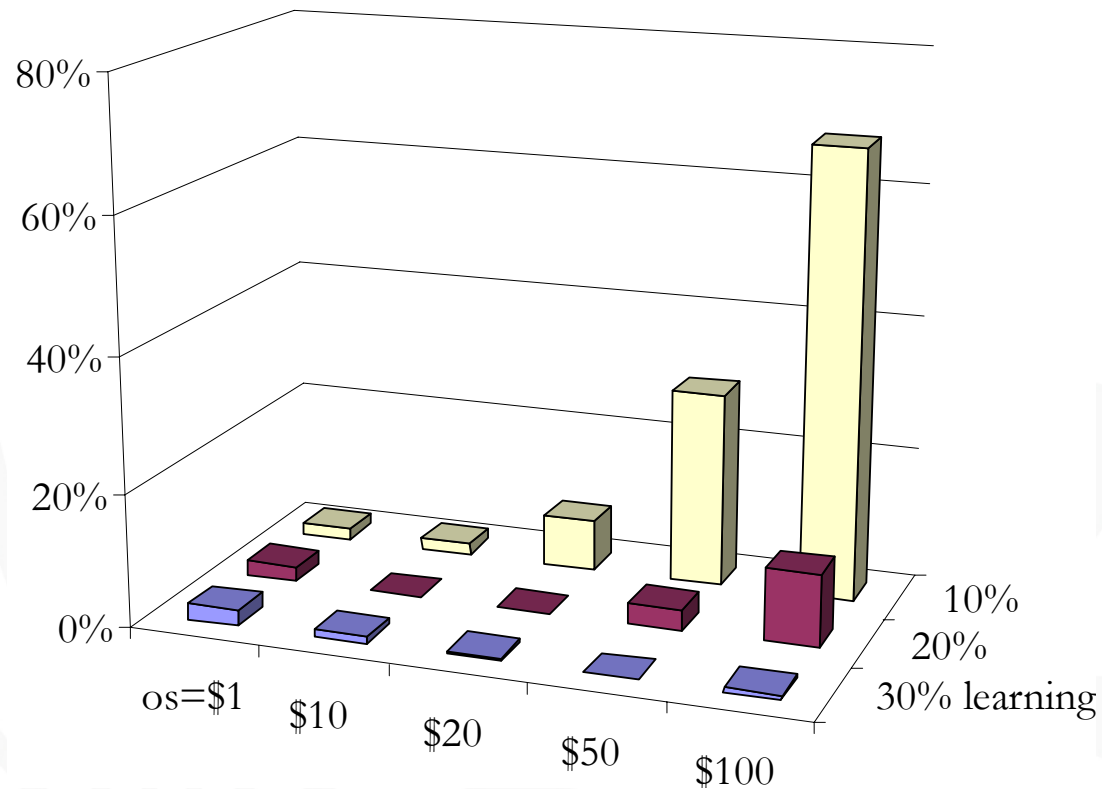
- When spare capacity is cheap
 - simpler hiring schemes do well
 - less need to track employees carefully

⇒ heuristics that ignore learning
- When spare capacity is tight
 - simpler hiring schemes can do poorly
 - value to tracking employees carefully

⇒ more complex approaches: DP, IPA

Optimal vs. LP Heuristic

- W/O training LT: LP does very well
- When there is a one-period leadtime: LP capacity often falls short



Our Findings

- LP heuristic performs well, when ...
 - ... training is fast -- LP can wait to observe the number of employees available and hire accordingly
 - ... training takes time, but overtime is ample, or OS (or service failure) is cheap -- OT and OS can cover capacity shortfall when turnover is higher than average
 - The optimal “base level” structure simplifies calculation
- LP performs poorly, when ...
 - ... overtime is limited and OS (or service failure) is expensive
 - If MDP is too complicated to use, then at least add some “buffer capacity” in conjunction with LP

Summary

- A new framework to study employee staffing problems
 - DP approach, allows randomness in the system
 - Natural incorporation of the scheduling and routing problem
 - Optimal policy and structural properties
- Numerical results show when it is worthwhile to more carefully monitor system capacity