



Changing the rules of business™

From Planning to Operations: The Ever-Shrinking Optimization Time Horizon

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Deriving Benefit from Increased Solver Power



- Revisit previously shelved applications
- Build bigger, more accurate models
 - *Example:* Recent supply-chain model with 10 million constraints, 19 million variables (solve in 1.5 hours)
- Optimize “globally”, over entities that were previously treated separately
- *Move from the traditional Operations Research domain of planning to (real-time) operations: Business execution*

Three Success Stories



Using real-time optimization

“Tales from the cutting edge”

Ann Bixby & Brian Downs, Aspen Technologies

“The dance of the thirty-ton trucks”

Martin Durbin, Decisive Analytics

Karla Hoffman, George Mason University

Real-time production-line scheduling for front-end semiconductor fabs

ILOG

- **Each of these applications uses optimization**
 - Linear and Mixed-Integer Programming
- **Question: Did increased solving power really make a difference? Could we have done this 5 years ago?**



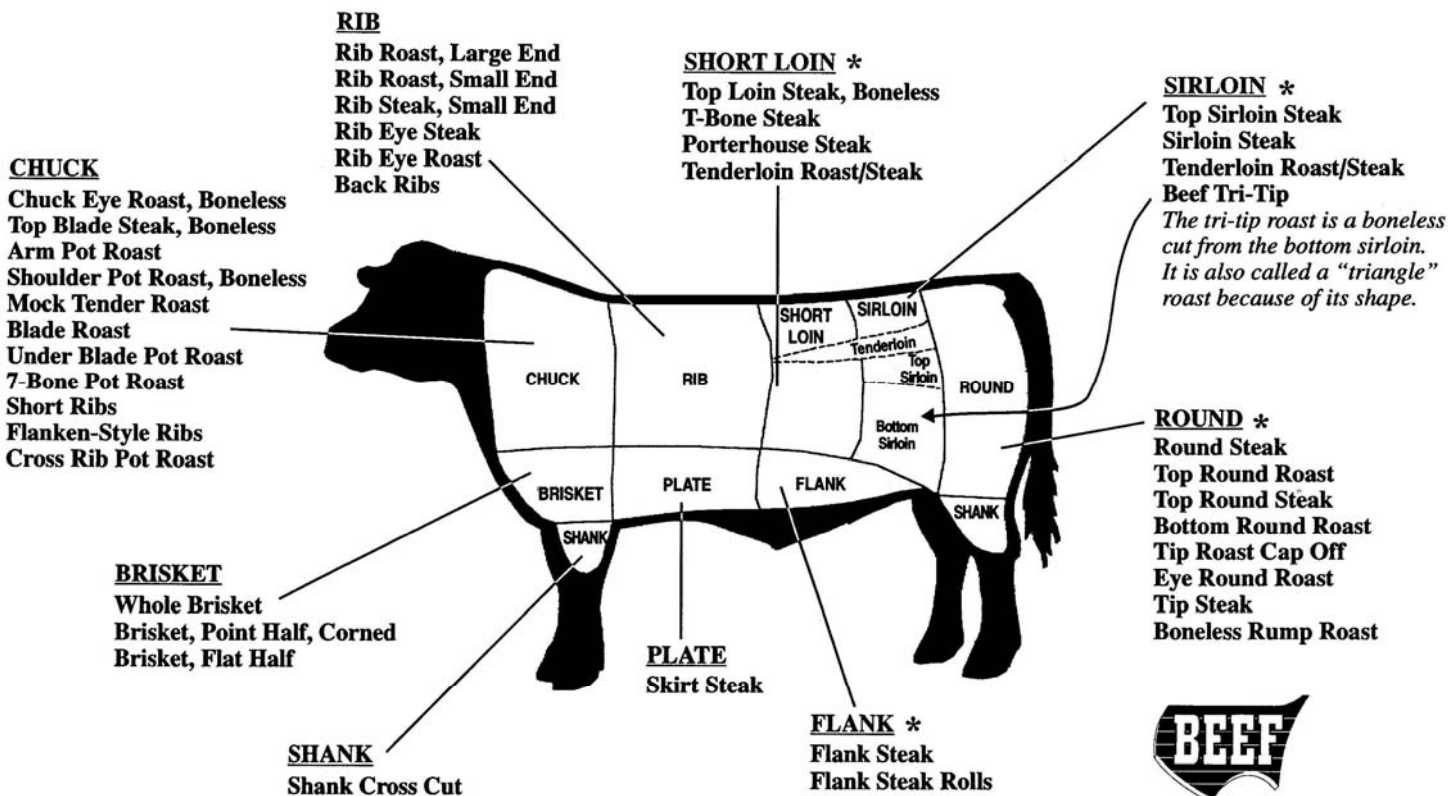
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**Tales from the Cutting Edge: A
real-time capable-to-promise
model for meat disaggregation
at Swift & Co.**



— BEEF CUTS —

Where They Come From



National Cattlemen's Beef Association
 444 North Michigan Avenue
 Chicago, Illinois 60611
 (312) 467-5520

* Beef primals that feature cuts lowest in fat.

Beef Disaggregation



- **The problem**

- 5 meat processing plants
- Carcass inventory at each plant at shift start must be processed by shift end. Cut into 7 primals, USDA graded, “disaggregated” into pieces, and packaged.
- This process must be scheduled, taking into account existing orders and *current forecast*.
- **Schedule must interact with the sales process.**

A Carcass Disaggregation Tree



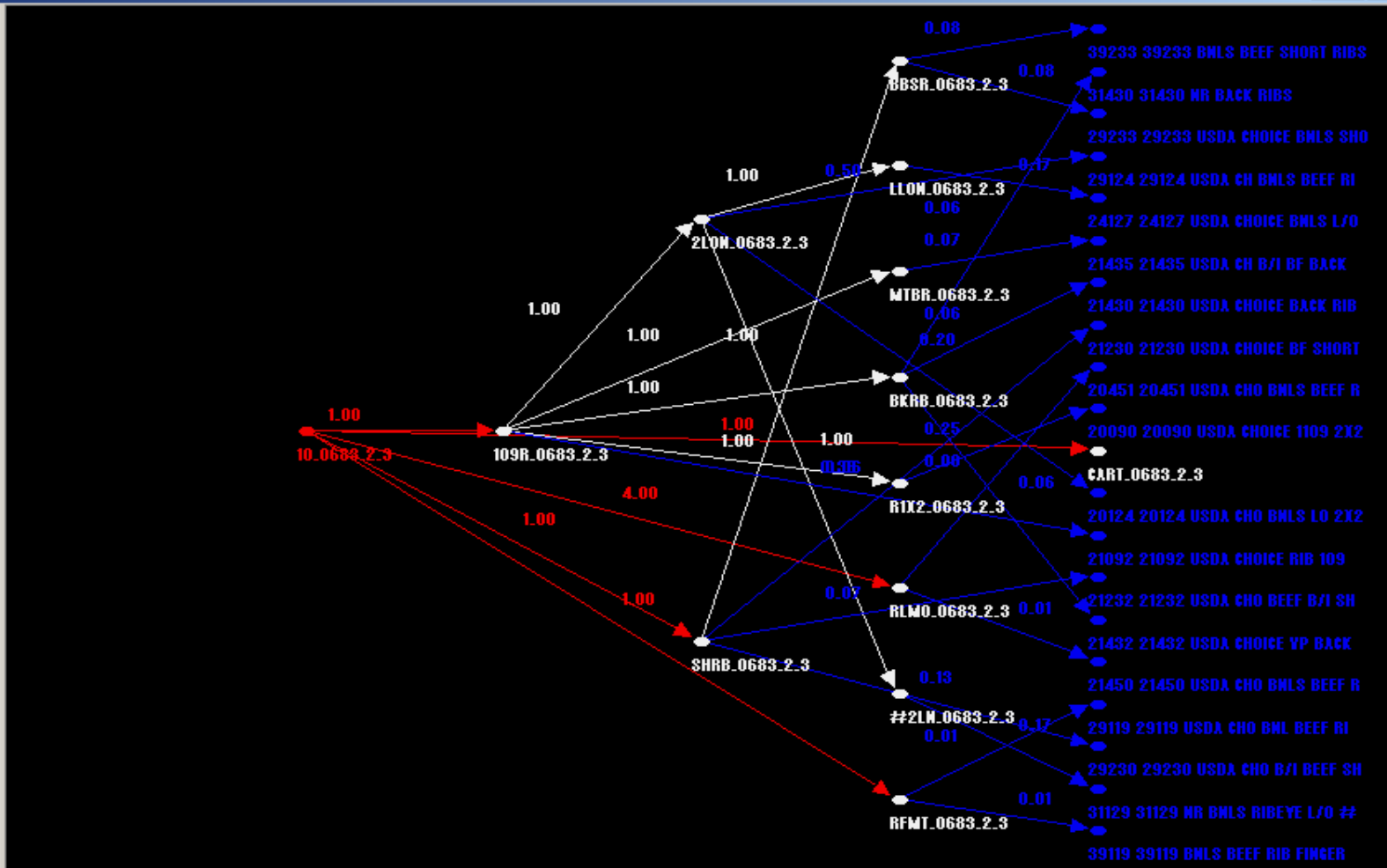
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File Data View Menus Options Window Help



Beef Disaggregation

- [-] Pack. Operations
 - [+] Carcass Data
 - [+] Inventory Data
 - [+] Demand Data
 - [-] Production Data
 - [-] Cutting Opr. Costs
 - [-] Pkg. Opr. Costs
 - [-] Frozen Cutting Sch.
 - [-] Frozen Pack. Sched.
 - [-] Carcass Disagg.
 - [-] Bill of Materials
 - [-] Condemnation %
 - [-] Fin. Prod. Recipes
 - Disagg. Diagram
 - [-] RCODE Limit Def'ns
 - [-] RCODE Limits
 - [-] FIN Limit Def'ns
 - [-] FIN Limit
- [+] Solution Generation
- [-] Solution Analysis
 - [+] Schedule Report
 - [+] Revenue Report
 - [-] Detail Reports
 - [-] Raw Material Usage
 - [-] Cut Schedule
 - [-] Co-Prod. Production
 - [-] Piece Production
 - [-] Box Production
 - [-] Fresh Sales
 - [-] Frozen Sales
 - [-] Fresh Short Dem.
 - [-] Frozen Short Dem.
 - [-] Fresh Inventory
 - [-] Frozen Inventory
 - [-] Inv. Storage Viol'n.
 - [-] RCODE Lim. Viol'n.
 - [-] FIN Lim. Viol'n.
- [+] Summary Reports
- [-] Scenario Management



Select Primal:

[View Co-Products for Selected Node](#)

What Drove the Application



- **The process**

- The schedule decides for each carcass a full disaggregation and packaging plan.
- When you take an order, you would like to know what you are “capable” of supplying, not just what’s in the schedule. This requires “moving up the tree”: **HUMANS can’t do it – not during a sales call!**

- **The result**

- Lost sales, unfulfilled orders, dissatisfied customers.

Beef Disaggregation



The solution

- Started as 1 million variable “textbook” LP model.
 - After one year of model reductions (many very complex), the model was reduced to meet memory and *resolve-time limits (< 10 seconds)*
- The Environment:
 - 300 queries and commits (LPs) handled per hour by each model
 - A total of 45 models are running fully automated handling queries and commits 24 hours per day
- The savings:
 - **\$13 million/year (determined by internal benefits study)**
 - **Inventory sold increased from 10% to 80%**
 - **Most important: Business changed fundamentally**

An LP Instance from a CTP Model



Size: 241321 constraints; 297517 variables

- Resolve-time requirement:** **<10 seconds**

- Query solve:** Resolve from advanced basis with a small number of added rows and columns
 - CPLEX 9.0 (2004) 0.7 secs
 - CPLEX 5.0 (1997) 1.2 secs
 - CPLEX 1.0 (1988) 4.4 secs

- Machine speed adjustment:**
 - CPLEX 5.0 (1997 PC -- 20x slower) 24 secs

Was increased solving power essential to this application?



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The Dance of the 30-Ton Trucks: **Real-time dispatching of** **concrete trucks for Virginia** **Concrete**



Concrete Delivery



- **The Background**

- Virginia Concrete is a part of Florida Rock.
- They deliver 500-700 loads per day to 150 customers, a total of 5000-6000 cubic yards of concrete per day.
- Deliveries occur from 10 plants with 125-150 trucks.

- **A key characteristic of the business**

- 95% of orders change before being delivered → The delivery schedule is always out-of-date.

- **The key driver for this application**

- The recognition that GPS provided a potentially very valuable technology for their business.
- **The result:** A major program to introduce GPS technology and the necessary IT infrastructure.

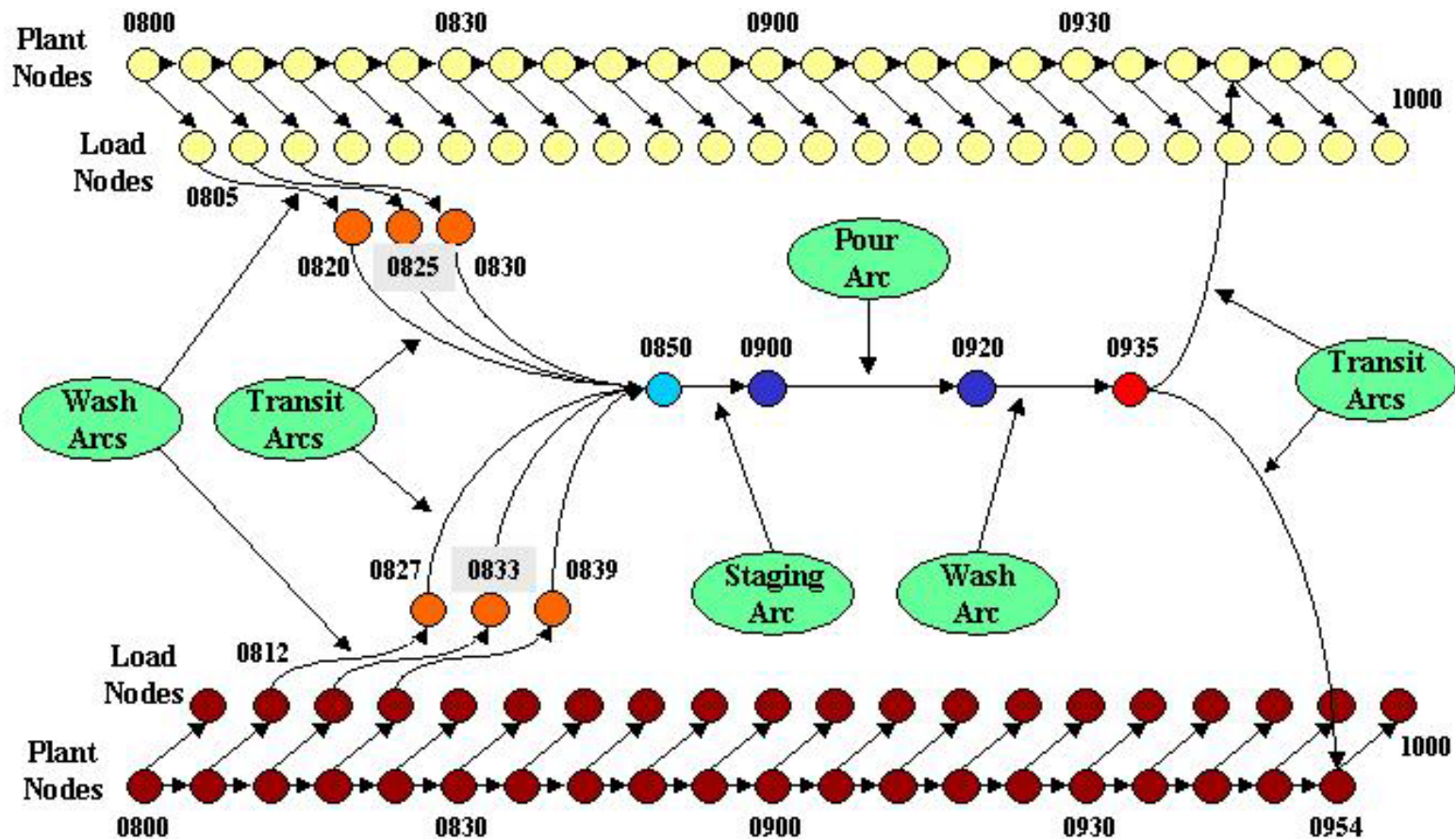
The Optimization Solution



- **Developing the solution**

- **The initial expectation:** Based upon experience, heuristics were expected to be the only viable approach.
- **The plan:** Being aware of the advances in LP/MIP technology, at least give it a try.

The Model Structure: A space-time network

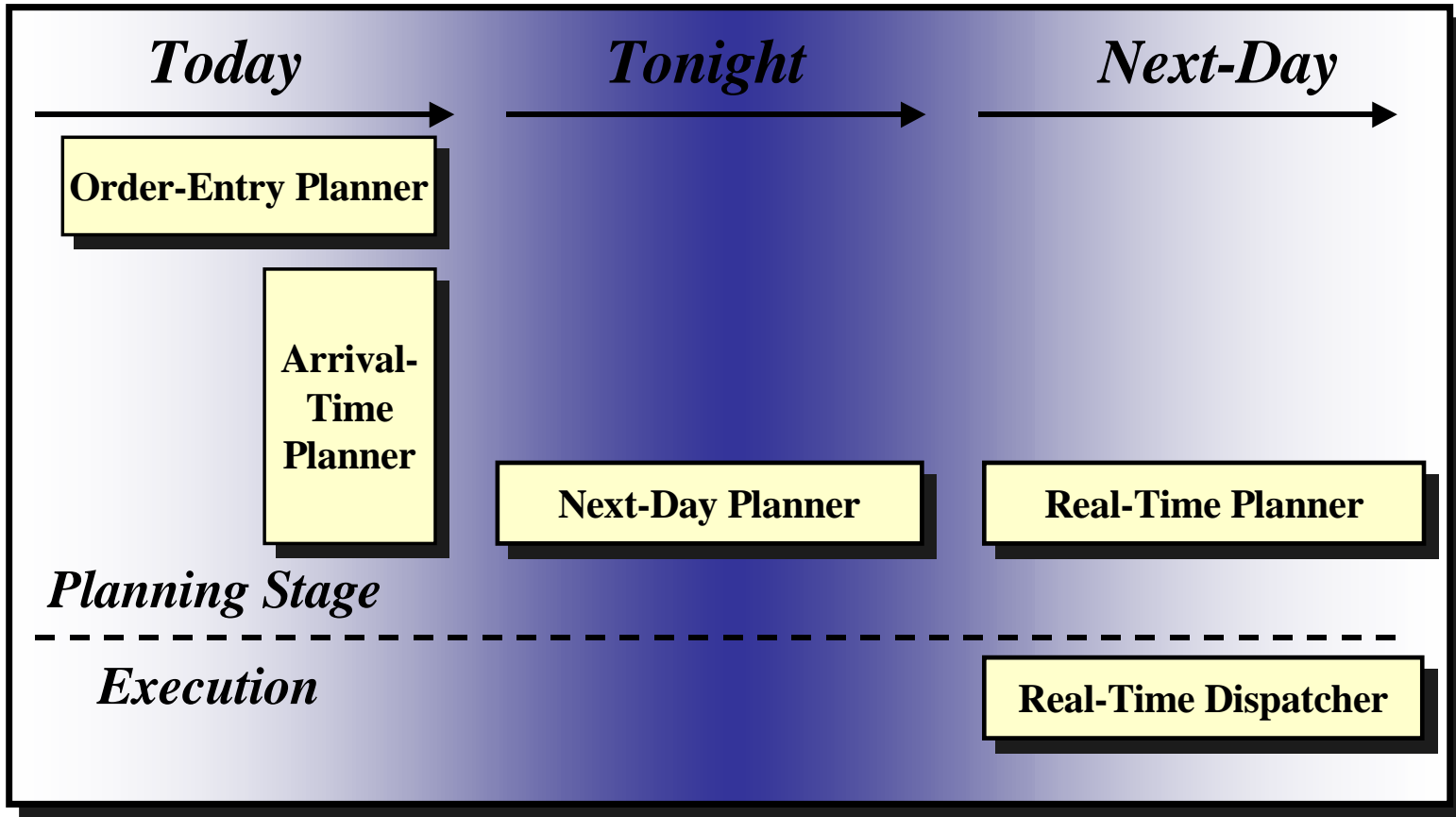


The Optimization Solution

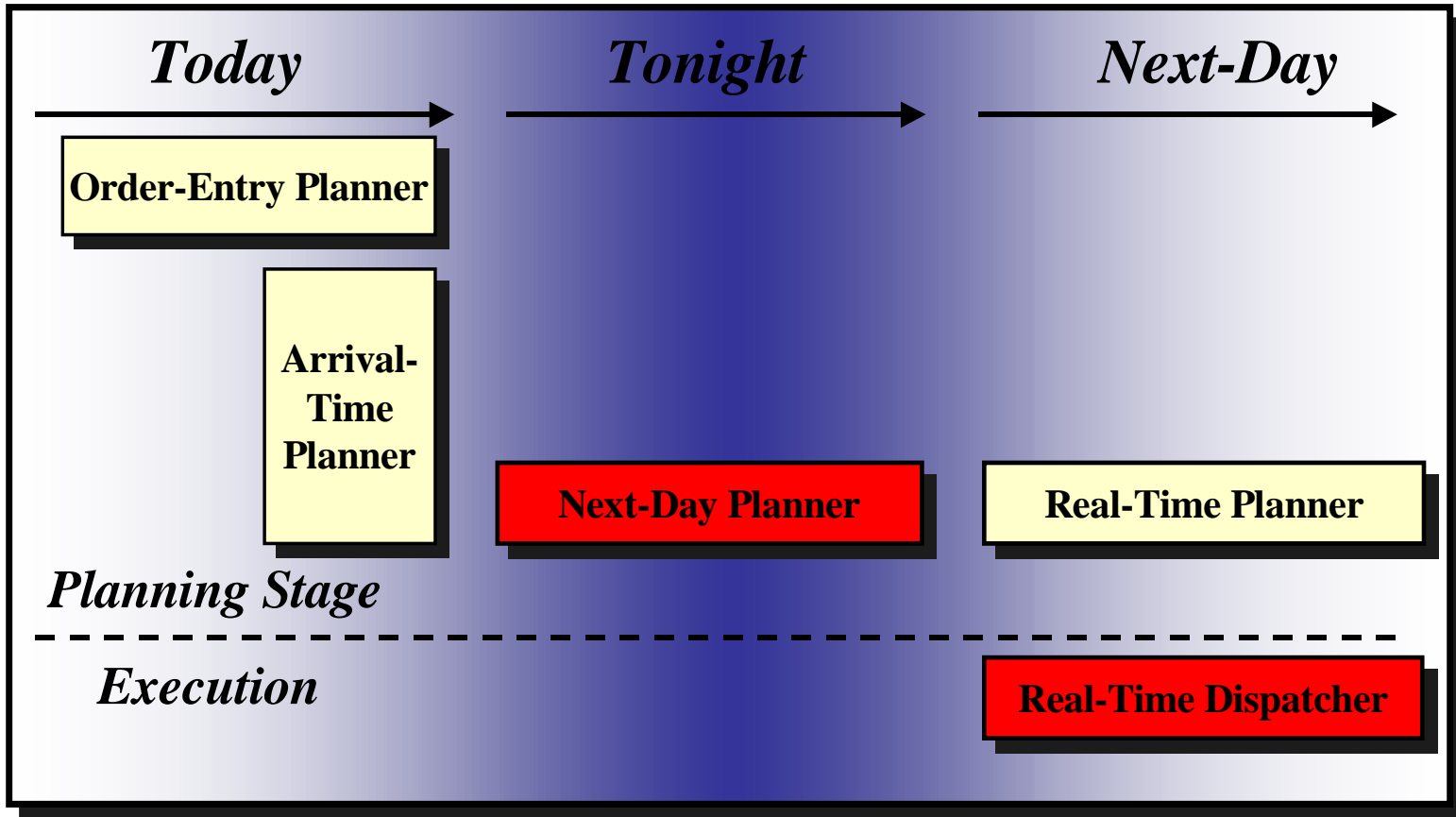


- **The characteristics of the solution**
 - **The key business benefits:**
 - Employee retention through reduced stress on dispatchers.
 - A fundamental change from Truck-Based to Demand-Based dispatching.
 - **The key OR modeling contribution: Dealing with infeasibilities.**

The Decision Support Tool



The Decision Support Tool



Key Modules

Return on Investment



- **Benefits**
 - Eliminated employee retention problems
 - Quality of schedule less dependent on dispatchers
 - Schedule is now DEMAND-based rather than TRUCK-based (estimated savings of \$750,000/year)
- **Florida Rock is expanding and promoting this application**
 - Now being deployed company wide (10x increase in trucks and plants)
 - FR is promoting industry wide as scheduling best practice

□ Model sizes:

- Next-Day Planner: 25000 cons, 200000 vars (2000 binary)
Time Window to solve = 2 hours (4 hours accepted)
- Real-Time Dispatcher: 10000 cons, 75000 vars (300 binary)
Time Window to solve = 15 seconds (30 seconds accepted)

□ Summary: Where LP/MIP technology progress made a difference:

- A. Dual simplex algorithm
- B. Heuristics in MIP

□ Next-Day Planner LPs – Solving the Root:

CPLEX 1.0 (1988)	primal	>40 hrs
CPLEX 3.0 (1994)	dual	18 mins
CPLEX 9.0 (2004)	dual	12 mins

The MIPs

Next-Day Planner MIPs – 2 hour window

Algorithm	Mean Time	First Solution
CPLEX 5.0 (1997)	5.1 hrs	4.1 hrs
CPLEX 8.1 (2003)	0.8 hrs	0.2 hrs

Real-Time Dispatching MIPs – 15 second window

Time Limit	15 secs	30 secs	60 secs
CPLEX 5.0	no feasibles	20% feasible	80% feasible
CPLEX 8.1 gaps	10.3%	1.5%	0.05%



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Fab PowerOps (FPO):

**A shop-floor scheduling solution for
front-end semiconductor Fabs**



Fab PowerOps: Overview



- **Semicon background**
- **The scheduling problem**
- **Using optimization to schedule fab production**
 - **The Diffusion module**



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Semicon Background

Semiconductor History



- **1947 Transistor invented**
 - Bardeen, Brattin, Shockley at Bell Labs
- **1958 Integrated circuit introduced – circuits on a single, planar substrate**
 - Kilby (TI), Noyce (Fairchild)
- **1960s – 90s Manufacturing processes revolutionized**
 - 1964: Gordon Moore (Fairchild) predicted device density would double every 18 months
 - Rapid price drops began in mid sixties
- **1990 – Present: Focus on production issues**
 - Automation
 - Cost control
 - Process control and efficiency

Semiconductors Today



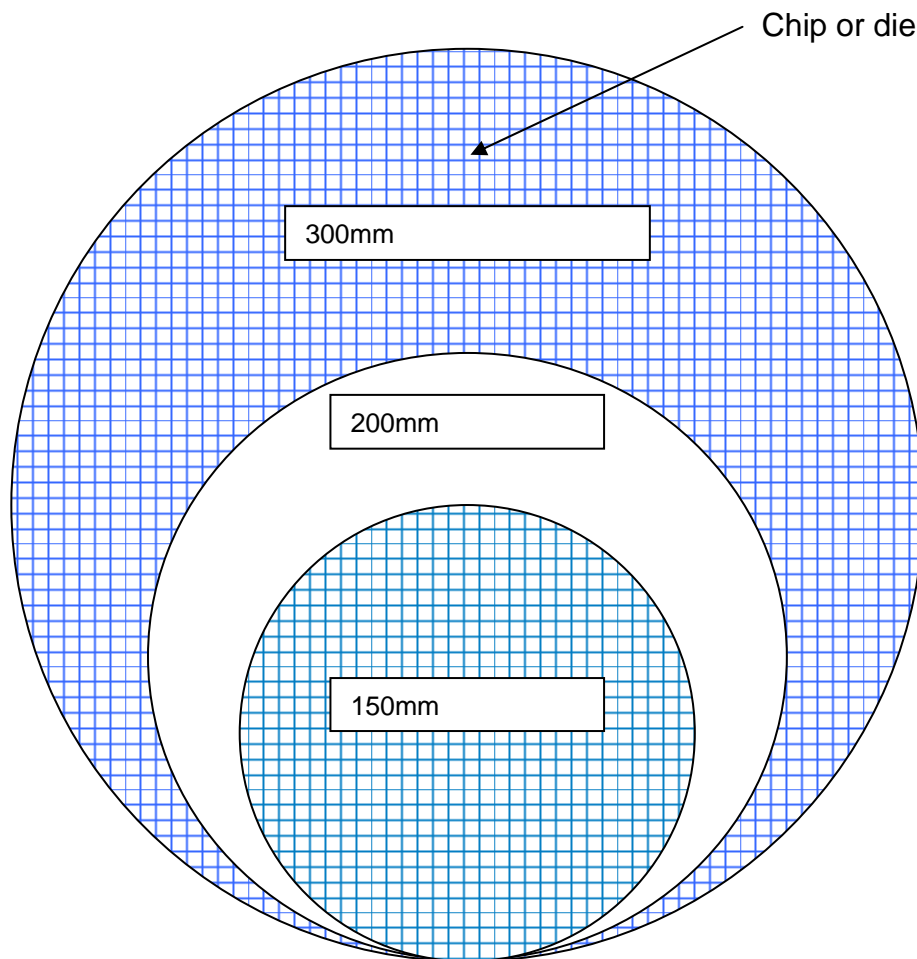
- **Leading the world in manufacturing**
 - \$200+ billion/year business
 - \$40+ billion/year consumer of other products
 - New capital expenditure leader in US manufacturing
- **Technology driven**
 - Current generation of Fabs: 300 mm (12 inches)
 - Feature sizes down to 65 nanometers
 - New Fabs cost \$3-4 billion
- **Tremendous price pressure**
 - DRAM drops by 10x every 24 months



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The Scheduling Problem

Wafer size has grown steadily



Some facts:

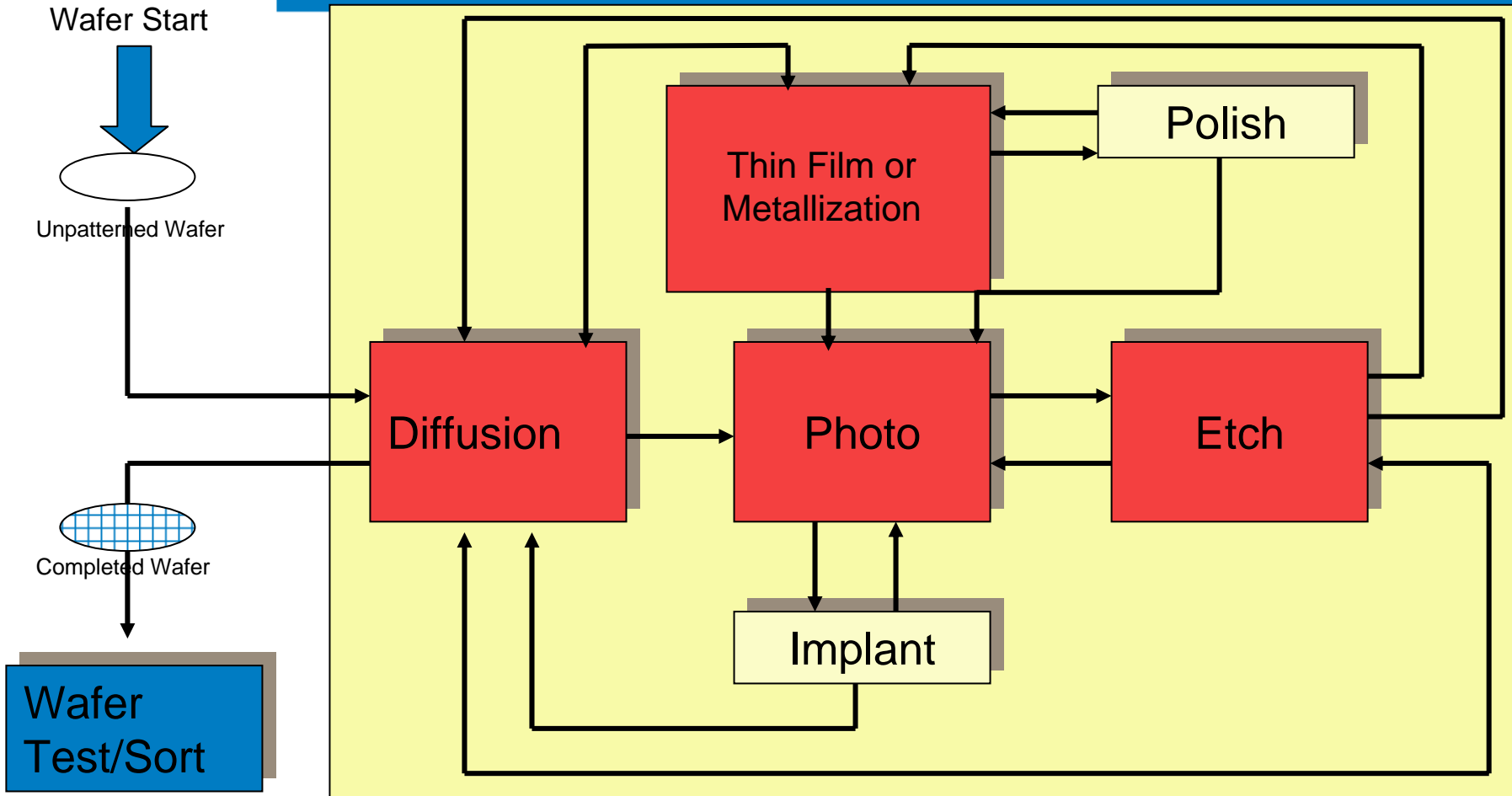
- 300 mm wafers current state-of-the-art
- 500+ chips (dies) per wafer
- Process may require over 300 steps in 30 or more “layers”
- Processing occurs in **lots** of 6-25 wafers (300 mm – held in FOUPs)
- Takes 1-3 months to process a lot

A Cyclic Fabrication Process



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Main front-end processes



The Lot Scheduling Problem



Issues

- **Efficient scheduling of lots clearly important**
- **Scheduling must be flexible and must occur in real-time**
 - **Processes are delicate**
 - **Some processing times are unpredictable**
 - **“Super Hot Lots” disrupt scheduling**

Use rules + simulation

- **Real-time dispatching rules are the norm**
 - **Typically involve 30 or more rules**
 - Evolve with time
 - Very complex
 - Order dependent
 - **Must be tuned daily, typically using simulation**
 - **Very brittle and hard to maintain**
 - Can take several weeks to adjust to process changes
- **The solution discussed here is the only optimization-based alternative to rules-based dispatching**
 - **Optimization considered better, but not practical**
 - **Operations people often recognize that “scheduling” is better than “dispatching”**

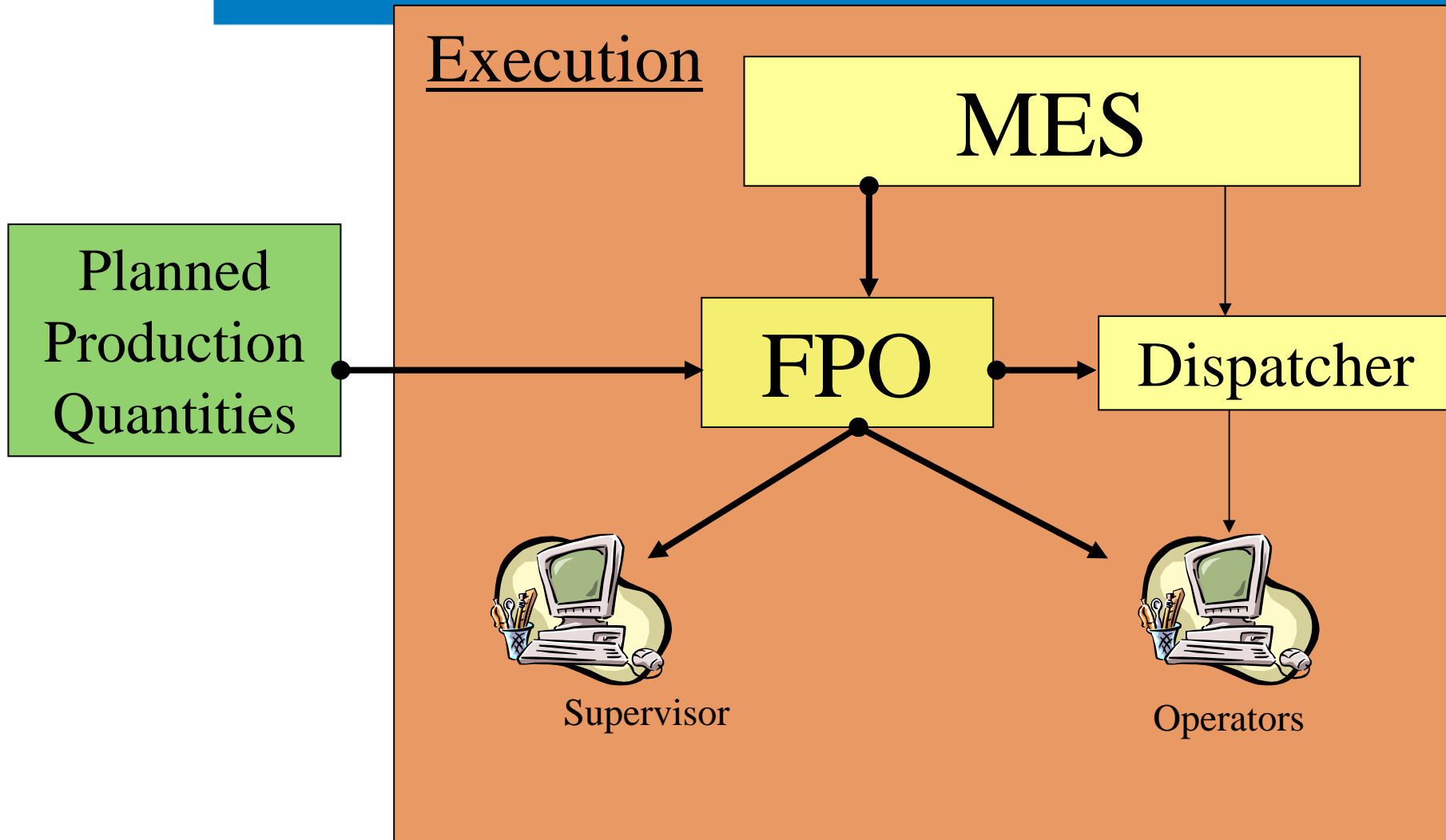


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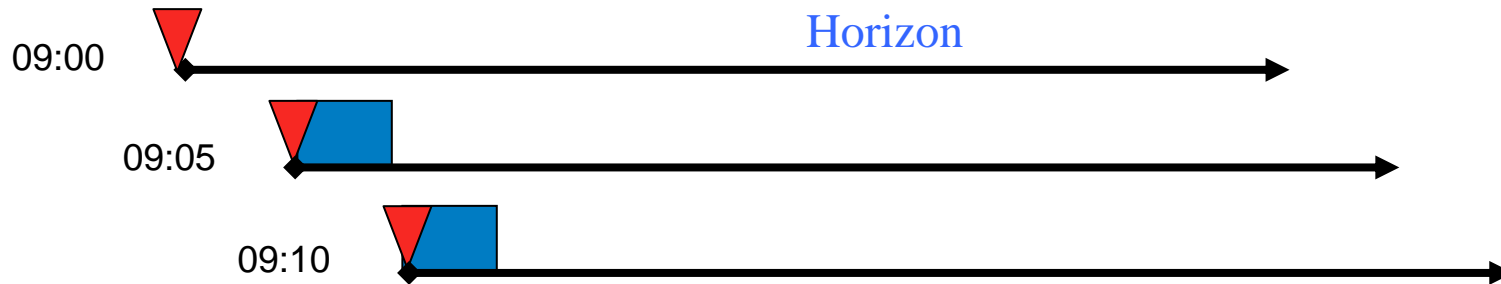
Using Optimization to Schedule FAB Production

- **Independent module for each key process**
 - **Photolithography**
 - **Diffusion (clean + tube furnace)**
 - **Etching**
 - **Thin Film (CVD)**
 - **Implant**
 - **Chemical-Mechanical Polishing (CMP)**
- **Optimization model**
 - **Decomposition**
 - **Uses Constraint Programming + Mixed-Integer Programming**
 - **Some constraints are hard, some handled with objectives**
 - Focused on modeling tools rather than processes → translates from one Fab to the next.

FPO fits into the existing software environment



FPO Working Modes (all modules)



- **Horizon usually 8 or 12 hours (one shift)**
- **Rolling horizon**
- **Runs frequently (every 5 minutes)**
- **Generates solutions in seconds**
 - **Mode 1: Beginning of the shift**
 - **Mode 2: During the shift**
- **Time fence management**

What Enables this Solution?



- **Processes have become highly automated**
 - **Particularly in 300 mm Fabs**
- **Optimization models are “local” to individual processes, not global to entire Fab**
- **New decomposition algorithm**
 - **Took three years to develop**
 - **Key features exploited: Large # of parallel machines with weak precedence constraints**
- **Advances in optimization technology**

Diffusion Model: Impossible until very recently

CPLEX 5.0 (1998):

CPLEX Error 1001: Out of memory.

Error termination, no integer solution.

Current MIP best bound = $-3.9084392492e+02$ (gap is infinite)

Solution time = **16520.82 sec.** Iterations = 24359727 Nodes = 854226

CPLEX 9.0 (present):

	Node	Left	Objective	IInf	Best Integer	Best Node	ItCnt	Gap	
	0	0	393.2257	1322		393.2257	4853		
			366.4625	1185		Cuts: 703	8483		(mostly Gomory cuts)
*	720+	672		0	348.3725	366.3402	28464	5.2%	16 seconds
*	1314+	1092		0	354.8399	366.3359	43629	3.2%	25 seconds
*	3060+	2623		0	355.9241	366.2938	94792	2.9%	59 seconds
*	4000+	2770		0	357.6452	366.2146	127312	2.4%	80 seconds
*	6056	4400		0	357.9718	365.7744	220862	2.2%	137 seconds

Time limit exceeded, integer feasible: Objective = $3.5797175137e+02$

Current MIP best bound = $3.6560278193e+02$ (gap = **7.63103, 2.13%**)

Solution time = **180.01 sec.** Iterations = 309099 Nodes = 7841 (6124)

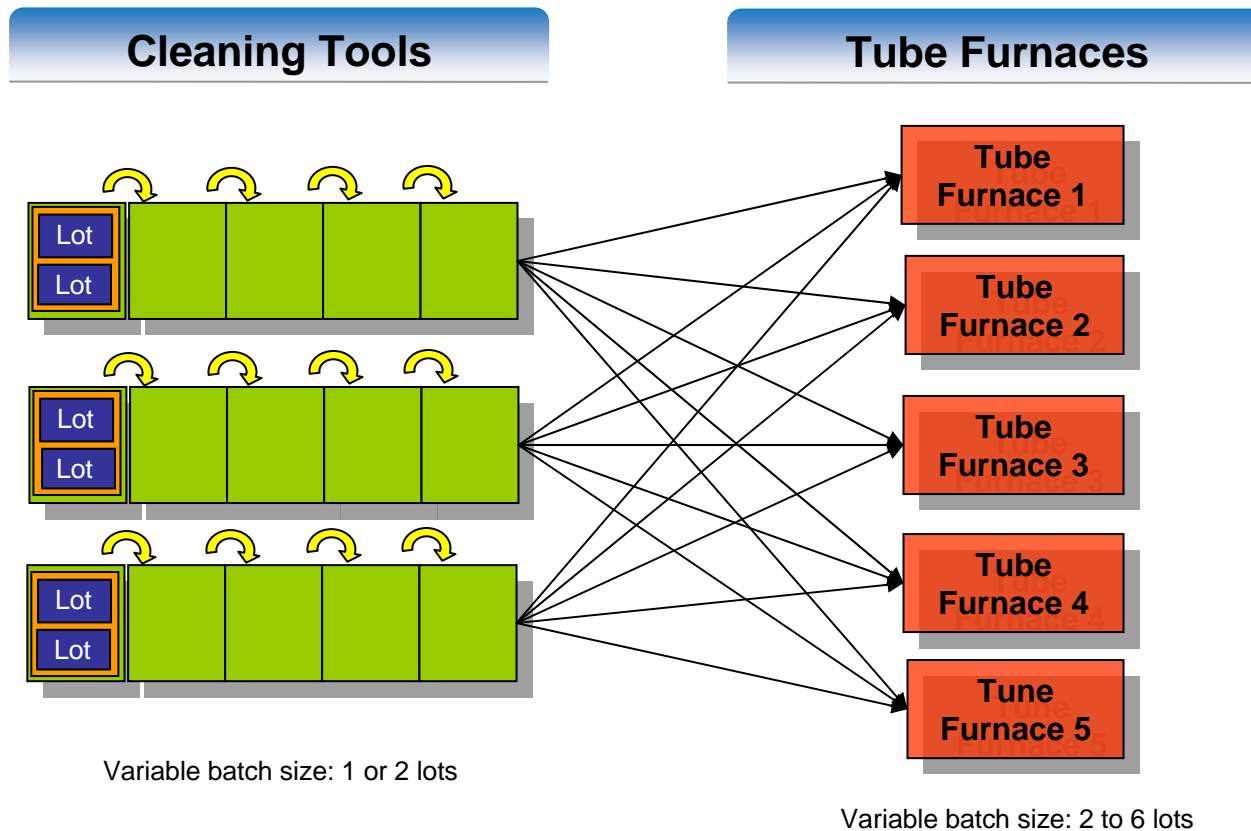


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The Diffusion Module

Most complex Fab process

- Two processes with dynamic batching and process time windows



Process Behavior

- **Tool status**
 - Production, standby, down
- **Tool pre-reservation**
- **Tool balancing**
- **Lot processing events and phases**
 - Operation start, process start, process end, cooling, unloading, operation complete
- **Lot type, class and priorities**
- **Multiple-lot FOUPs**
 - Single lot/single recipe, multiple lot/single recipe, multiple lot/multiple recipe
- **Test steps and scripts**
 - Measurement, metrology
- **Inhibits**
 - Product, route, route/operation, recipe, tool, tool/recipe, process definition/tool, module process definition
- **Hold mechanism**
- **Phase-in**

Precise Models of Diffusion Process



Tool Behavior

- **Wet tools**
- **Tube furnaces**
- **RTP furnaces**

Tool Behavior

- **Wet tool constraints**
 - **Dump-fill constraints**
 - **Picking:** Single wafer, all wafer
 - **Recipe balancing:** Alternate recipes
 - **ML/MR:** Quals
 - **Buffer Management**

Tool Behavior

- **Tube furnace constraints**
 - **Batch size constraints:** Tool level, tool recipe level, step/tool recipe level
 - **Process types:** Nitride, poly, copper anneal, cure, standard oxidation, gate oxidation, etc.
 - **Maximum number of FOUPs**
 - **Buffer Management**
 - **BTM (boat transfer module) management**

Schedule Objectives

- **(Super) Hot lot management**
- **Throughput**
- **Batch size**
- **Batch time fence**
- **Idle time**
- **Wet capacity**