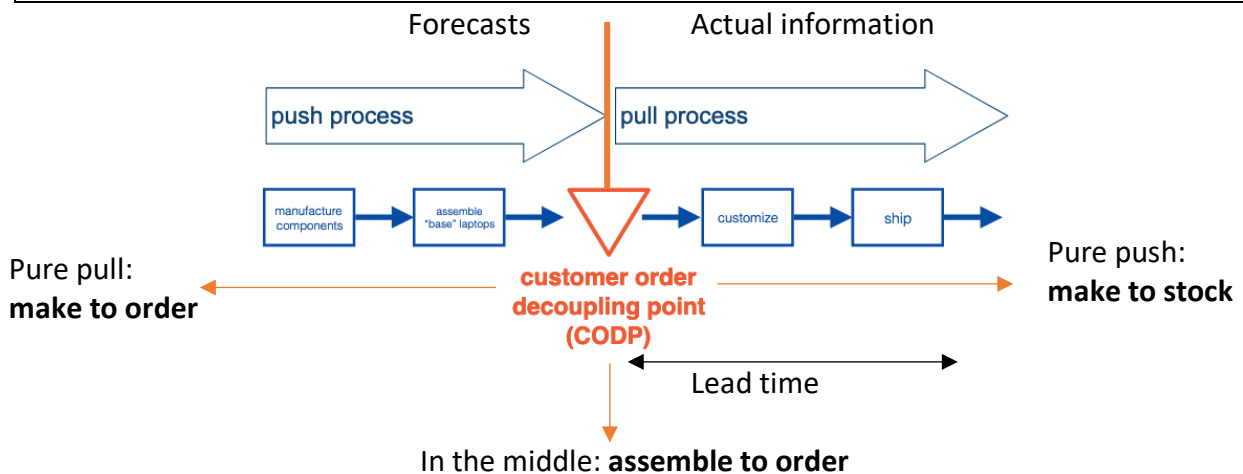


Understanding the Supply Chain



Strategic fit

Competitive strategy: price? Variety? Time? Quality?

Functional strategy: supply chain, ...

→ **ALIGNEMENT!!**

Achieving strategic fit:

1. Understand uncertainty

- Actual demand uncertainty

+ customer needs

→ **implied** demand uncertainty

Low ←————→ High

- | | |
|----------------------|-----------------------|
| ○ Easy to forecast | ○ Hard to forecast |
| ○ Low leftover risk | ○ High leftover risk |
| ○ Low product margin | ○ High product margin |

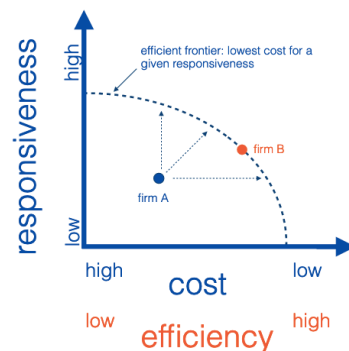
- Supply uncertainty

2. Understand SC capabilities

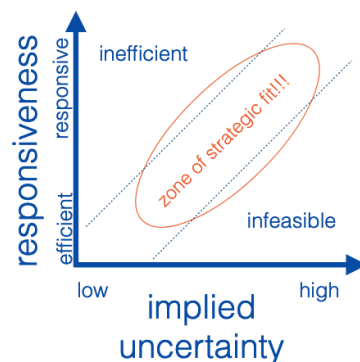
- Responsiveness ~ ability to react to uncertainty

>< Efficiency: if uncertainty is low

- Responsiveness ↑ → CODP more downstream → lead time ↓
- Careful: capacity utilization
- Trade-off:



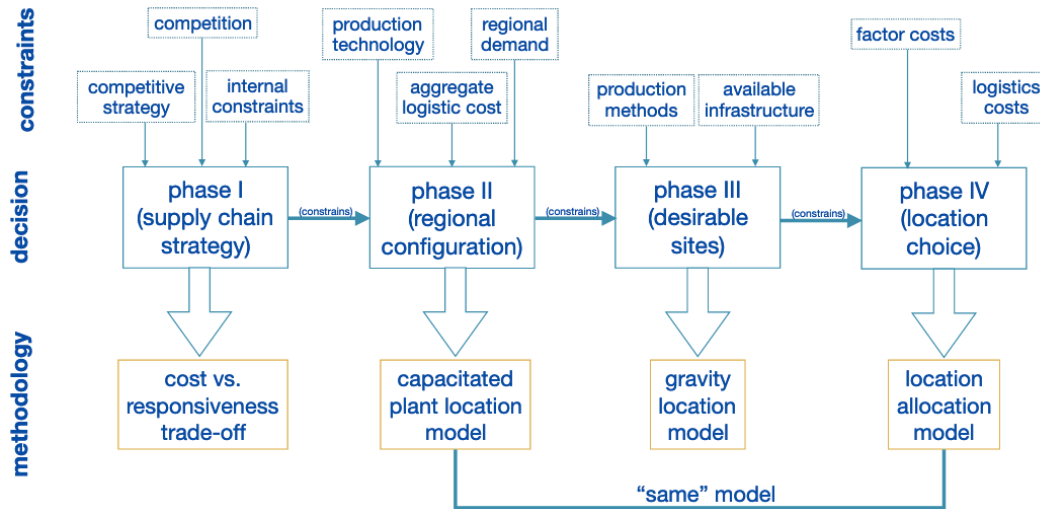
- SC strategy (responsiveness) **depends on** competitive strategy!!!



3. Redesign

Zara case → very high responsiveness

Framework for network design:



Phase I: supply chain strategy

Main trade-off:

Customer service

- Lead time
- Product variety
- Product availability
- Time to market
- Order visibility
- After-sales

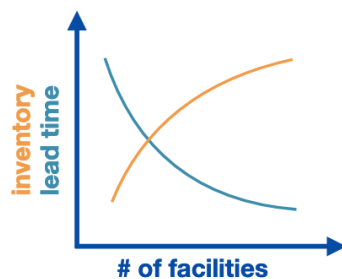
Supply chain costs

- Inventory
- Transportation
- Handling
- Facilities
- Information

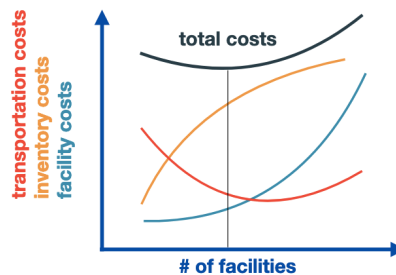
↳ Depends on competitive strategy!

Number of facilities:

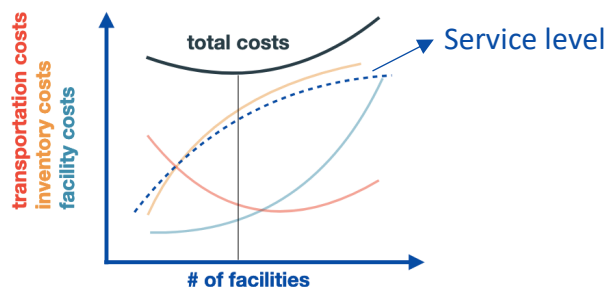
→ Impacts lead times



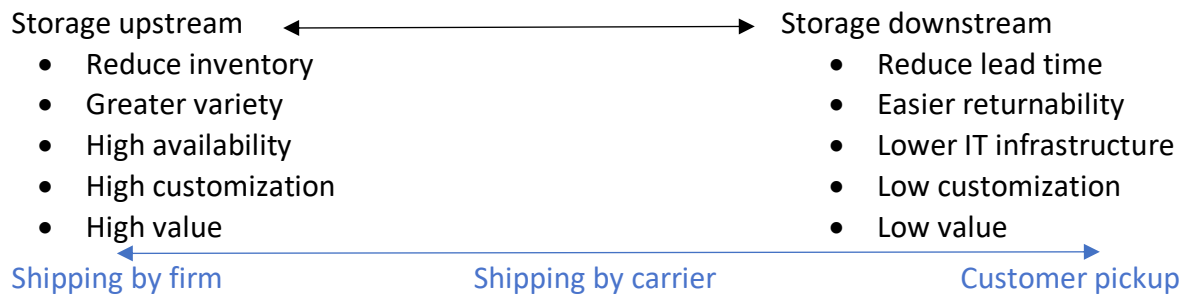
and inventory/transportation/handling/facility costs:



→ Service level:

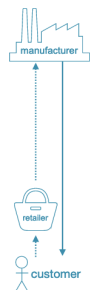


Role of intermediaries:

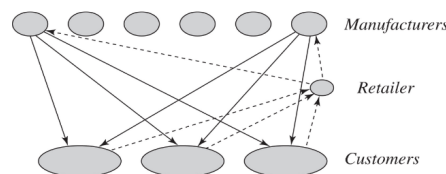


Design options:

- Manufacturer storage with direct shipping (drop-shipping)



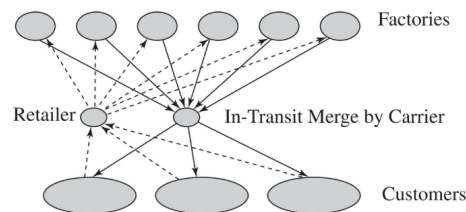
- Low response time
- High level of variety + high availability
- Low inventory costs + low handling costs



- Manufacturer storage with direct shipping and in-transit merge (= cross dock)



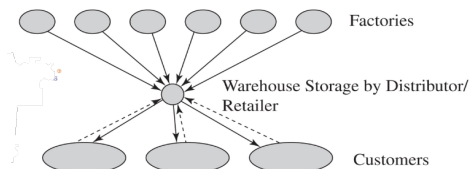
- Customer perspective: similar to drop-shipping
- Lower transportation costs than drop-shipping



- Distributor storage with carrier delivery



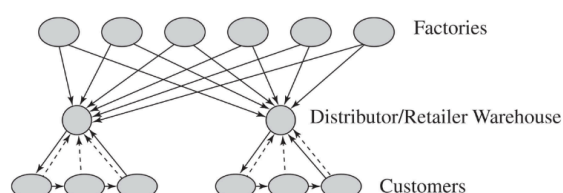
- Higher response time than manufacturer storage
- Higher cost to provide same availability as manufacturer storage
- Higher inventory costs than manufacturer storage



- Distributor storage with last-mile delivery



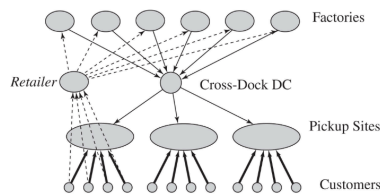
- Higher service level (same/next-day delivery)
- Very expensive!



- Manufacturer/distributor storage with customer pickup



- High response time
- Low transportation cost
- Low customer experience



- Retail storage with customer pickup



- Lower product variety
- High inventory!

+ internet

- Online sales
- **Omnichannel logistics**

Phase II: regional configuration

→ allocate capacity and demand

Linear programming:

1. Define variables
2. Define objective function (minimize total relevant costs)
3. Constraints
4. Solution
 - If $\sum K = \sum D$ (**balanced** transportation problem) → all constraints binding
 - If $\sum K > \sum D$ (**unbalanced** transportation problem) → unused supply
 - If $\sum K < \sum D$ (**unbalanced and infeasible** problem) → unmet demand

Phase III: desirable sites

Load-distance model

- Finite alternatives: supply, demand and potential locations as points in a plane
- Minimize load x distance

Gravity model

- Supply and demand as points in a plane
- Optimal location anywhere in the plane
- Minimize $\sum d_n D_n F_n$

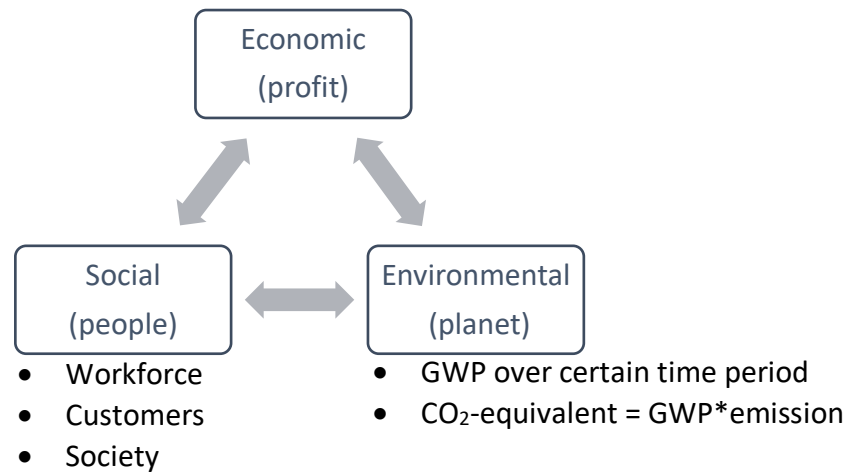
Factor-rating

- Finite alternatives to compare via relevant factors
- Assign weight to each factor

Phase IV: location choice

Same model as phase II

Sustainability

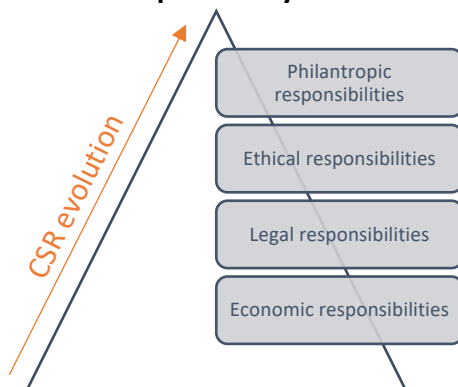


Sourcing

Trends:

- Globalization: few hubs
 - Specialization: multiple tiny nodes
 - Complex, long supply chains
 - Lean processes
- High risk!

Corporate Social Responsibility:



Socially Responsible Purchasing:

- Goods, services, capabilities and knowledge
 - Sustainable purchasing: procedures and guidelines
- >> Responsible purchasing: personal responsibility

Sustainability adoption:

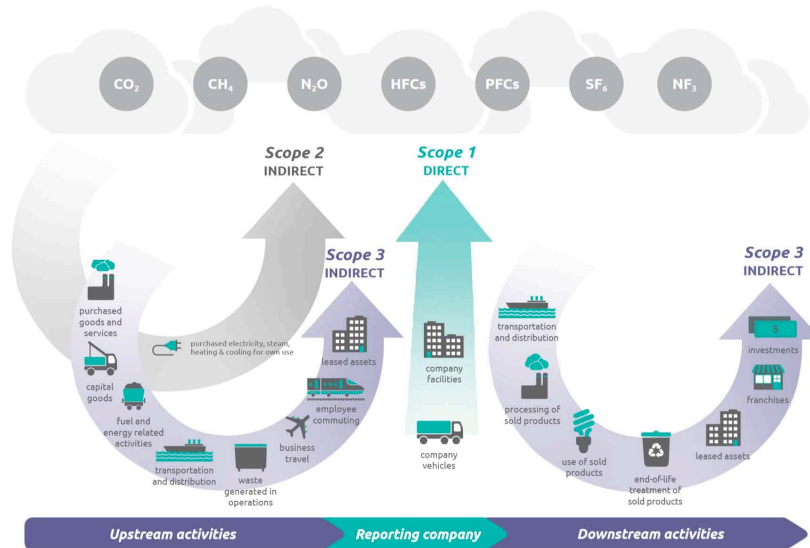


SCND

Measuring carbon footprint...

- **Cradle-to-grave**
- **Well to tank** (for transport)
- **Tank to wheels** (for transport)

Greenhouse Gases Protocol scopes



- **Scope 1**
 - Direct emissions
 - Own company
- **Scope 2**
 - Indirect emissions
- **Scope 3**
 - All other emissions

→ ASSUMPTIONS have a HUGE impact on results!!!

Estimating emissions: e.g. Network for Transport Measures

Inventory control: deterministic demand

Joint replenishment

- **Full aggregation**
= order everything together
- **Power-of-two policy**
= all items ordered in intervals of $2^k T_L$
 1. Define baseline interval T_L
 2. Find smallest k so that $TC(2^{k+1}T_L) \geq TC(2^k T_L)$
- **Tailored aggregation**
= optimize frequency
 1. Determine the fast mover: for all products optimal $\bar{n}_i \rightarrow$ choose maximum \bar{n}
 2. Determine multiplier of slow movers, *without S*: $\bar{n}_i \rightarrow \bar{m}_i \rightarrow m_i$
 3. Adjustment for fixed cost: new $n \rightarrow n_i$

Inventory control: stochastic demand

Forecasting

= finding underlying structure in demand based on past information!

- Basis for planning
- Relevant for both push **and** pull processes
- **Always** wrong!!
More accurate if ...
 - ... aggregated
 - ... short term
 - ... less demand uncertainty

Time-series methods: observed demand = systematic component + random component

↳ Underlying structure in demand:
this is what we want to forecast

= Level + Trend + Seasonality + ε

- Multiplicative methods
Forecast = $L * T * S$
- Additive methods
Forecast = $L + T + S$
- **Mixed methods**
Forecast = $(L + T) * S$

Static methods

→ **Parameters** are fixed (forecast itself ≠ static!!)

Mixed model: $F_t = (L + Tt)S_i$

1. Collect data
2. De-seasonalize historical demand
 - a. Determine periodicity

Period	Demand	Lag 1	Lag 2	Lag 3	...
1	D_1				
2	D_2	D_1			
3	D_3	D_2	D_1		
4	D_4	D_3	D_2	D_1	
5	D_5	D_4	D_3	D_2	...
...
Correlation		$\text{Corr}(D, L_1)$	$\text{Corr}(D, L_2)$	$\text{Corr}(D, L_3)$...

If correlation is high → periodicity $p = \text{lag value}$

- b. Determine average demand \bar{D}_t over p periods
3. Estimate L and T
E.g. via linear regression: $\bar{D}_t = L + Tt + \varepsilon_t \rightarrow E[\bar{D}_t] = L + Tt$
4. Estimate S_i
 - a. For every period: $\bar{S}_t = \frac{D_t}{E[\bar{D}_t]}$
 - b. S_i = average over each season
5. Calculate F_t

Adaptive methods

→ **Parameters** are updated in time: forecast changes in time!!

1. (Initialise)
2. Compute forecast
3. Observe demand
4. Recompute parameters

Moving average: $F_{t+x} = L_t$

- Average over past N periods
- Throws away all data from before $t-N$

Simple exponential smoothing: $F_{t+x} = L_t$

- Updates forecast based on previous forecast +/- previous error
- Sum of previous demands, where past demand becomes less and less relevant
- Initialize L_0
- Smoothing parameter $\alpha \rightarrow 1$: more importance to recent observations
 $\rightarrow 0$: more importance to past data

Without new information, all future forecasts are the same!!

Holt's model/double exponential smoothing: $F_{t+x} = L_t + xT_t$

- Initialize L_0 and T_0
- Smoothing parameters α and β

Winter's model/triple exponential smoothing: $F_{t+x} = (L_t + xT_t)S_{t+x}$

- Initialize L_0 , T_0 and S_i
- Smoothing parameters α , β and γ

Future forecasts no longer constant!!

Measuring forecast error:

- Forecast error $\varepsilon_t = F_t - D_t$
- Absolute deviation $A_t = |\varepsilon_t|$
- **Mean squared error** $MSE = \frac{1}{n} \sum_{t=1}^n \varepsilon_t^2$
- Mean absolute deviation $MAD = \frac{1}{n} \sum_{t=1}^n A_t$
→ if errors are normally distributed: $\sigma_F = 1.25MAD$
- Mean absolute percentage error $MAPE = \frac{1}{n} \sum_{t=1}^n \left| \frac{\varepsilon_t}{D_t} \right| 100$
- Systematic over/under estimations $BIAS = \sum_{t=1}^n \varepsilon_t$
- Tracking signal $TS = BIAS/MAD$
→ can compare different forecasts; between -6 and 6 OK

Aggregate production planning

Assumption: forecast = true (~ deterministic demand)

Include uncertainty via e.g. scenario analysis

Aggregate decisions necessary because they **take time!**

↳ Aggregate unit

- ! Additional constraints = boundary conditions
- → Net Demand = demand incl. beginning and end inventory

Chase strategy

= production follows demand: $P_t = D_t$

Evaluation:

- + Minimise inventory/backlog: $I_t = I_{t-1} + P_t - D_t = I_{t-1} \approx 0$
- Difficult/expensive in practice

→ Useful when inventory/backlog costs are high

Level strategy

= maintain constant output rate: $P = \frac{\sum_t \text{Net Demand}}{\text{nr of periods}}$

If no backorders allowed:

- Determine **"worst" period**, i.e. period with highest average cumulative net demand
- $P = \text{average cumulative N.D. of "worst" period} = \frac{\text{Cumulative Net Demand "worst" period}}{\text{"worst" period}}$

Evaluation:

- + Stability of capacity/workforce
- Service level (= how much fulfilled without backlog) may suffer
- Difficult to estimate true cost of inventory/backlog

→ Useful when capacity changes are expensive

Mixed strategy

= tailored: use combination of capacity, inventory and backlog

- + Only strategy that **minimises costs!**
- Difficult to calculate: **linear programming**
- Not very intuitive

Supply Chain Coordination: deterministic demand

No coordination: **double marginalization** → sub-optimal profits

→ Local optimisation ≠ global optimisation

→ Coordination: **all firms happy + global optimum!**

Discounts

- **Volume based discount**

Set C_r such that

- Retailer buys D_{SC}^* → discount if $D_r \geq D_{SC}^*$

- Profit retailer ≥ profit of individual optimum

→ $\pi_r = (p - \text{discounted } C_r) \cdot D_{SC}^* \geq \text{profit of individual optimum}$

- **Two-part tariff**

Franchise fee & $C_r = C_m$

- Profit manufacturer = franchise fee ≥ profit of individual optimum

- Profit retailer = profit SC* – franchise fee ≥ profit of individual optimum

- **Lot-size based quantity discount**

If price cannot be set by retailer (e.g. commodity market)

- Retailer must buy Q_{SC}^* → discount if $Q_r = xQ_{SC}^*$

- Cost retailer ≤ cost of individual optimum

→ $TC_r = D \cdot \text{discounted } C_r + \frac{D}{Q_{SC}^*} \cdot S_r + \frac{Q_{SC}^*}{2} \cdot h \cdot \text{discounted } C_r \leq \text{cost of individual optimum}$

Supply Chain Coordination: stochastic demand

Coordination: **all firms happy + global optimum!**

The newsvendor model

Trade-off: overage ($Q > D$) ↔ underage ($Q < D$) } Optimum: $C_o \Pr(D \leq Q^*) = C_u \Pr(D \geq Q^*)$
 ! $E[OH] > 0$ AND $E[SF] > 0$!

$$\Pr(D \leq Q^*) = \frac{C_u}{C_u + C_o} = \text{critical ratio}$$

- If $CR < 0,5$ → rather have underage → $Q^* < \mu$
- If $CR = 0,5$ → $\Pr[\text{overage}] = \Pr[\text{underage}] = 0,5$ → $Q^* = \mu$
- If $CR > 0,5$ → rather have overage → $Q^* > \mu$

Contracts

- Increase S_r to increase Q_r : **buyback contract**

Set C_r and buyback price b such that

- $\text{Retailer buys } Q_{SC}^*: \Pr(D \leq Q_{SC}^*) = CR_{SC}^* \rightarrow b = \text{function of } C_r$
- $\pi_r \geq \pi_r^* \rightarrow C_r = \text{function of } b$
- $\pi_m \geq \pi_m^* \rightarrow \pi_r \leq \pi_{r2} \rightarrow C_r = \text{function of } b$

\rightarrow A **sustainable contract** will have $C_{r1} \leq C_r \leq C_{r2}$ and $b_1 \leq b \leq b_2$

!!! There exists a b for every C_r that maximises π_{SC} ,

BUT this does NOT necessarily make **manufacturer AND retailer happy !!!**

- Decrease C_r to increase Q_r : **revenue sharing contract**

$\rightarrow C_m = C_r \rightarrow \pi_m = 0$

\rightarrow Lower C_r AND require percentage $(1-r)$ of revenue

- $CR = CR_{SC}^* \Leftrightarrow Q_r^* = Q_{SC}^* \rightarrow r^* = \text{linearly increasing in } C_r \rightarrow \pi_r = \text{function of } C_r$
- $\pi_r \geq \pi_r^* \rightarrow C_{r1} \leq C_r$
- $\pi_m \geq \pi_m^* \rightarrow \pi_r \leq \pi_{r2} \rightarrow C_r \leq C_{r2}$

Note: buyback contract \approx revenue sharing contract

- **Real options contract**

1. Retailer buys Q call options at price C_r

2. Each option can be exercised at price E

\rightarrow **Retailer has no physical inventory left!** ($E[OH]$ = unexercised options)

\rightarrow Manufacturer is left with excess inventory/capacity

- $CR = CR_{SC}^* \Leftrightarrow Q_r^* = Q_{SC}^* \rightarrow E^* \rightarrow \pi_r = \text{function of } C_r$
- $\pi_r \geq \pi_r^* \rightarrow C_{r1} \leq C_r$
- $\pi_m \geq \pi_m^* \rightarrow \pi_r \leq \pi_{r2} \rightarrow C_r \leq C_{r2}$