

A stochastic programming approach for planning remanufacturing activities under uncertain returns and demand forecasts

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Plan

- 1 Introduction
- 2 State of the art
- 3 Deterministic optimization problem
- 4 Stochastic optimization problem
- 5 Preliminary computational results
- 6 Conclusion and perspectives

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Context

Circular economy

"In a world with growing pressures on resources and the environment, the EU has no choice but to go for the transition to a resource-efficient and ultimately regenerative circular economy."

Manifesto for a Resource Efficient Europe, European Commission, 2012

Context

Circular economy

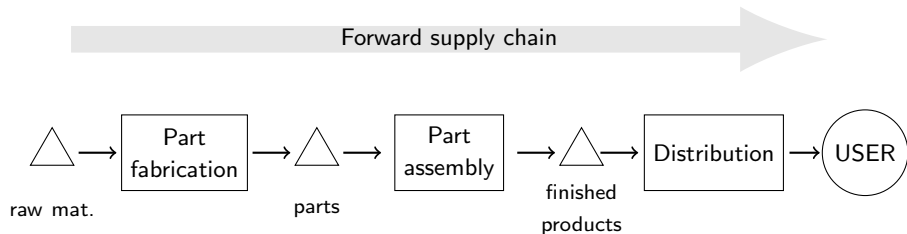
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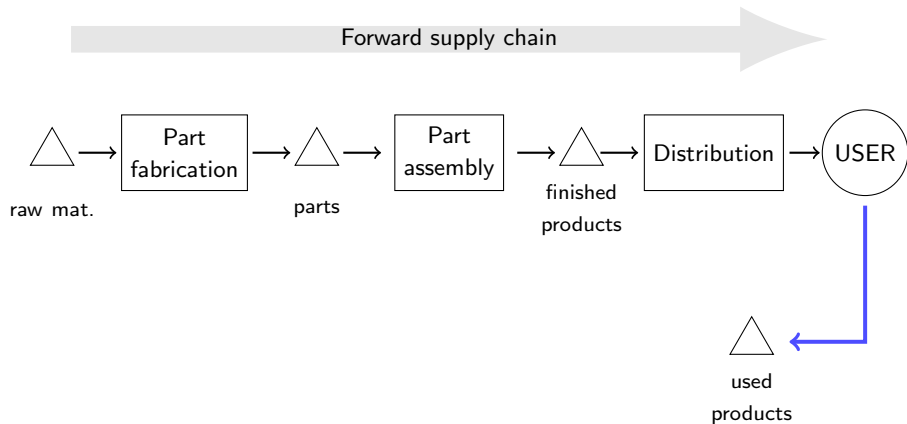
Reverse supply chains

- One possible way towards developing a circular economy
- Set of logistics and rehabilitation activities to transform end-of-life products returned by customers into once again usable products
- Main advantages:
 - Reduction in waste generation and environmental pollution
 - Decrease of natural resource consumption

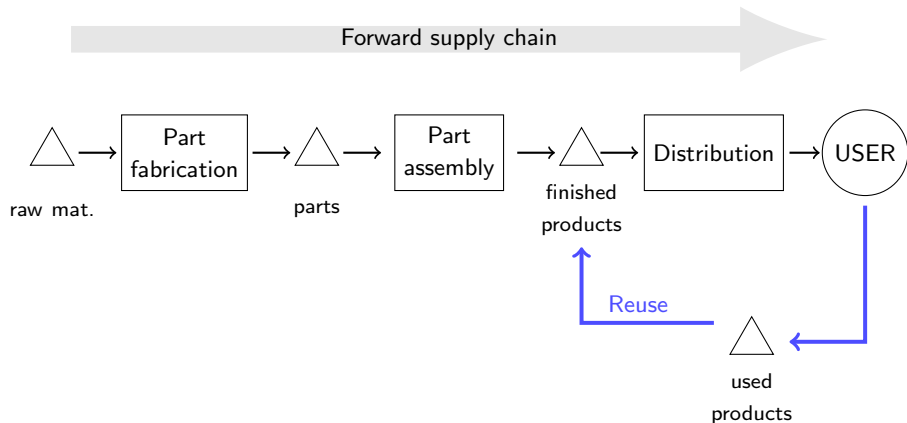
Reverse logistics activities



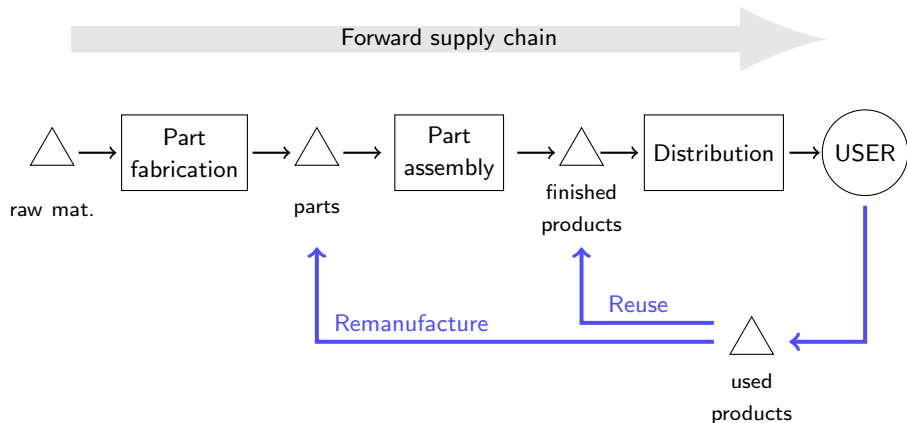
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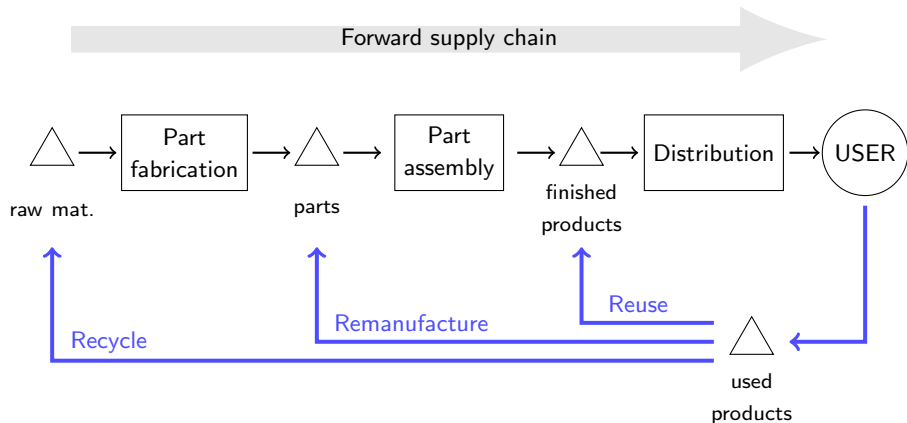
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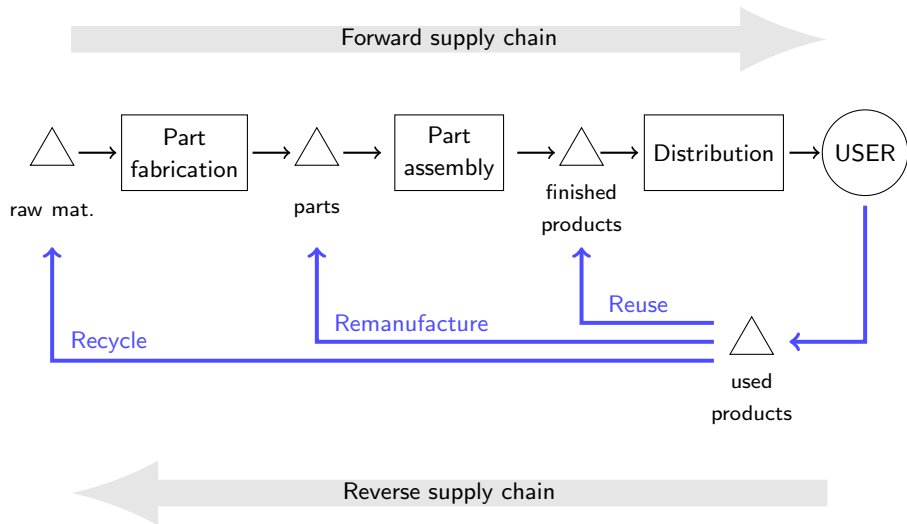
Reverse logistics activities



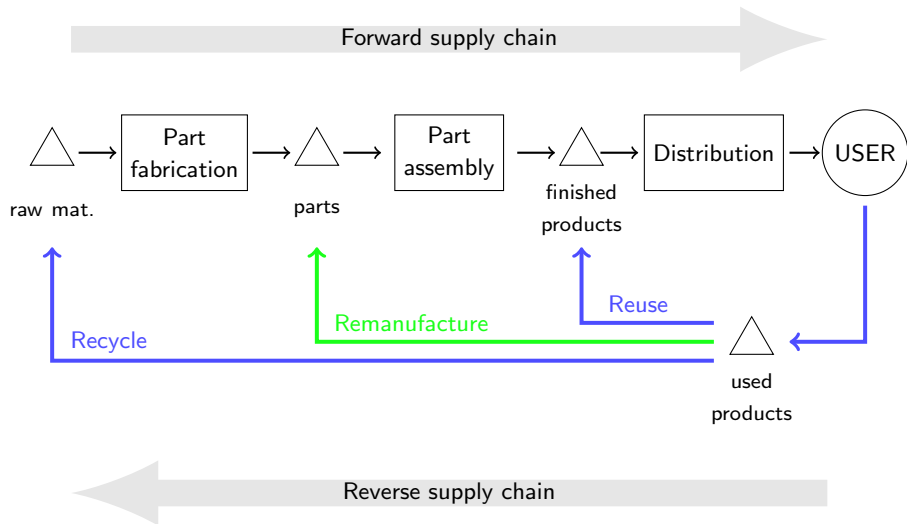
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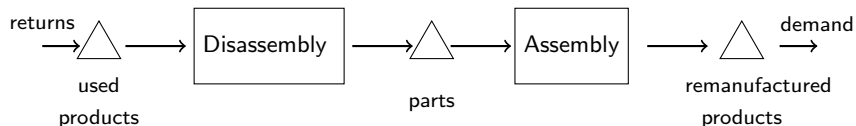


Reverse logistics activities



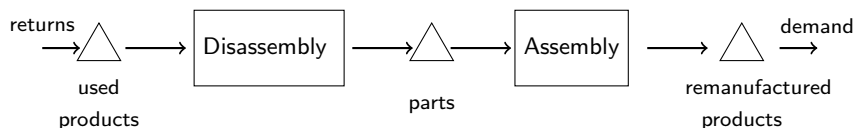
Remanufacturing planning

Remanufacturing system



Remanufacturing planning

Remanufacturing system



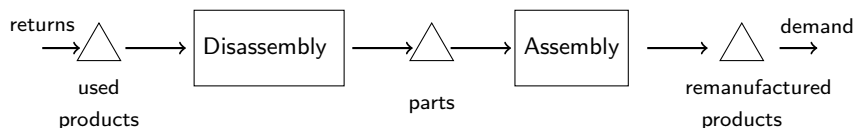
Aggregate production planning

Decide how many:

- used products to disassemble
- remanufactured products to assemble

Remanufacturing planning

Remanufacturing system



Aggregate production planning

Decide how many:

- used products to disassemble
- remanufactured products to assemble

so as to:

- satisfy customer demand
- respect technical constraints: capacity, bill of materials, inventory balance
- minimize total production costs.

Uncertain returns/demand

A specific feature in reverse logistics

No binding agreement with the end users to manage product returns

Lack of control on the product returns quantity and quality

→ High level of uncertainty in the input data of the optimization problem

[Fleischmann *et al.*, 1997]

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Our proposal

A two-stage stochastic programming approach to take into account the uncertainty on:

- returns quantity / quality
- customer demand

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Overview of the literature

Aggregate production planning for remanufacturing

- General literature reviews
[Aksali and Cetinkaya, 2011] , [Lage and Godinho, 2012]
- Deterministic optimization problems
[Jayaraman, 2006] , [Qu and Williams, 2008]
[Corominas *et al.*, 2012] , [Fall *et al.*, 2013] , [Han *et al.*, 2013]
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Classification based on

- the production planning problem
- the uncertainty modeling
- the solution approach

Classification (1)

Production planning problem

	Products	Periods	Prod. stages	Activities
Li <i>et al.</i> , 2009	1	T	1	Remanuf.
Shi <i>et al.</i> , 2010	N	1	1	Manuf./Remanuf.
Denizel <i>et al.</i> , 2010	1	T	2	Remanuf.
Rouf and Zhang, 2011	N	1	2	Remanuf.
Mahapatra <i>et al.</i> , 2012	N	T	1	Manuf./Remanuf.
Our work	N	T	2	Remanuf.

- + A realistic production planning problem
- Focus only on the remanufacturing activities

Classification (2)

Uncertainty modeling

	Demand	Returns quantity	Returns quality	Uncertainty representation
Li <i>et al.</i> , 2009	X	X		Discrete random variables
Shi <i>et al.</i> , 2010	X	X		Continuous random variables
Denizel <i>et al.</i> , 2010			X	Scenario tree
Rouf and Zhang, 2011	X	X		Continuous random variables
Mahapathra <i>et al.</i> , 2012			X	Deterministic
Our work	X	X	X	Scenarios

- + Uncertainty on demand / returns quantity / returns quality
- Simple representation via a set of discrete scenarios

Classification (3)

Solution approach

	Backlog Lost sales	Probabilistic constraints	Recourse actions	Solution approach
Li <i>et al.</i> , 2009	X			Stochastic dynamic prog.
Shi <i>et al.</i> , 2010	X	X		Single-stage stoch. prog.
Denizel <i>et al.</i> , 2010			X	Multi-stage stoch. prog.
Rouf and Zhang, 2011	X	X		Single-stage stoch. prog.
Mahapathra <i>et al.</i> , 2012			X	Recourse problem only
Our work	X		X	Two-stage stoch. prog.

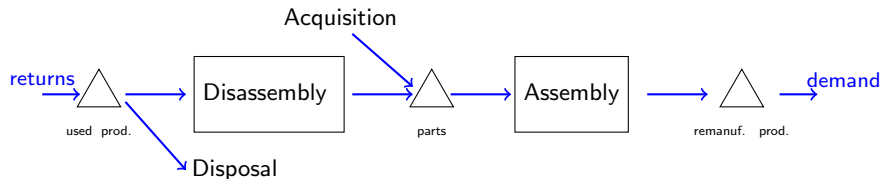
- + Definition of recourse actions applicable in practice
- Consideration of only 2 stages in the decision process

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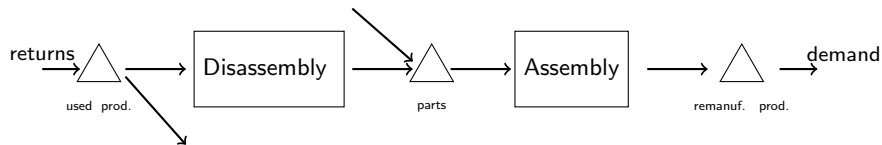
Problem description (1)

Product flows



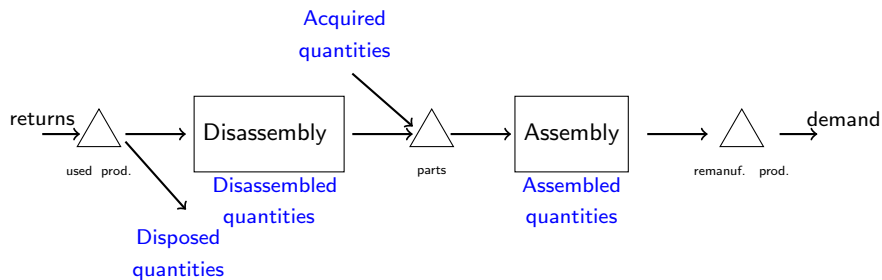
Problem description (3)

Main decisions



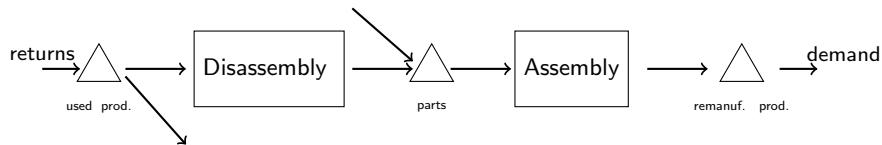
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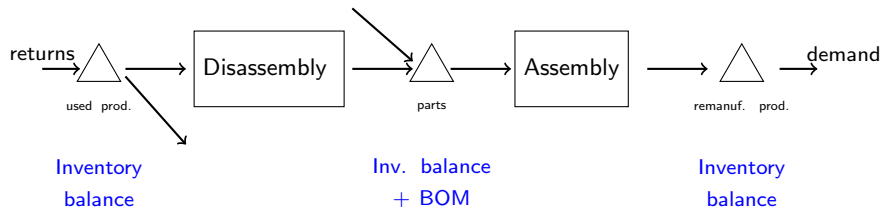
Problem description (3)

Constraints



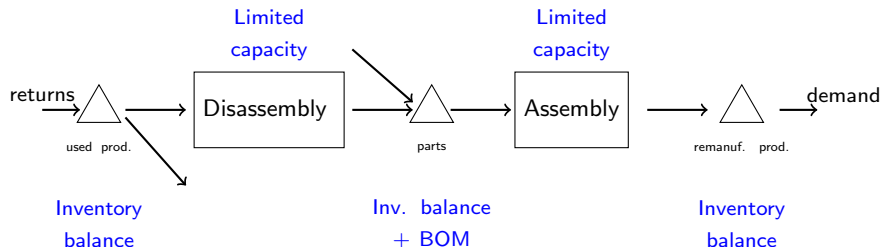
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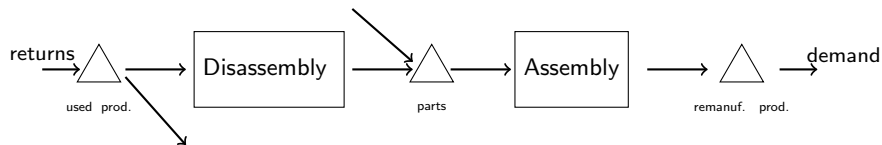
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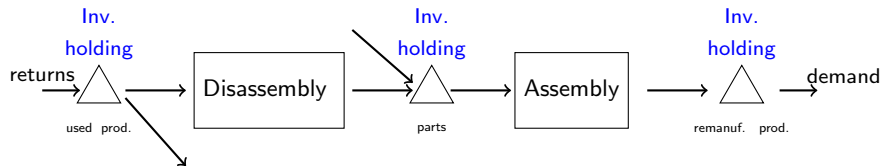
Problem description (4)

Costs



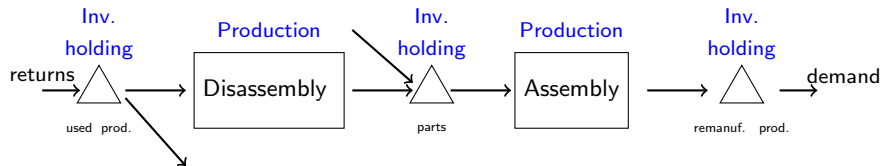
Problem description (4)

Costs



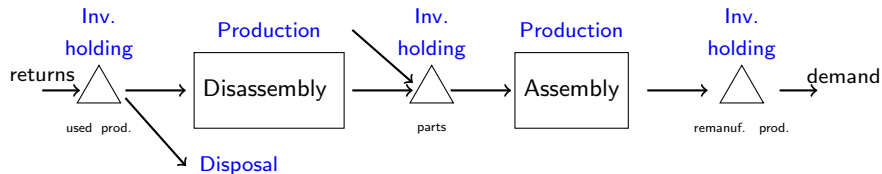
Problem description (4)

Costs



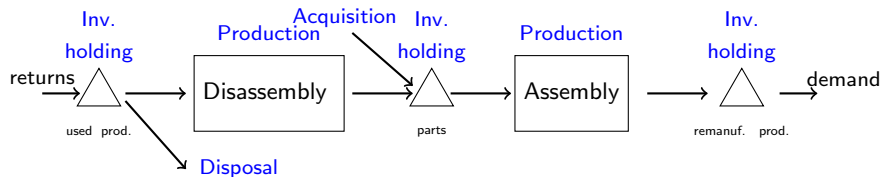
Problem description (4)

Costs



Problem description (4)

Costs



Problem description (5)

Quality of the returned products

Use of a finite set of discrete quality levels

For each returned product type and each quality level:

- A disassembly bill-of-material
- A per unit disassembly capacity consumption
- A per unit disassembly cost

[Jayaraman, 2006]

Assumption

The returned products have already been sorted and assigned to a quality level.

Linear programming formulation

Minimize total production costs

- = disassembly/assembly costs
- + inventory holding costs
- + new parts acquisition costs
- + used products disposal costs

subject to :

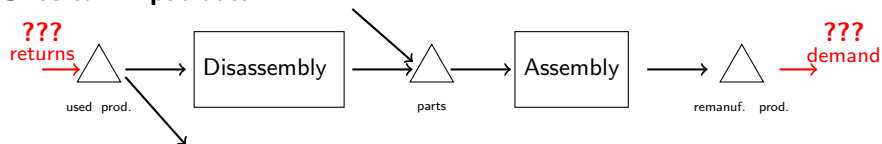
- capacity constraints
 - disassembly
 - assembly
- inventory balance + BOM constraints
 - returned products
 - parts
 - remanufactured products

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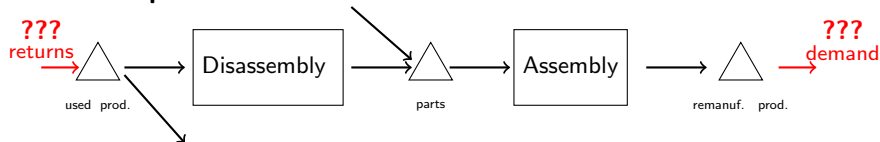
Uncertainty

Uncertain input data



Uncertainty

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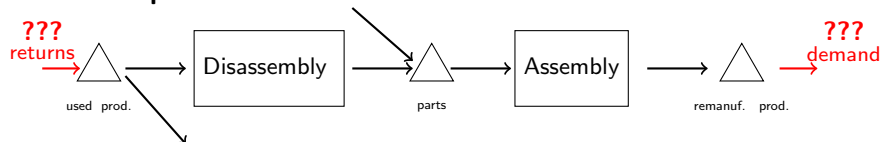


Consequences

- Demand
 - Impact limited to the remanufactured product inventory

Uncertainty

Uncertain input data



Consequences

- Demand
 - Impact limited to the remanufactured product inventory
- Returns quantity and quality
 - Disorganization of the disassembly and assembly production plan

Uncertainty representation

Continuous random variables

- Uncertainty mostly due to forecasting errors
- Forecasting errors = Normally distributed random variables
- Terms involving integrals in the mathematical formulation
- → Computational difficulties

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Continuous random variables

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- → Computational difficulties

A finite set of discrete scenarios

- Monte Carlo sampling of the continuous random variables
- A scenario = a possible realization of all uncertain parameters
 - For each period:
 - demand for each remanufactured product
 - returned quantity for each quality level of each used product
- The larger the sample size, the better the approximation.

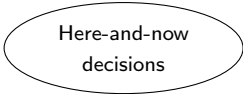
Two-stage stochastic programming approach

A two-stage decision process

Two-stage stochastic programming approach

A two-stage decision process

- 1 "Here-and-now" decisions
Before the realization of the uncertain parameters
Decisions common for all scenarios

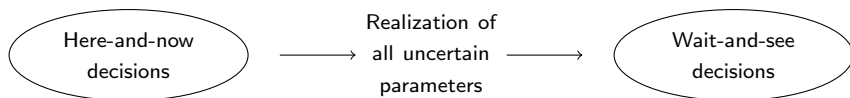


Here-and-now
decisions

Two-stage stochastic programming approach

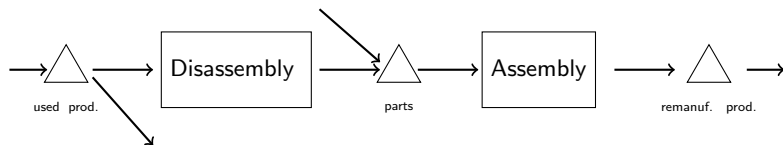
A two-stage decision process

- 1 "Here-and-now" decisions
Before the realization of the uncertain parameters
Decisions common for all scenarios
- 2 "Wait-and-see" decisions / Recourse actions
After the realization of the uncertain parameters
Decisions specific to each scenario



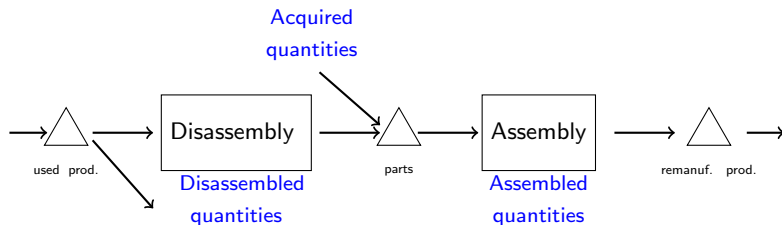
Two-stage stochastic programming approach

First stage decisions



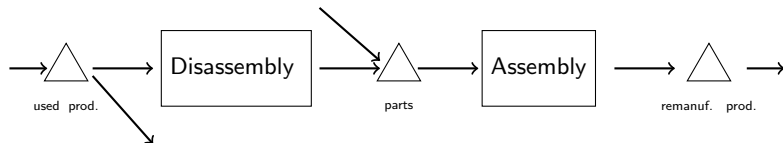
Two-stage stochastic programming approach

First stage decisions



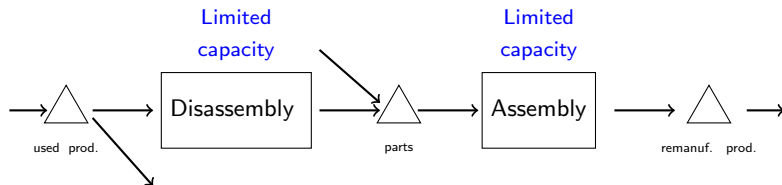
Two-stage stochastic programming approach

First stage constraints



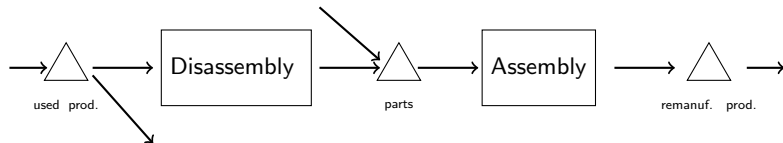
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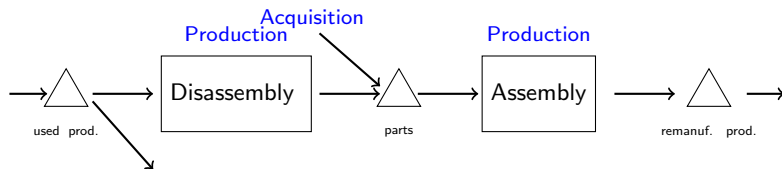
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First stage costs



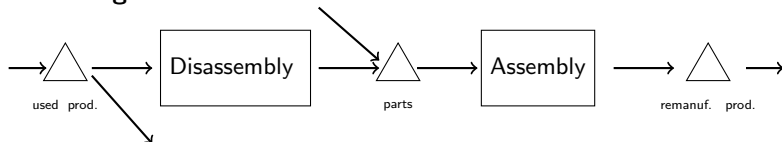
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First stage costs



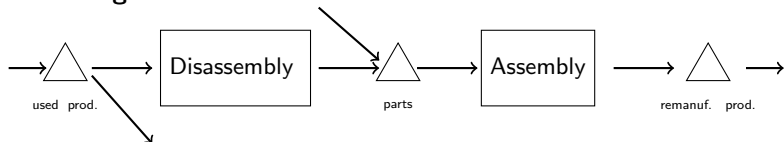
Two-stage stochastic programming approach

Second stage decisions



Two-stage stochastic programming approach

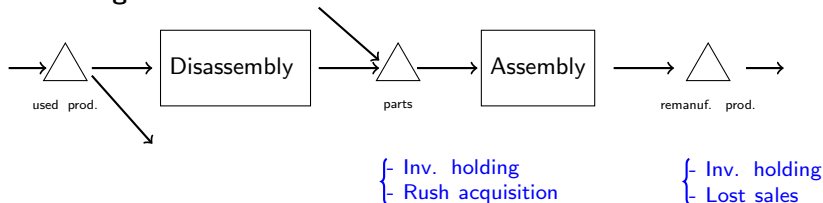
Second stage decisions



{ Inv. holding
{ Lost sales

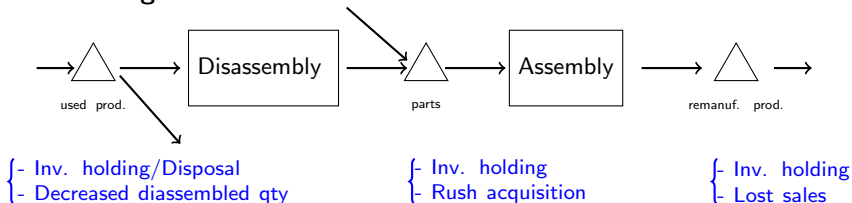
Two-stage stochastic programming approach

Second stage decisions



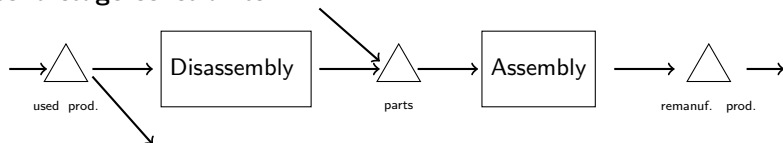
Two-stage stochastic programming approach

Second stage decisions



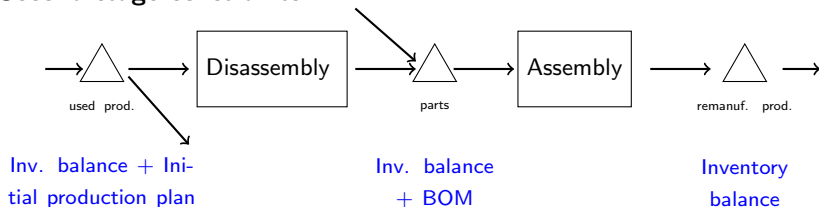
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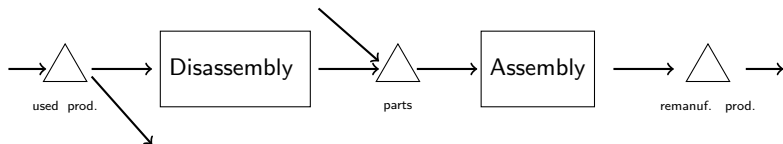


Two-stage stochastic programming approach

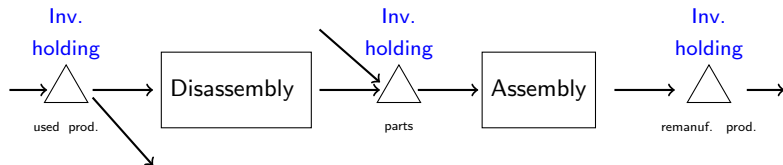
Second stage constraints



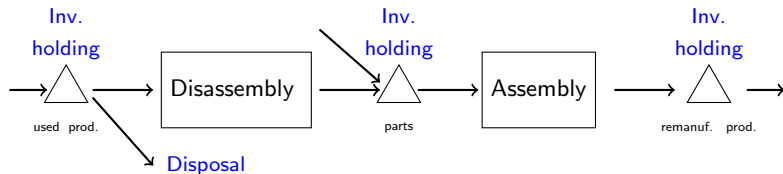
Second stage costs



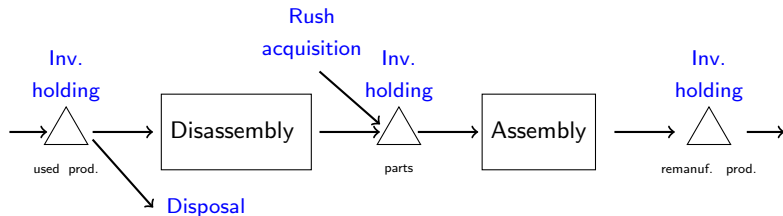
Second stage costs



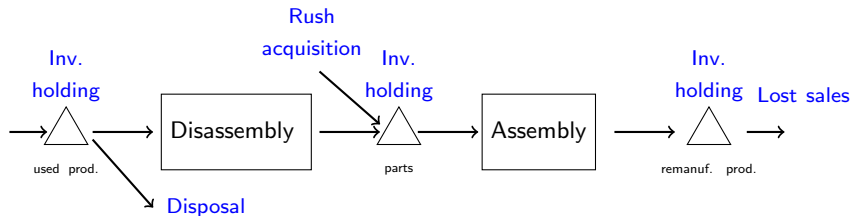
Second stage costs



Second stage costs



Second stage costs



Linear programming formulation

Minimize total production costs

- = first-stage decision costs
 - disassembly/assembly cost
 - new parts acquisition cost
- + expected value of the second-stage decision costs
 - inventory holding cost
 - used products disposal cost
 - new part rush acquisition cost
 - lost sales cost

subject to :

- first-stage constraints
 - disassembly capacity
 - assembly capacity
- second-stage constraints
 - inventory balance of returned products
 - inventory balance of parts
 - inventory balance of remanufactured products
 - link between the initial and adjusted disassembled quantities

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Instances

Instance size

	Used/remanuf. products	Quality levels	Parts	Periods
Instance 1	2	1	1	2

Numerical values of the deterministic parameters

Case study presented in [Jayaraman, 2006]

Remanufacturing of mobile phones

Scenario generation

- Random parameters: uniform probability distribution
- Expected value based on the case study presented in [Jayaraman, 2006]
- Support interval: $\pm 10\%$ around the expected value

Preliminary results

Scenarios	Variables	Constraints	Optimal cost
1	11	6	58.8
10	189	135	75.7
100	617	309	243.5
1000	19994	13995	352.7
2000	39994	27995	483.2
2600	-	-	-

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Conclusion

Remanufacturing planning under uncertainty

- Aggregate production planning
- Multi-product multi-period problem
- Disassembly/Reassembly coordination
- Uncertainty on the demand, returns quantity and quality

Conclusion

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Two-stage stochastic programming approach

- Uncertainty represented by a set of discrete scenarios
- First-stage decisions: initial production and supply planning
- Second-stage decisions: planning adjustments applicable in practice
- Formulation of a large-size linear program

Perspectives

Short-term perspectives

- Develop a efficient solution approach based on the L-shaped method
- Assess the practical interest of using stochastic programming

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Short-term perspectives

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- Assess the practical interest of using stochastic programming

Mid-term perspectives

- Improve the production planning problem representation:
 - more activities: sorting/grading, part refurbishing
 - hybrid manufacturing/remanufacturing system
 - non-linear production costs
- Improve the uncertainty representation
 - multi-stage decision process

Thank you for your attention !