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DRIVING INNOVATION IN MANUFACTURING

Adhesive bonding process optimization via Gaussian Process models

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KNOWLEDGE IN ACTION

Adhesive bonding used in wide variety of applications



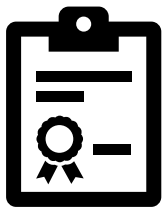
<https://www.dupont.co.uk/products/betaseal.html>



<https://blog.paryleneconformalcoating.com/whats-the-difference-between-potting-and-conformal-coating/>



<https://www.livios.be/nl/bouwinformatie/ruwbouw/muren/verlijmen-versus-dun-metselen/>



Quality of the final bond should be guaranteed during the production process

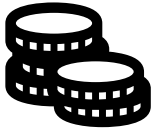
Finding + controlling optimal process parameters is critical for final quality of adhesive bond



Problem



Large number of process parameters which (might) influence final bond quality



Costly experiments: too expensive to optimize by testing them all



Traditional approach: expert knowledge + trial & error



- Settle for suboptimal solution
- Depends largely on experience + knowledge process designer



Use Design of Experiments together with Gaussian process models for optimization

- Adhesive bonding process
- Traditional approach
- Gaussian process models
- Results
- Conclusions

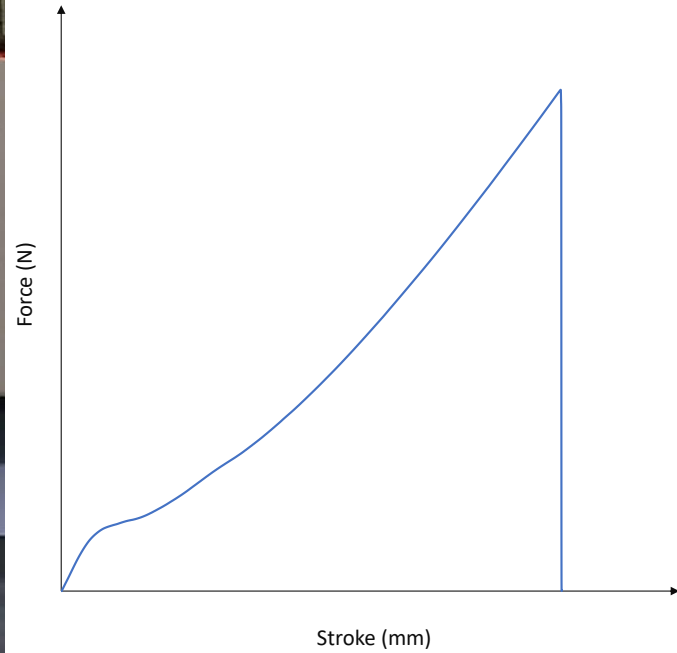
Adhesive bonding process



Adhesive bonding process



Destructive testing



Potting of sensor in PPS housing

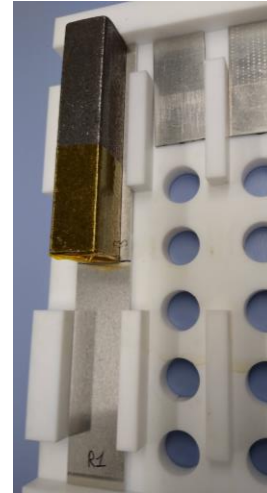
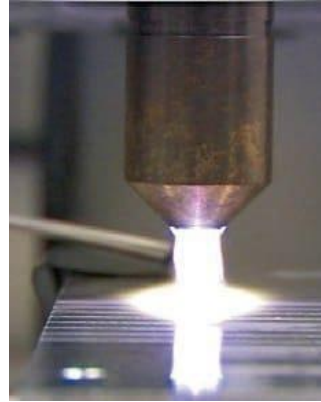
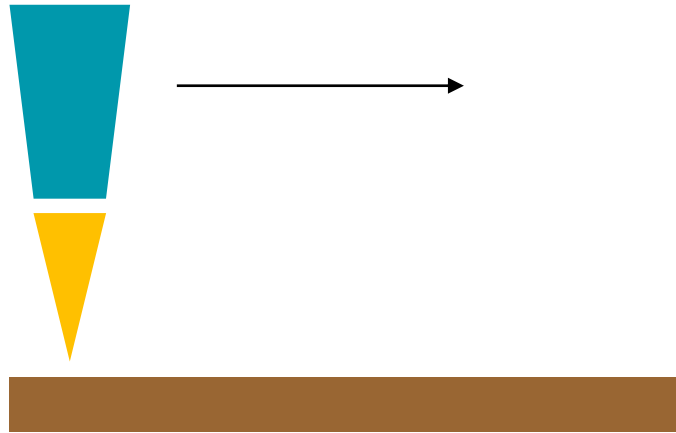


Bond quality



- Break stress
- Failure mode

Adhesive bonding process



Adhesive bonding process – traditional approach

Huge amount of possible combinations ($>10^{100}$)

Variable name	Range
Cleaning	[0, 1]
Plasma power	[300, 500]
Plasma distance between nozzle and substrate	[0.4, 20]
Plasma speed	[5, 250]
Real time plasma-gluing	[1, 120]
Number of passes	[1, 50]



Adhesive bonding process – traditional approach



! Intermediate measurements don't take all effects plasma into account → possible suboptimal solution

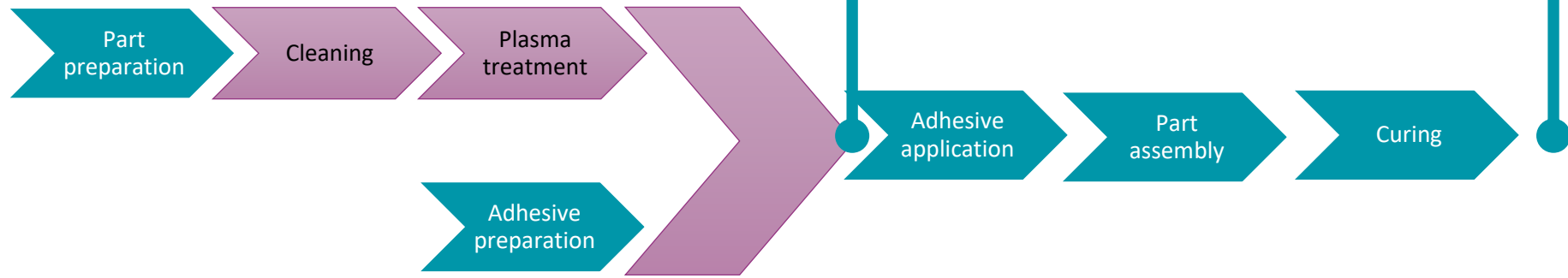
! Strongly dependent on experience + knowledge process designer

Huge amount of possible combinations (>10¹⁰⁰)

Variable name	Range
Cleaning	[0, 1]
Plasma power	[300, 500]
Plasma distance between nozzle and substrate	[0.4, 20]
Plasma speed	[5, 250]
Real time plasma-gluing	[1, 120]
Number of passes	[1, 50]

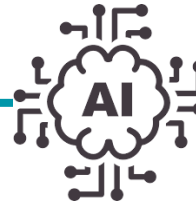
Intermediate measurements (contact angle)

Strength measurements



Adhesive bonding process – Gaussian process models

Proposes new experiments



Analyzes data



Strength measurements

Huge amount of possible combinations ($>10^{100}$)

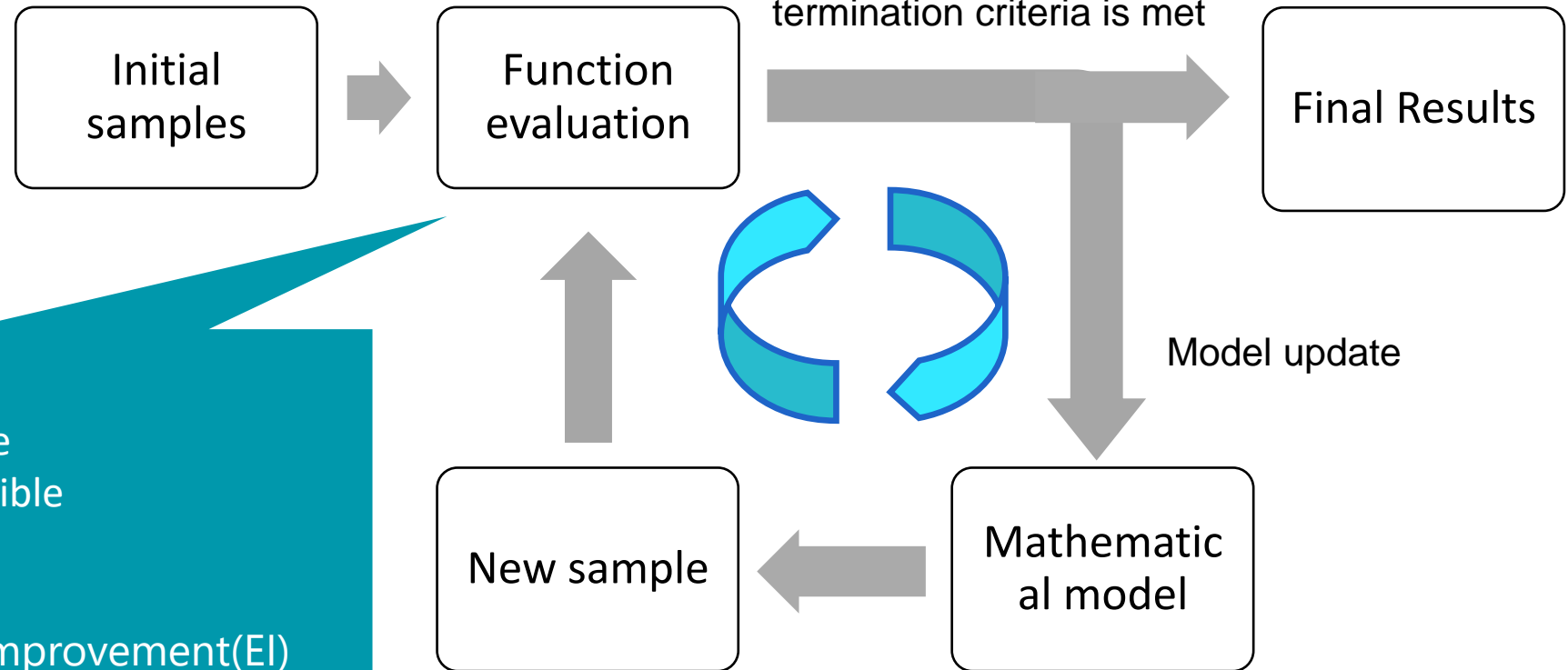
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Number of passes	[1, 50]



Methodology adhesive bonding process – Gaussian process models

DoE:
Latin Hypercube

Budget for finding final
solution consumed



Optimization criteria:

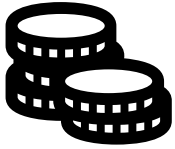
- Break stress as high as possible
- Production cost as low as possible

Bayesian Optimization using

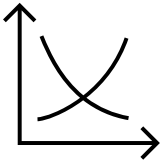
- Single objective: Expected Improvement(EI) acquisition function
- Multi-objective : Hypervolume expected improvement (HVEI) acquisition function

Adapted from (Ivo, 2019)

Termination criteria



Budget 3000 EUR for finding optimal solution

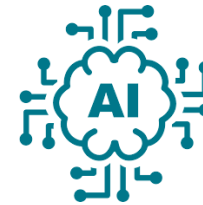


Cost model made about how much it would cost to search for optimal settings in traditional + Gaussian Process Bayesian Optimization
+ cost model for calculating production cost with selected solution



Traditional approach

Contact angle tests: 6 x 30 samples
Strength test: 1 x 30 samples



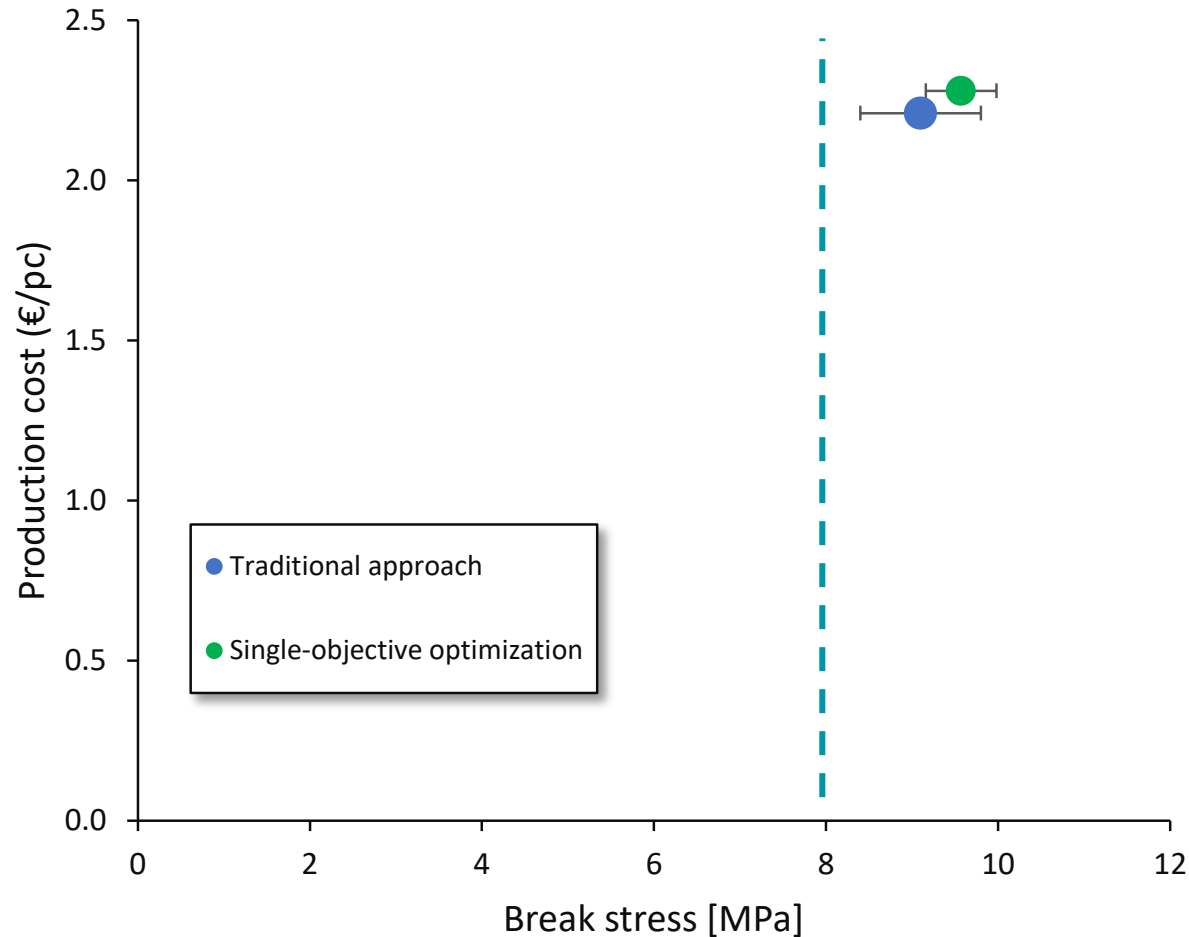
Gaussian processes

Strength test: 9 x 30 samples

Consume budget of 3000 EUR for each approach + compare both final solutions

Results

	Cleaning	Power (W)	Distance (mm)	Speed (mm/s)	Number of passes (-)	Time after plasma (min)	Break stress (Mpa)	StDev
Traditional approach	TRUE	450	10	20	4	30	9.10	8%



- 1) Define optimal solution with **traditional approach**
 - perform contact angle measurements
 - Select interesting points for full test cycle
 - Select only feasible solutions
- 2) Start with **single objective optimization**
Optimization criteria: Break stress as high as possible

Criteria for feasible solution:

- Break stress above 8MPa
- No adhesion failure
- No visual damage due to plasma
- StDev < 15% of break stress

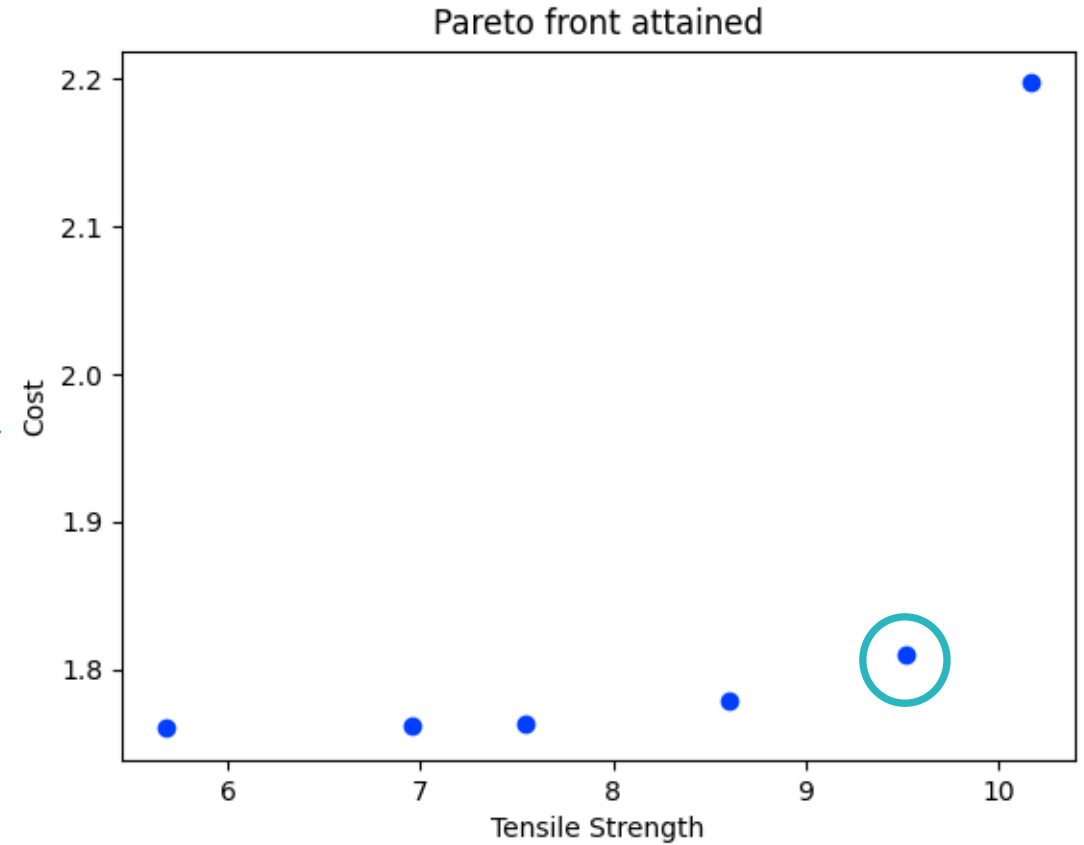
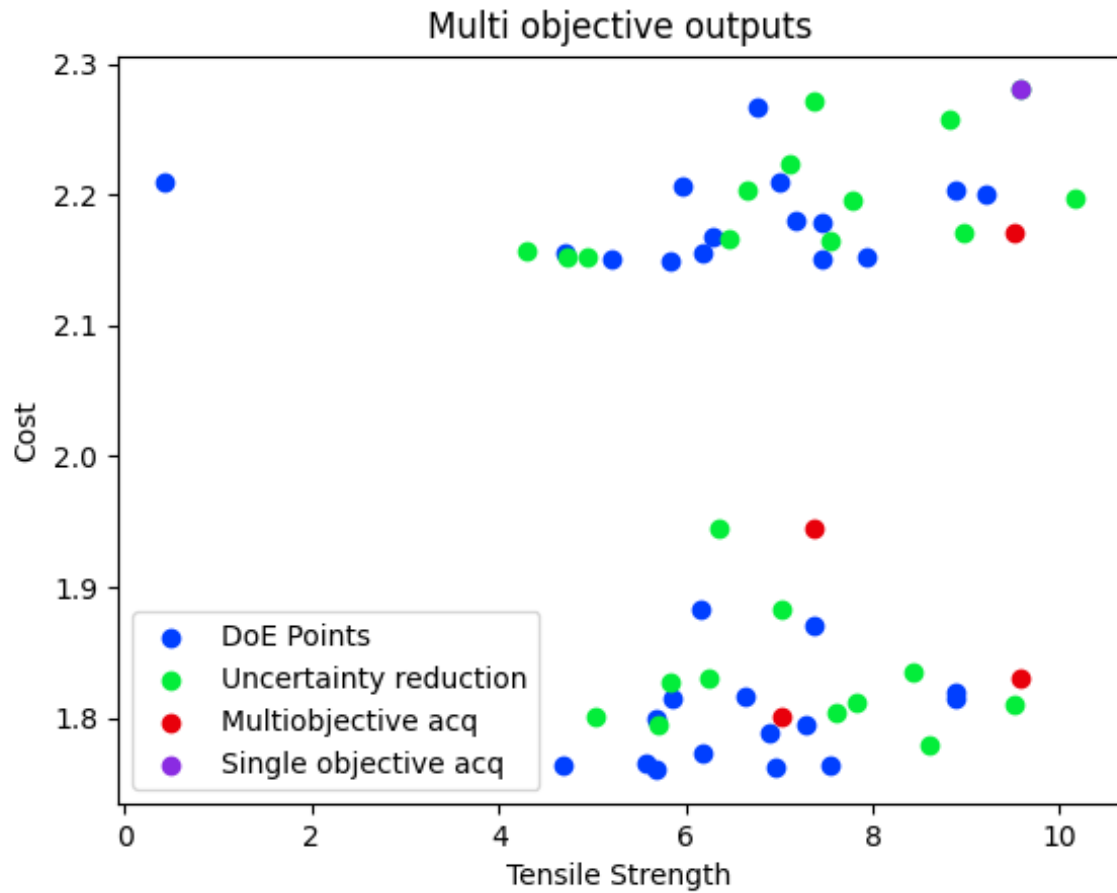
After initial DoE +1 single objective optimization

(67% of budget used)

→ break stress + cost comparable with traditional approach

→ Can we reduce production cost further?

Results multi-objective optimization



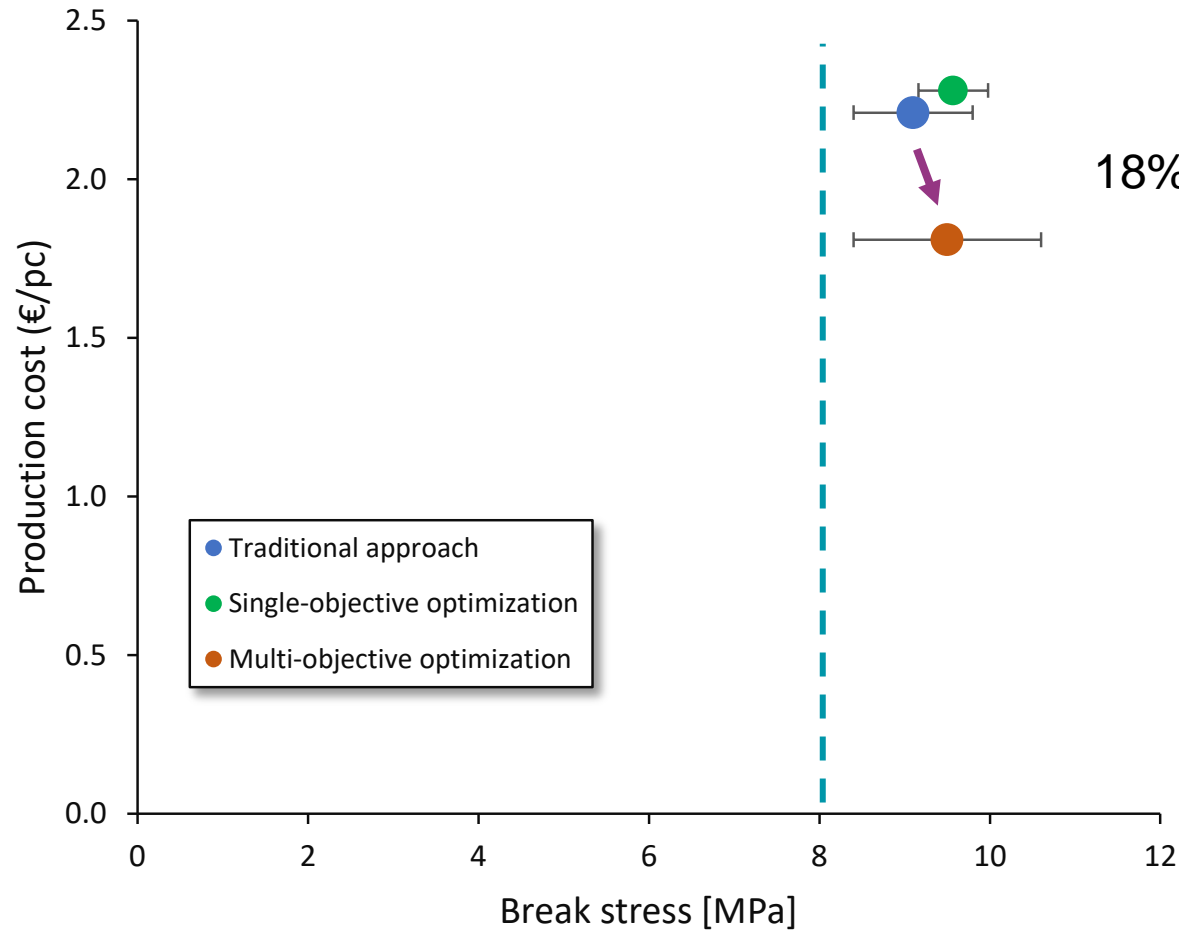
Multi-objective optimization + uncertainty reduction for remaining budget
→ Pareto front attained → interesting point selected

Results

	Cleaning	Power (W)	Distance (mm)	Speed (mm/s)	Number of passes (-)	Time after plasma (min)	Break stress (Mpa)	StDev
Traditional approach	TRUE	450	10	20	4	30	9.10	8%
Single objective optimization	TRUE	330	6	115	49	64	9.57	4%
Multi-objective optimization	FALSE	340	6	60	10	10	9.52	12%

Criteria for feasible solution:

- Break stress above 8MPa
- No adhesion failure
- No visual damage due to plasma
- StDev < 15% of break stress




Conclusions & outline

- Use of DoE & Gaussian process models is valuable approach for optimizing adhesive bonding process parameters
- Bayesian optimization yields similar break stress as traditional approach
- 18% reduction in production cost for final solution obtained via multi-objective optimization
- Trade-off analysis interesting to define when it is no longer beneficial to search for a more optimal solution
- Extend to more process settings (>10 parameters) to cover multiple production steps which mutually influence each other



Questions



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