

AI-Powered Delay Prediction for Portfolio Management

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Who am I?



Portugal → Netherlands



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AI & Data Analysis



Thesis @ DevOn



Guided by Experts



What am I talking about?

AI-Powered Delay Prediction for PPM

Struggles of Predicting Delays

AI for Portfolio Management



Existing Literature - The work of Dr. Kula



Delay Risk Management



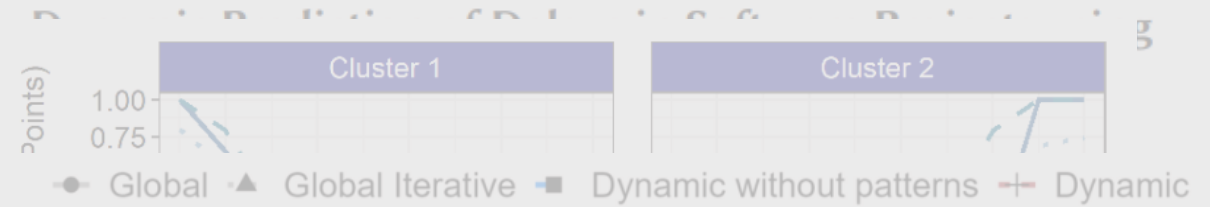
Dynamic AI Model



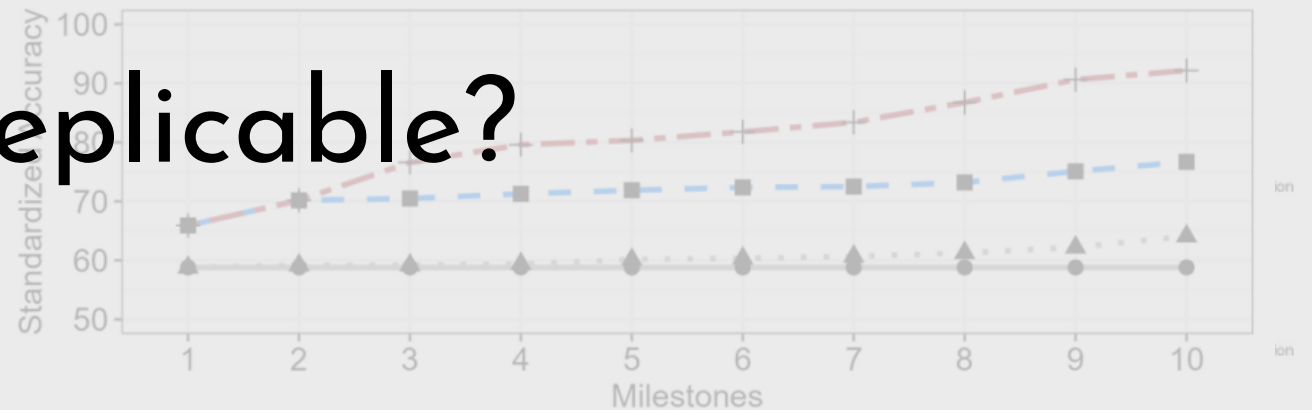
Delay Patterns



Great Results



Replicable?



(a) Standardized Accuracy over time

cent delay patterns across milestones in epic deliveries at
 dyn stat
 ING: 25th percentile: dotted; centroid: solid; and 75th per-
 centile: dashed.
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Replication environment

- ⏻ Different industry
 - ⏻ Faster moving
 - ⏻ Less regulated
 - ⏻ More demand-driven
- ⏻ Similar investments
 - ⏻ Agility
 - ⏻ Data-driven decision-making



How effective is Dr. Kula's solution
when applied to CCH's Project Portfolio?

Data Cleaning and Preprocessing



🔌 7463 epics

🔌 4040 clean epics

🔌 54% usability rate



🔌 2164 epics

- non-delivered
- no target dates
- < 10 planned sprints
- < 10 actual sprints

🔌 354 clean epics

🔌 16% usability rate

Collecting Delay Factors

Risk factor	Predictor variable	Description
Task dependencies	1. out-degree	Number of outgoing dependencies of an epic on other epics
Organizational stability	2. changed-leads	Number of changed tribe leads during the current and previous epic
Team stability	3. stability-ratio	Median of the ratio of team members that did not change during the current and previous epic
Skills and knowledge	4. dev-age-ing	Median of the number of years the developers working on the epic have been working at ING
Team familiarity	5. team-existence	Median of the number of years teams have existed in their current composition of team members
Team commitment	6. hist-performance	Median of the ratio of on-time delivered epics over all teams working on the epic
Work in progress	7. dev-workload	Median of the number of story points assigned to a developer per sprint
Bugs or incidents	8. nr-incidents	Number of incidents that occurred during the development phase of the epic
	9. unplanned-stories	Number of unplanned stories (related to bug fixes or incidents) that have been added to the epic
Project size	10. nr-stories	Number of planned stories assigned to the epic
	11. nr-sprints	Number of sprints assigned to the epic
	12. team-size	Median team size in the epic
Project security	13. security-level	The ratio of user stories in the epic that need to pass a security testing process



Final Datasets



- ⏻ 4040 epics
- ⏻ 13 predictor variables



- ⏻ 354 epics
- ⏻ 7 predictor variables

Delay Patterns

- ⚡ Show **when** delays happen throughout an epic
- ⚡ Large distance between 25th and 75th percentiles
 - ⚡ Small dataset
 - ⚡ Unclear or non-existing patterns
- ⚡ Delay patterns are not indicative of overall delay

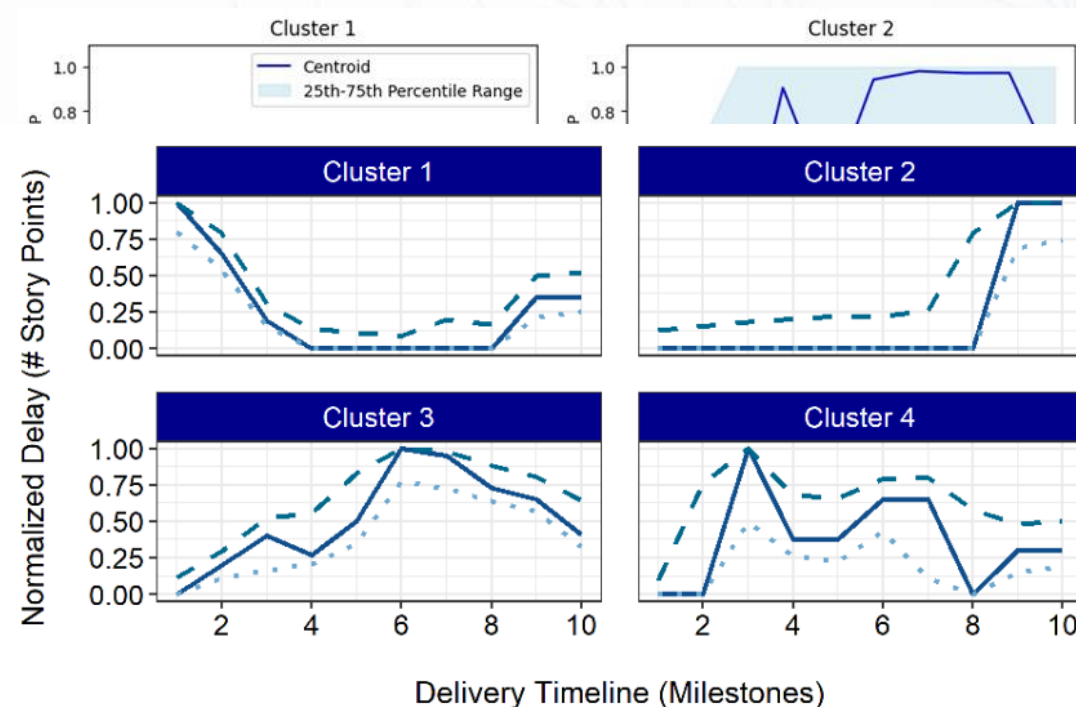
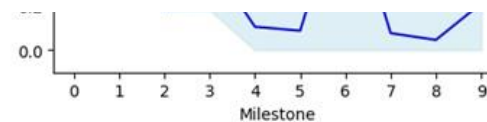


Figure 4: Four clusters of delay profiles representing recurrent delay patterns across milestones in epic deliveries at ING: 25th percentile: dotted; centroid: solid; and 75th percentile: dashed.



Performance Measurements

🔌 Epic delivered

🔌 on time: BRE = 0

🔌 50% longer: BRE = 0.5

Balanced Relative Error (BRE)

$$\text{If } \text{Act} - \text{Pln} \geq 0, \text{ then BRE} = \frac{\text{Act} - \text{Pln}}{\text{Planned Duration}}$$

$$\text{If } \text{Act} - \text{Pln} < 0, \text{ then BRE} = \frac{\text{Act} - \text{Pln}}{\text{Actual Duration}}$$

🔌 Average distance between predicted and actual delay

Mean Absolute Error (MAE)

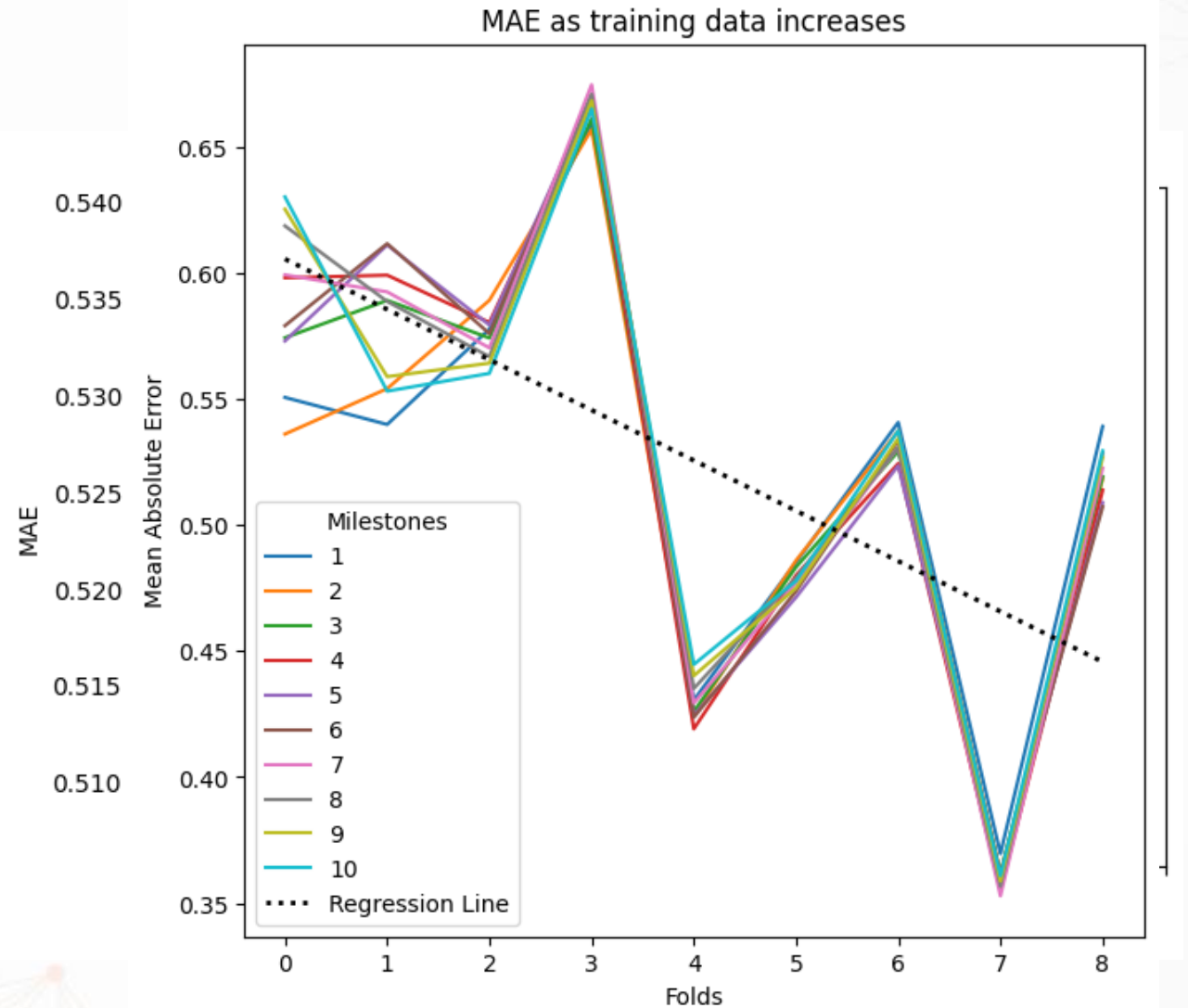
$$MAE = \frac{1}{N} \sum_{i=1}^N | \text{Actual BRE}_i - \text{Estimated BRE}_i |$$

🔌 The lower the MAE, the higher the accuracy



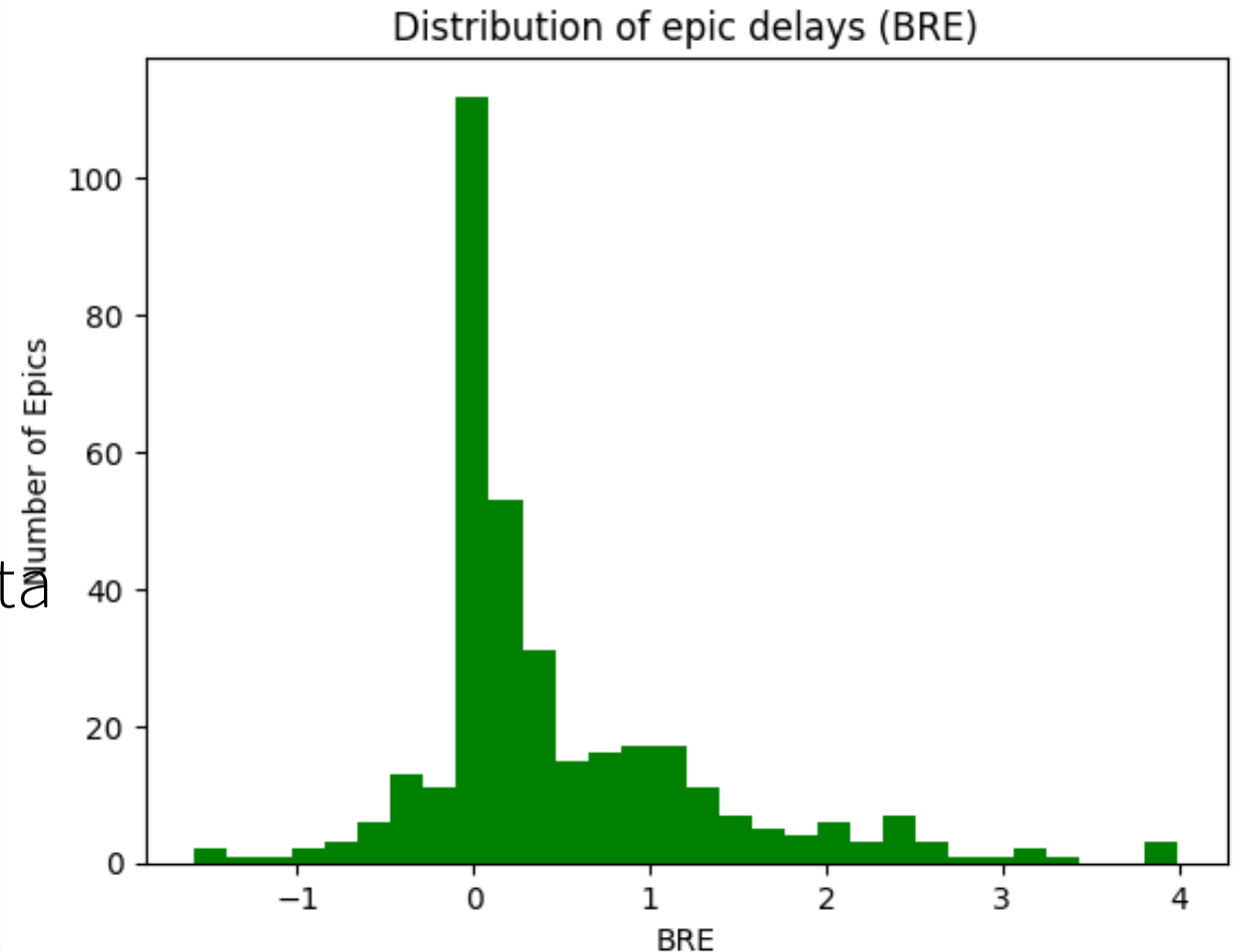
Predictive Results

- ⚡ MAE ≈ 0.5
 - ⚡ Off by half the epic's length
- ⚡ MAE at ING: 0.19 to 0.04
- ⚡ MAE goes down as amount of training data increases
 - ⚡ More data, better results!



How effective is Dr. Kula's solution when applied to Coca-Cola Hellenic's Project Portfolio?

- Mean Absolute Error is 0.5
- Effective outlier detector
- Performs better with more data
- Overall, not effective.



Data quality is everything.
Data takes time, start today!

Does CCH really need this?

Great solutions are tailored to organizational
needs, not wants.

Bottom-up change > Top-down enforcement

Thank you

Gaspar Rocha

