

# The Hidden Cost of Healthcare

Transforming medical equipment management with data and analytics

## Why it matters | Current practice | HANDLE Global | Question

**\$60 bn** annual medical equipment spending (U.S.)

**30%** of total costs of health systems)

**“The doctor with the loudest voice gets new equipment”**

**No data-driven decision like in other industries (e.g. transportation)**



**First to offer a data solution for asset management in healthcare**

**Can we use historical maintenance data to provide decision support for better asset replacement strategy?**

## Descriptive Analytics

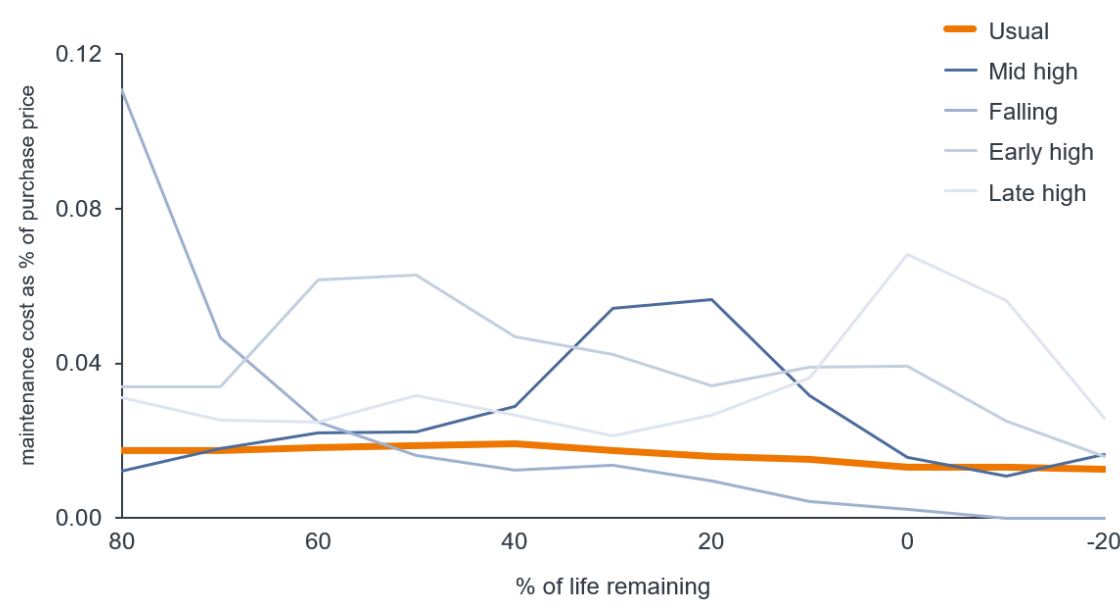
### Maintenance cost over lifecycle (clustering)

**Time series clustering on maintenance cost aggregated by product/category**

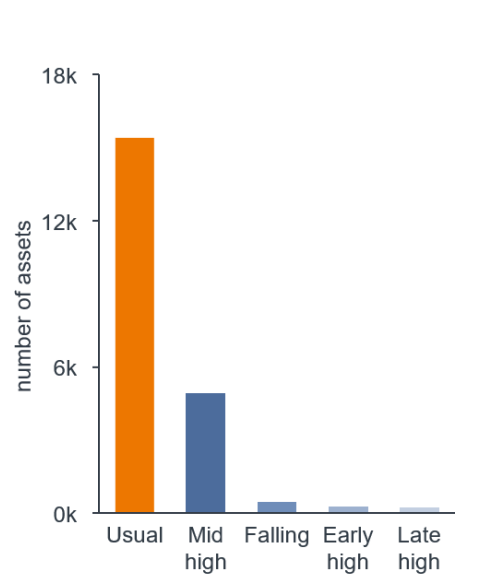
**Distance metric:** Dynamic time wrapping & Euclidean

**Method:** Agglomerative/hierarchical clustering

#### Cluster across life cycle



#### Assets per cluster



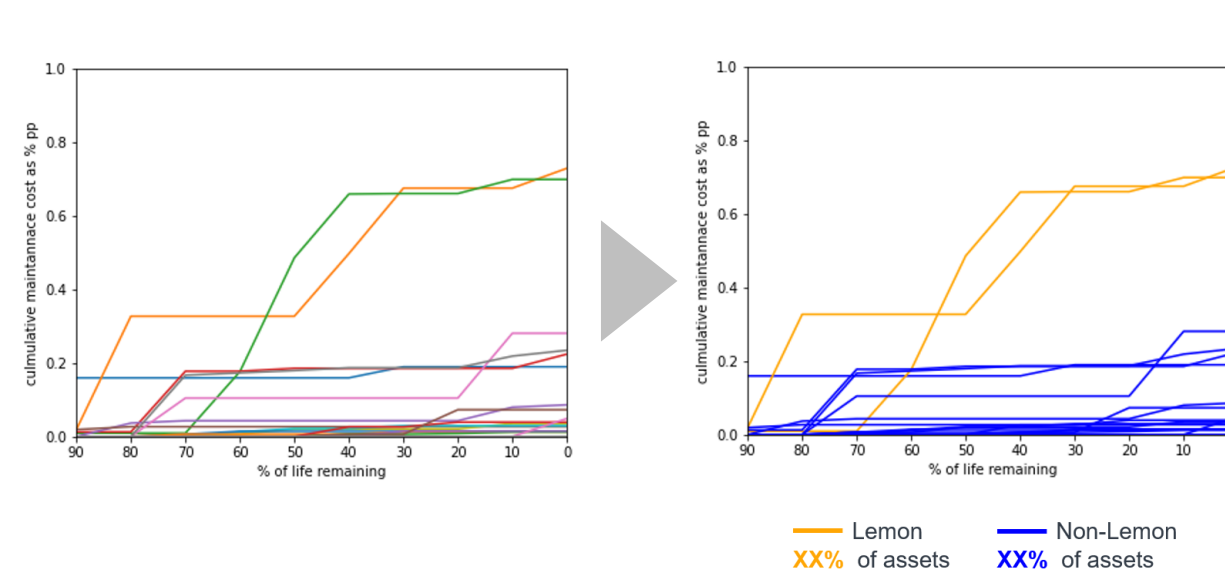
### Lemon analysis

**Lemon = asset with significantly higher life-time maintenance cost as its peers**

**1- Demeaning** of asset maintenance cost to achieve **global comparability**

**2- Select** global cutoff for lemons - **how much more expensive is a lemon?**

#### Example Mindray Passport2 (monitor)



#### Outcome

Handle Product ID Name	Lemon n	Non-Lemon
GE Healthcare SENDAdvantage	40%	60%
Philips L15710	29%	71%
GE Healthcare MAC5500HD	24%	76%
GE Healthcare SAM	23%	77%
Philips L93	22%	78%
Hill-Rom 4802	20%	80%
GE Healthcare MMS	20%	80%
Olympus GIFQ180	20%	80%
Philips FM30	17%	83%
Philips CX50	17%	83%
GE Healthcare ERC	13%	88%
GE Healthcare PRO400	12%	88%
Mindray PASSPORT2	11%	89%

**Ranking of product per percentage of lemon**

Majority of assets show stable maintenance cost over time

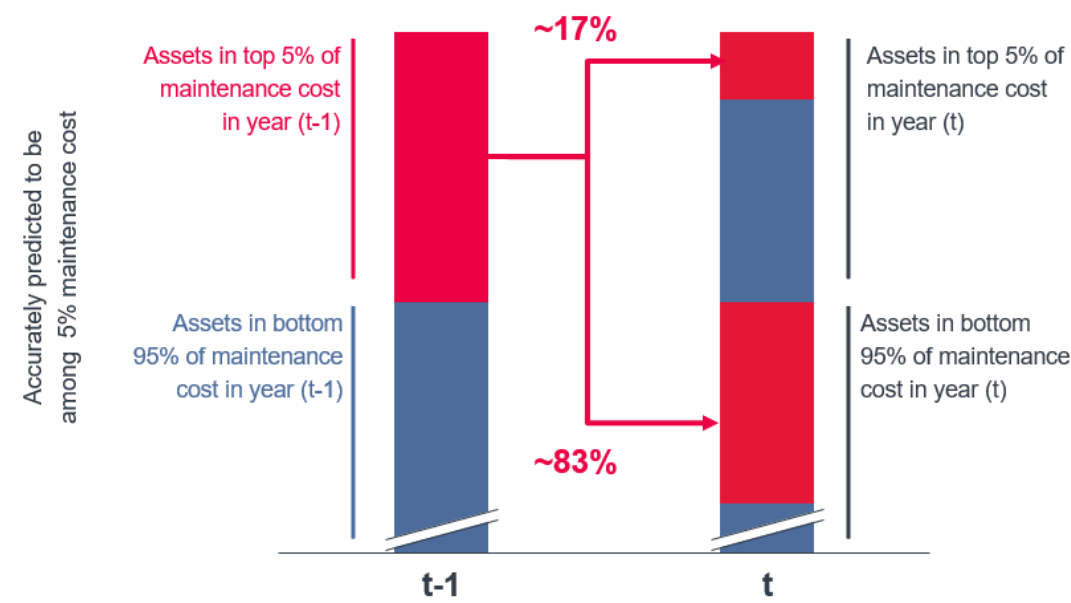
Products can be ranked by relative number of “lemons”

## Predictive Analytics

### Predictive setup

**Predicting expected annual maintenance cost to detect early costly assets.**

#### Comparison of high maintenance assets at t & t-1

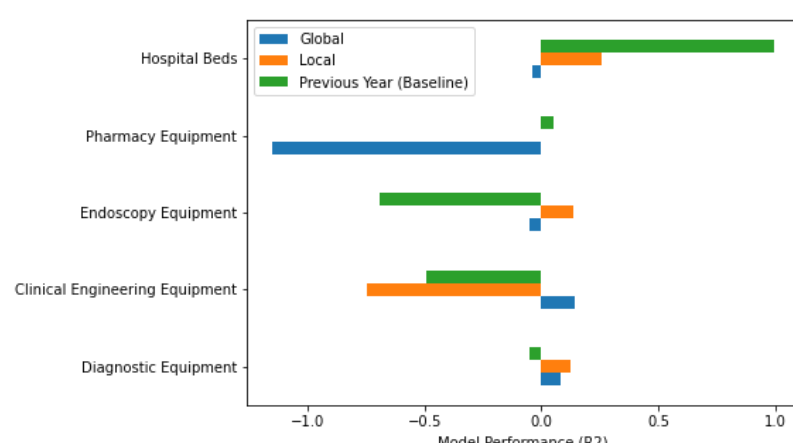


Only ~17% of high maintenance assets in year t-1 are again high maintenance year t. **Disaggregated cost is unstable**

Last year's cost is a bad predictor

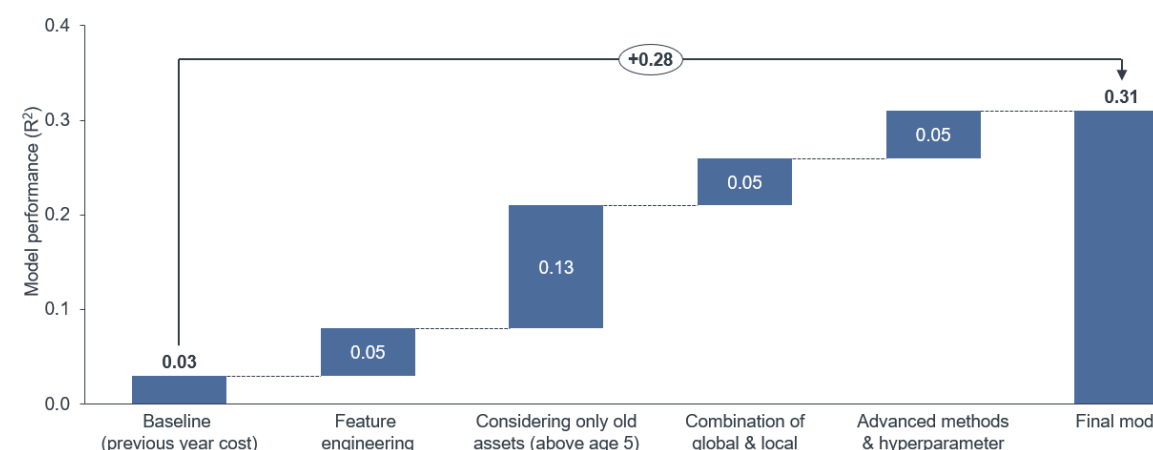
### A Global-Local-Baseline (GLB) approach

#### Example of GLB



For each category, we select the **best predictor** between a **local linear regression**, a **global gradient boosting** and the **previous year's cost**

#### R<sup>2</sup> improvement

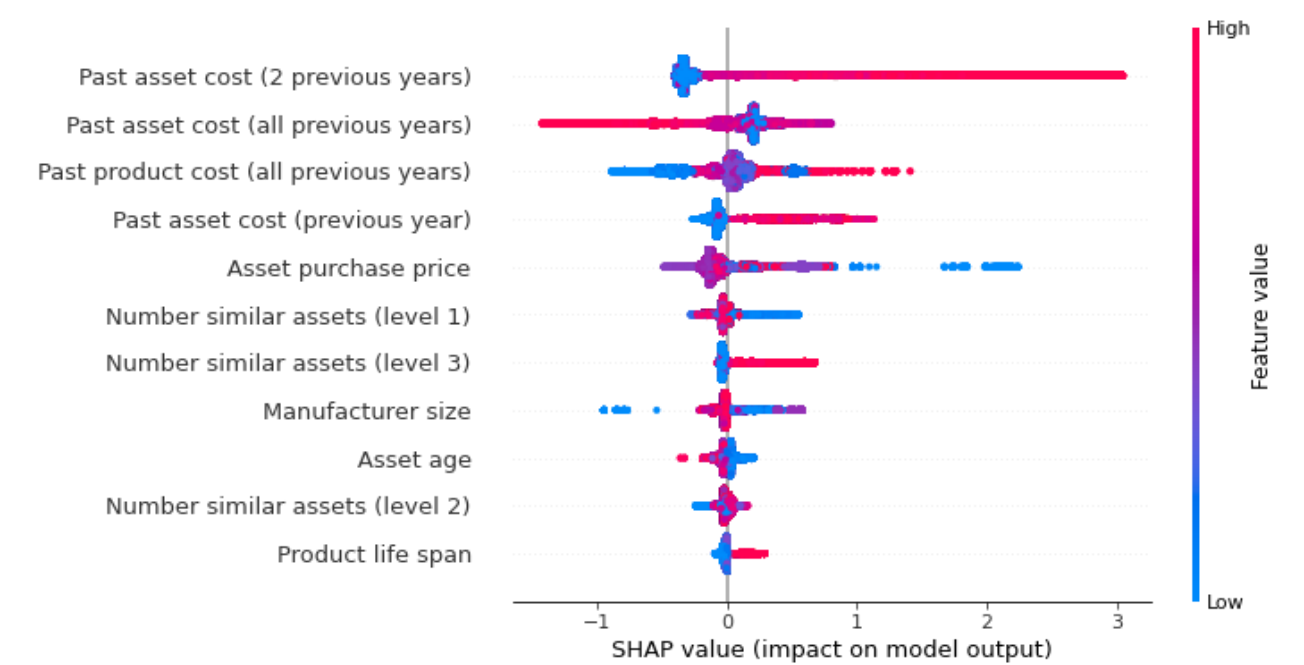


28% R<sup>2</sup> improvement with GLB

### Feature importance

Feature importance makes our model more interpretable, but also will guide future data collection and equipment handling.

#### Shap plot



Asset age is a weak predictor for maintenance cost.

Maintenance cost is predicably with MAE \$90 per asset per year

## Prescriptive Analytics

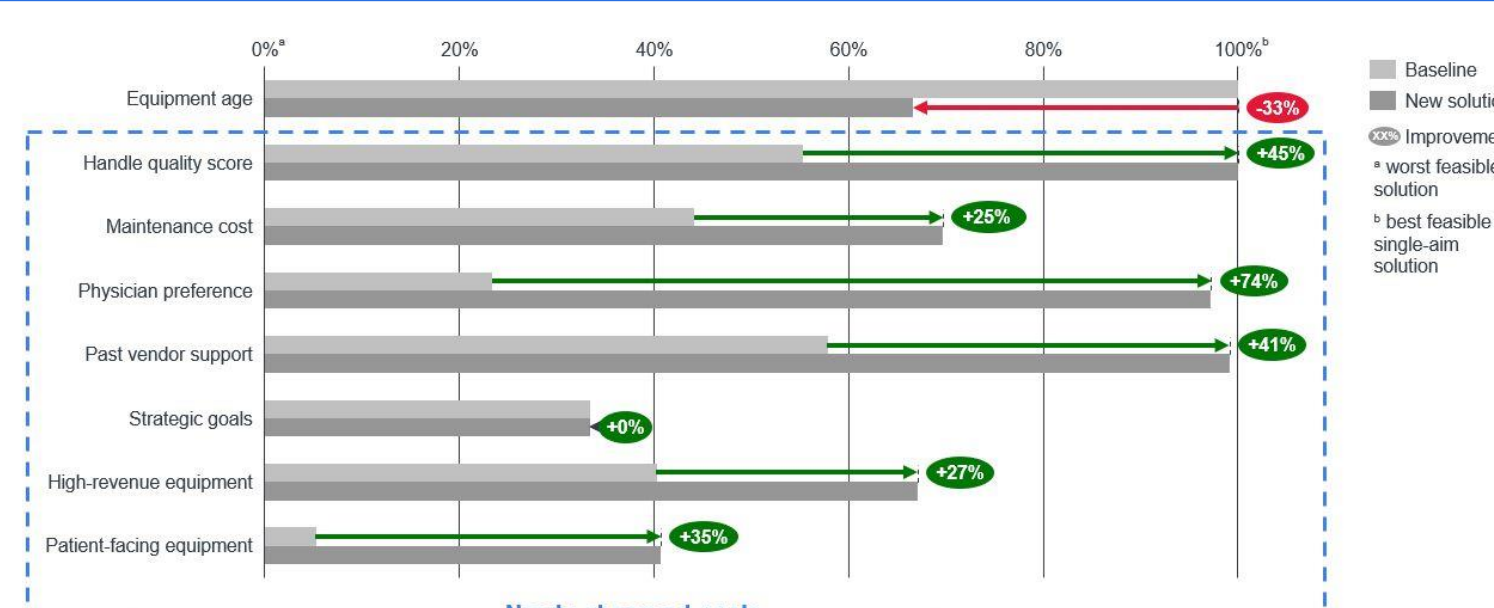
### What to optimize for?

- HANDLE score:** Asset quality
- Expected maintenance cost:** Cost and reliability
- Physician preference:** Physician satisfaction
- Past vendor support:** Risk of higher replacement effort
- Strategic goals:** E.g. Standardization
- High-revenue equipment:** Risk of revenue-loss
- Patient-facing equipment:** Patient experience

#### Technique:

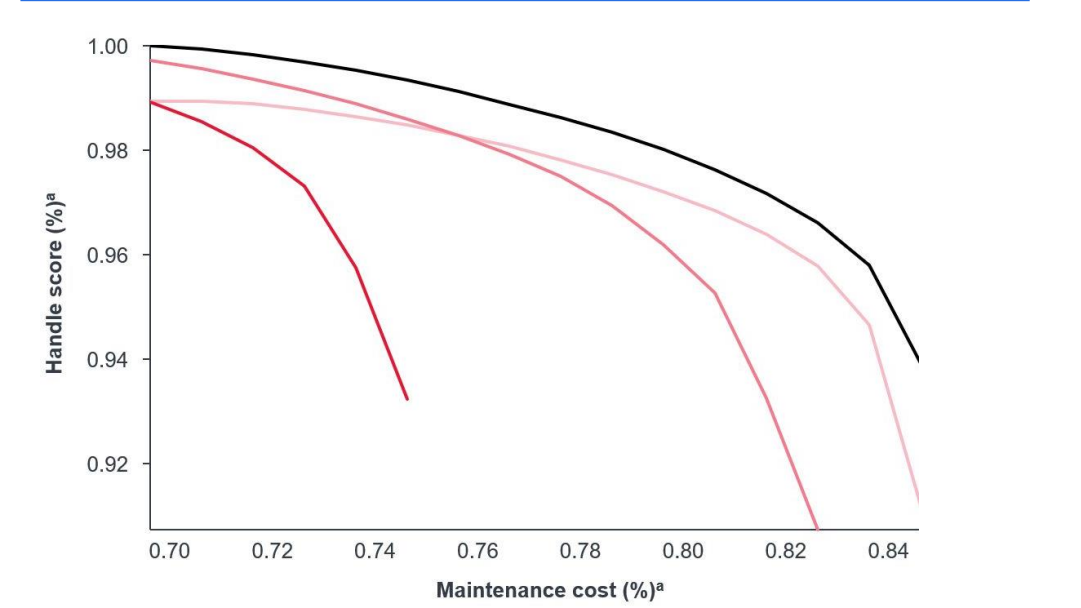
- Iteration over objectives** to find **Pareto optimal** solution
- Relaxation** of binary constrain for better scalability

### Improvement



**15% increased capital-effectiveness**

### Trade-off analysis



Original pareto frontier, +5% increase in strategic goals, +5% increase in patient-facing equipment, +5% increase in high-revenue equipment

**Pareto frontier to illustrate trade-offs between multiple objectives**

