

Leveraging Data Science to Improve Device Uptime

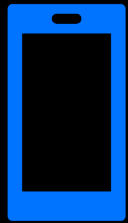
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Leveraging Data Science to Improve Device Uptime

Agenda



**Device uptime
and battery
performance**



The steps
Sense, analyze
and act



The algorithm
Remaining useful
life (RUL) of
battery



**Applications of
RUL**

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Device Uptime and Battery

Operational challenges that impact device uptime and availability

- Can the battery last the shift – does it have enough capacity
- How many batteries will have enough capacity after 6 months to remain useful for the business purpose
- How to identify batteries that are degraded and approaching end of life, so they can be replaced proactively

Battery challenges for the front-line workers

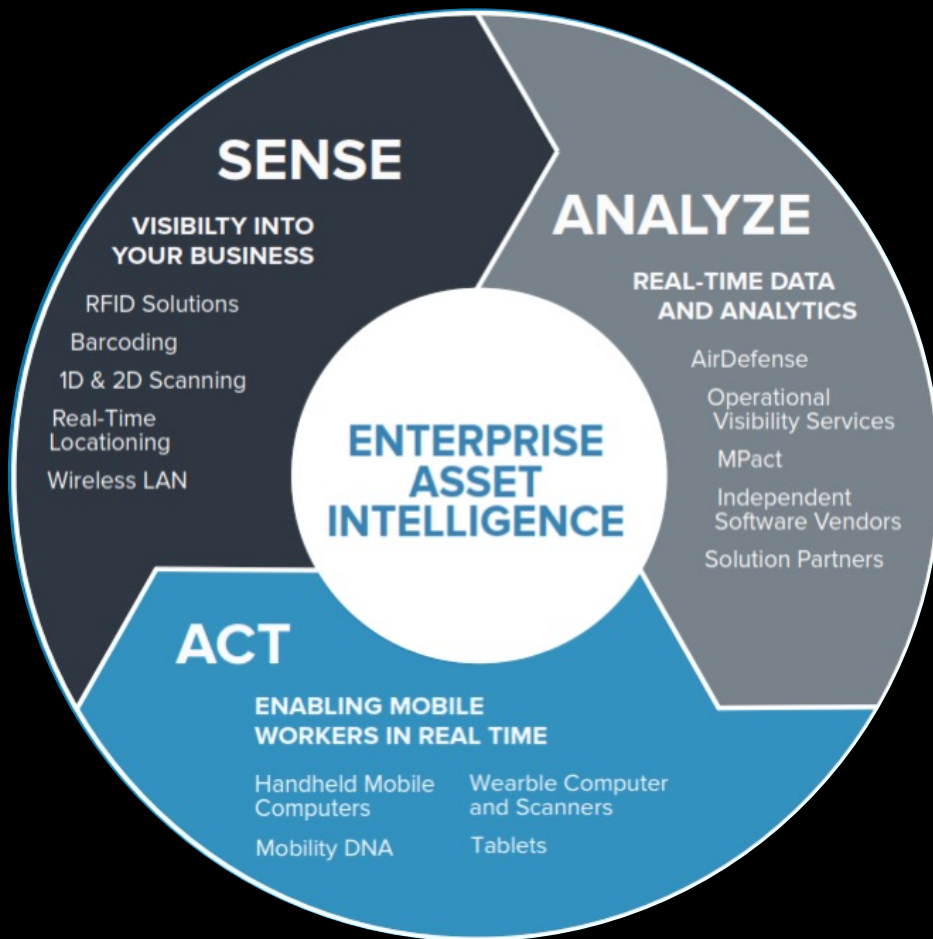
- Front-line workers need to have sufficient batteries capacity to handle site workload
- Front-line workers need the ability to identify batteries that should be replaced and mark them for removal from battery pool.

This translates to:

- Know your battery inventory
- What is the status of batteries and when they might need replacement
 - Forecasting for Operations Planning

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The Steps



Sense - Device data including battery related data coming from the battery pack. Consider any other relevant data from external sources



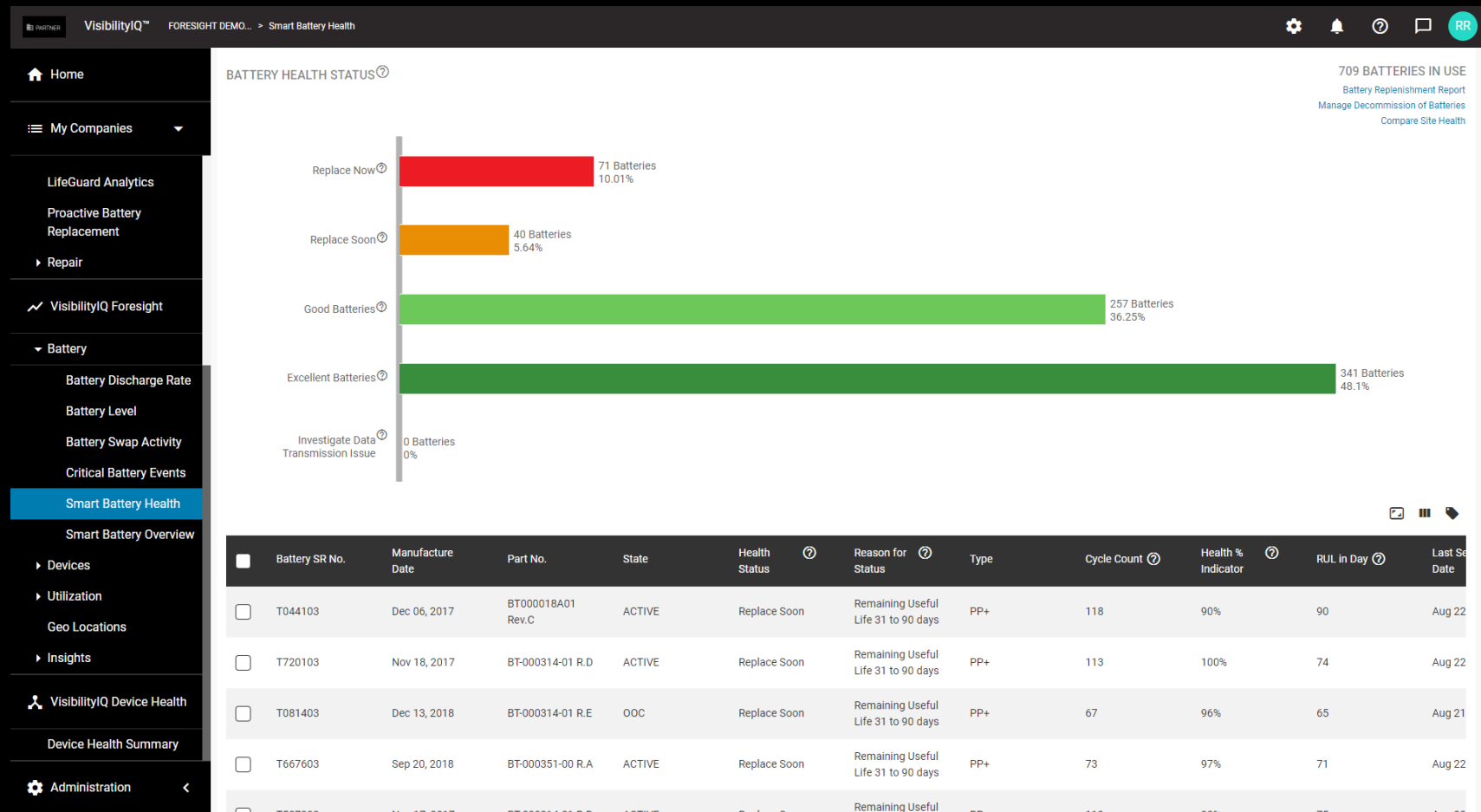
Analyze - Go through the data and apply relevant algorithms to generate insights like Remaining Useful Life of the battery



Act - Battery Replacement recommendations based on generated insights

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Outcome - Smart Battery Report



A sample report to recommend batteries to be replaced based on RUL

Zebra DevCon 2023

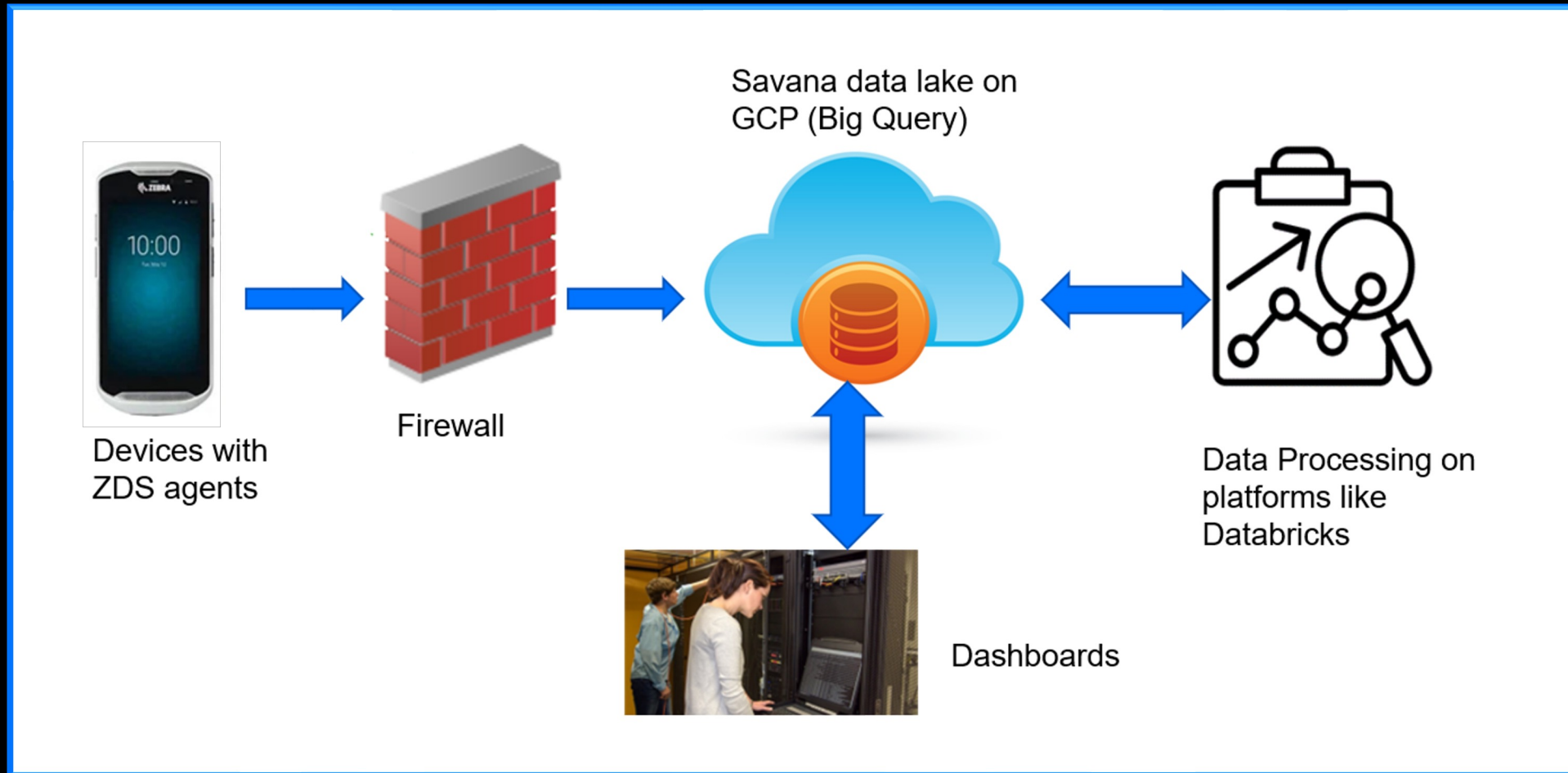


Sense – Device and Battery Data Captured in the Cloud



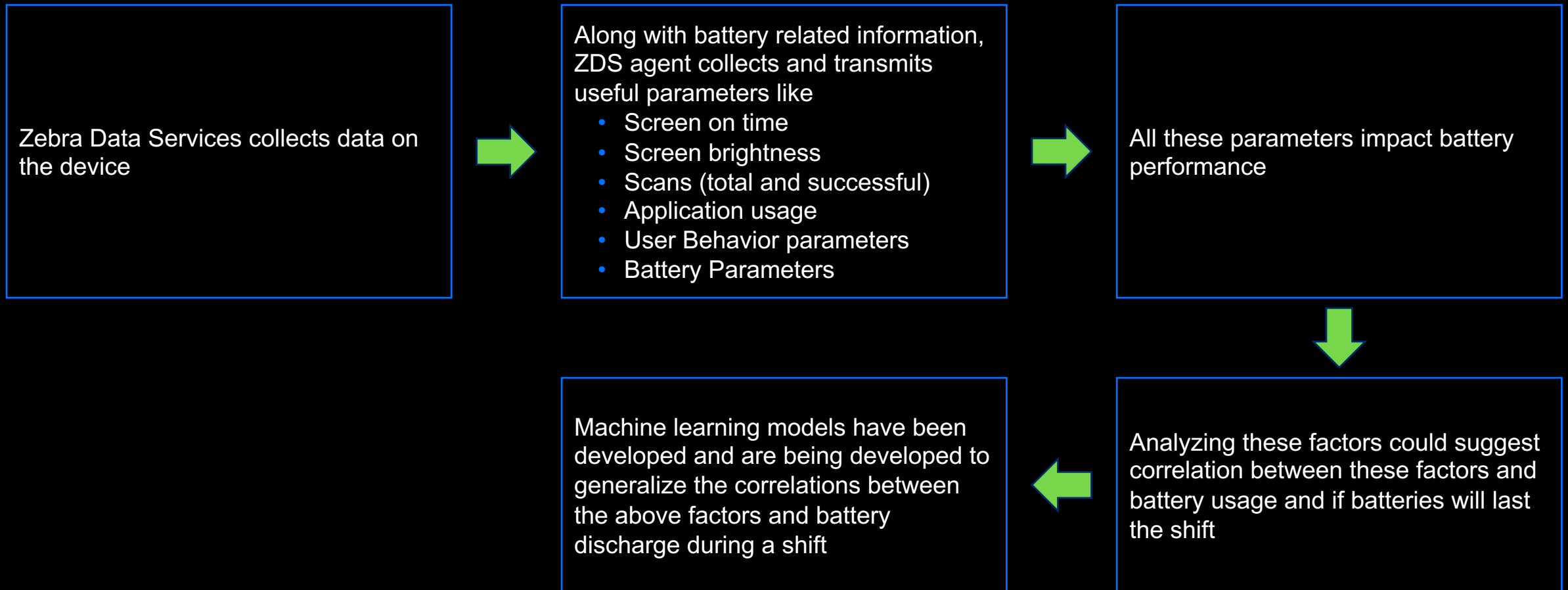
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Architecture Overview



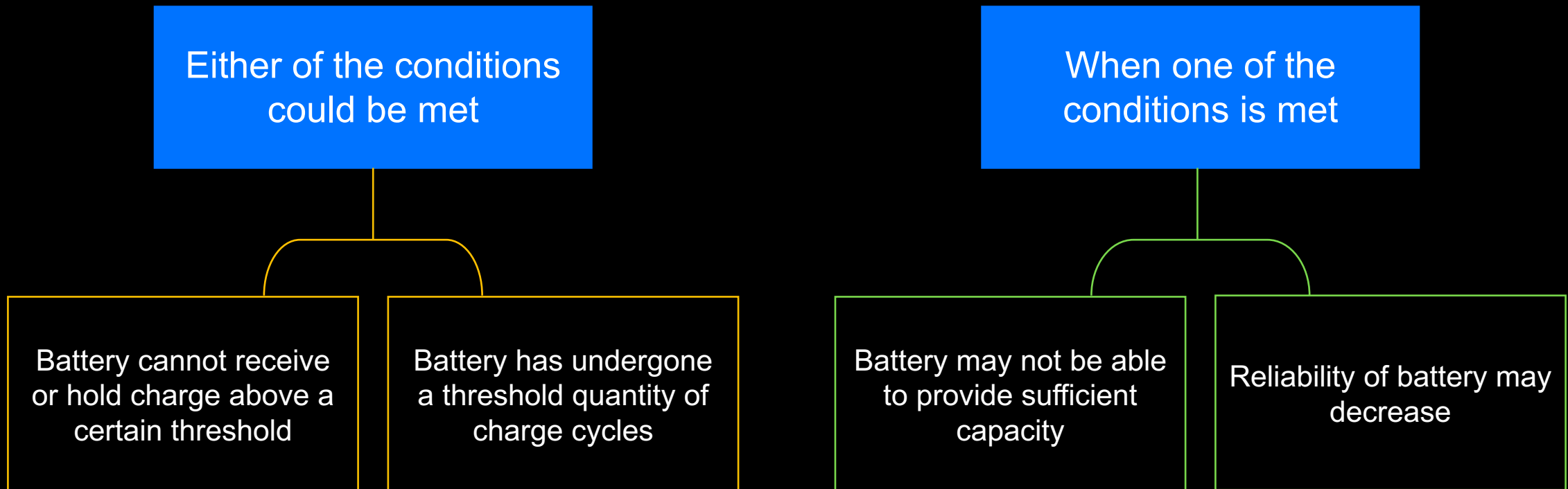
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Data Available from Zebra Data Services (ZDS) agent



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Defining Remaining Useful Life

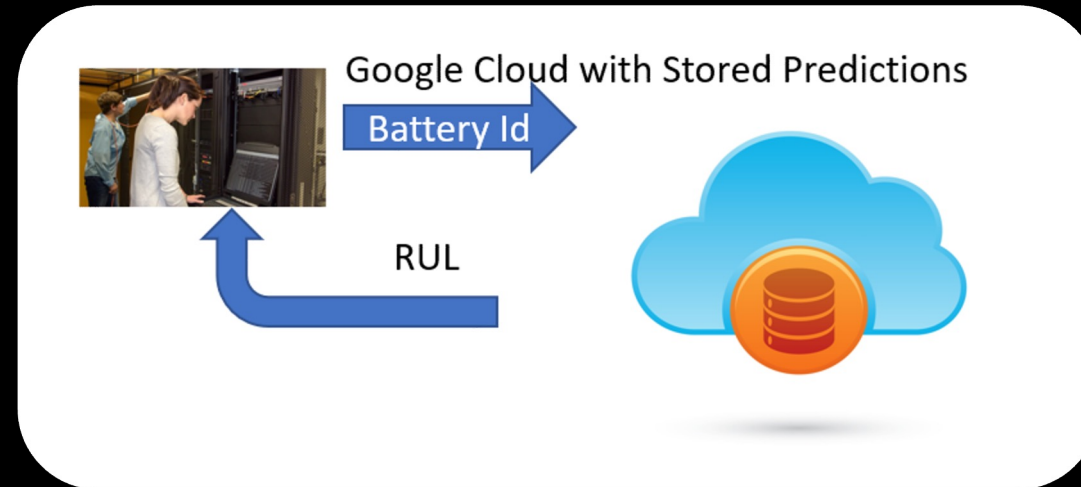


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Defining Remaining Useful Life

Factors that lead to battery degradation over time:

- Battery Cells
- Environment
- Usage patterns



Remaining Useful Life – The predicted number of days before the batteries key indicators fall below recommended levels

Number of days after which we expect the battery parameters to cross the manufacturer recommended threshold

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The Input Data

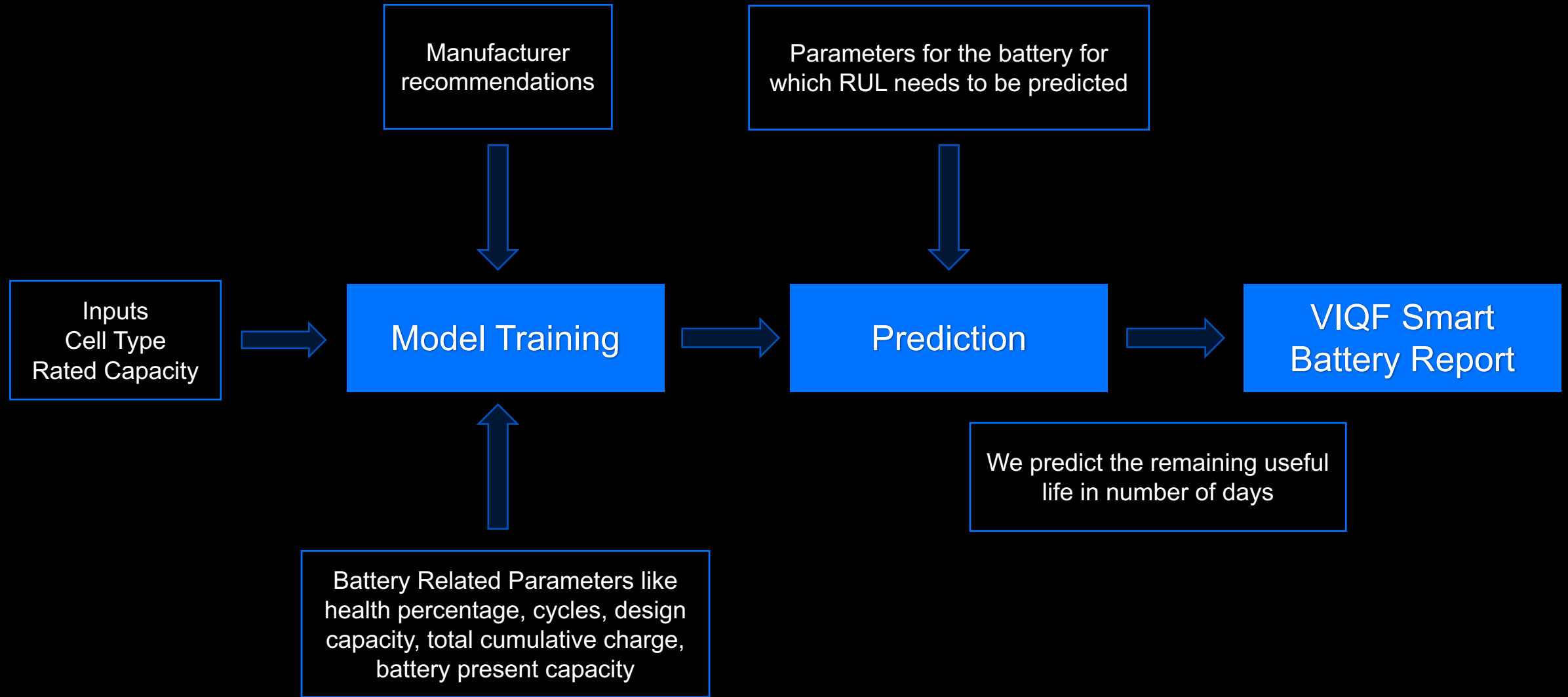


Battery Data being considered

- Data reported from the battery gauge
 - Battery Design Capacity – A measure of how much capacity a new battery of a particular type can deliver
 - Battery Present Capacity – a measure of how much charge the battery can deliver when fully charged.
 - Health Percentage = $\text{Battery Present Capacity} / \text{Battery Design Capacity}$.
 - Total cumulative charge – A measure of how much charge has flown through the battery since beginning of its use
 - Charge Cycles - A measures of how many complete charge and discharge cycles the battery has gone through. This is a usage measurement pattern of the battery
- Other Useful Data for RUL
 - Manufacturer recommendations on permissible health percentage and/or charge cycles

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The Process



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Data Analysis and Data Quality

Battery should be uniquely identifiable

Battery Charge cycles should be positive and monotonically increasing

Battery Present Capacity should go down over a period of time. It cannot go up defying laws of nature

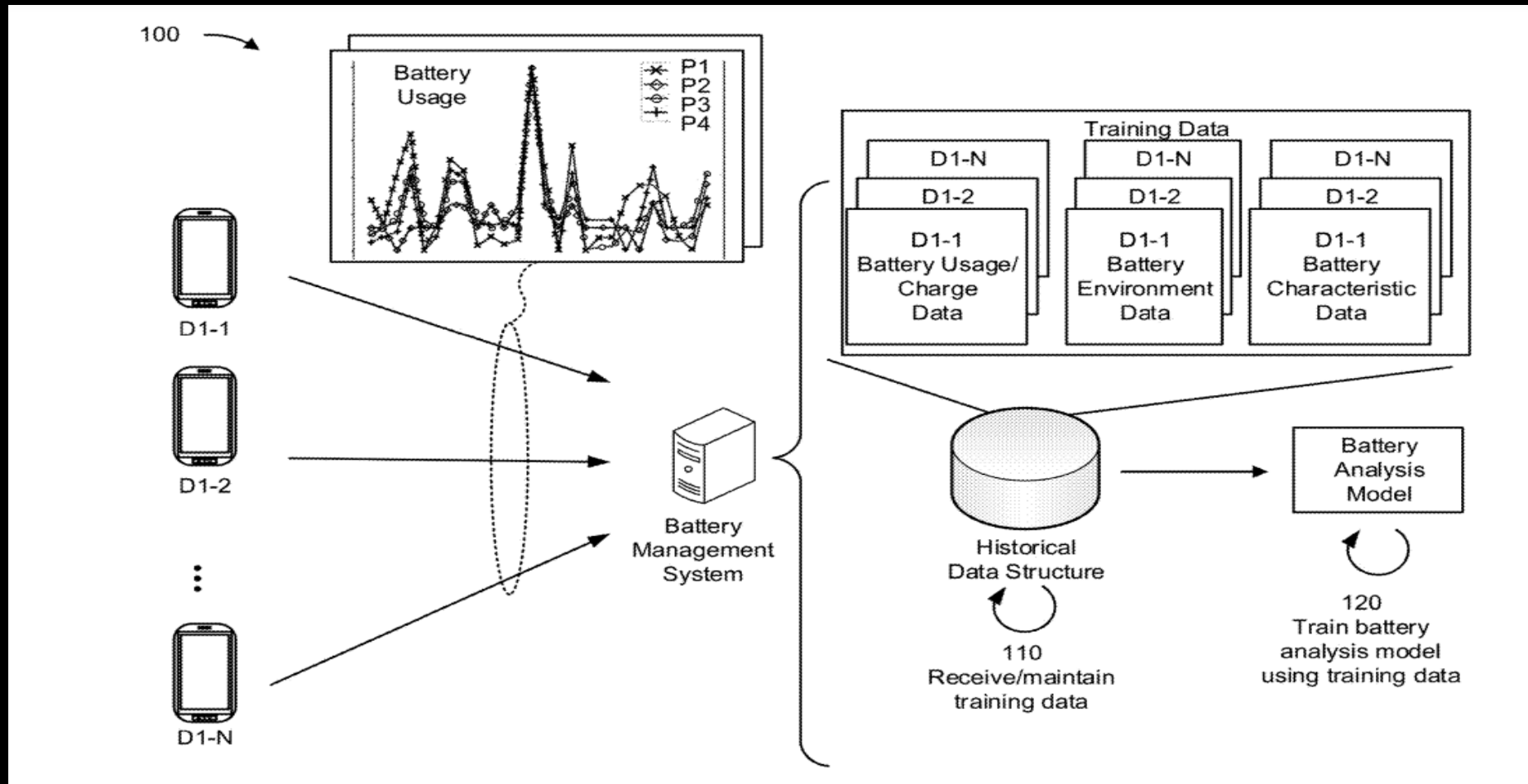
Battery Rated Capacity should not be negative or zero

Battery health percentage should be a number between 0 and 100

Data anomalies need to be handled

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Training



Reference - <https://patentimages.storage.googleapis.com/65/52/6d/57ad11a747540e/US20220294027A1.pdf>

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The Model

Model gives values for degradation rates of health

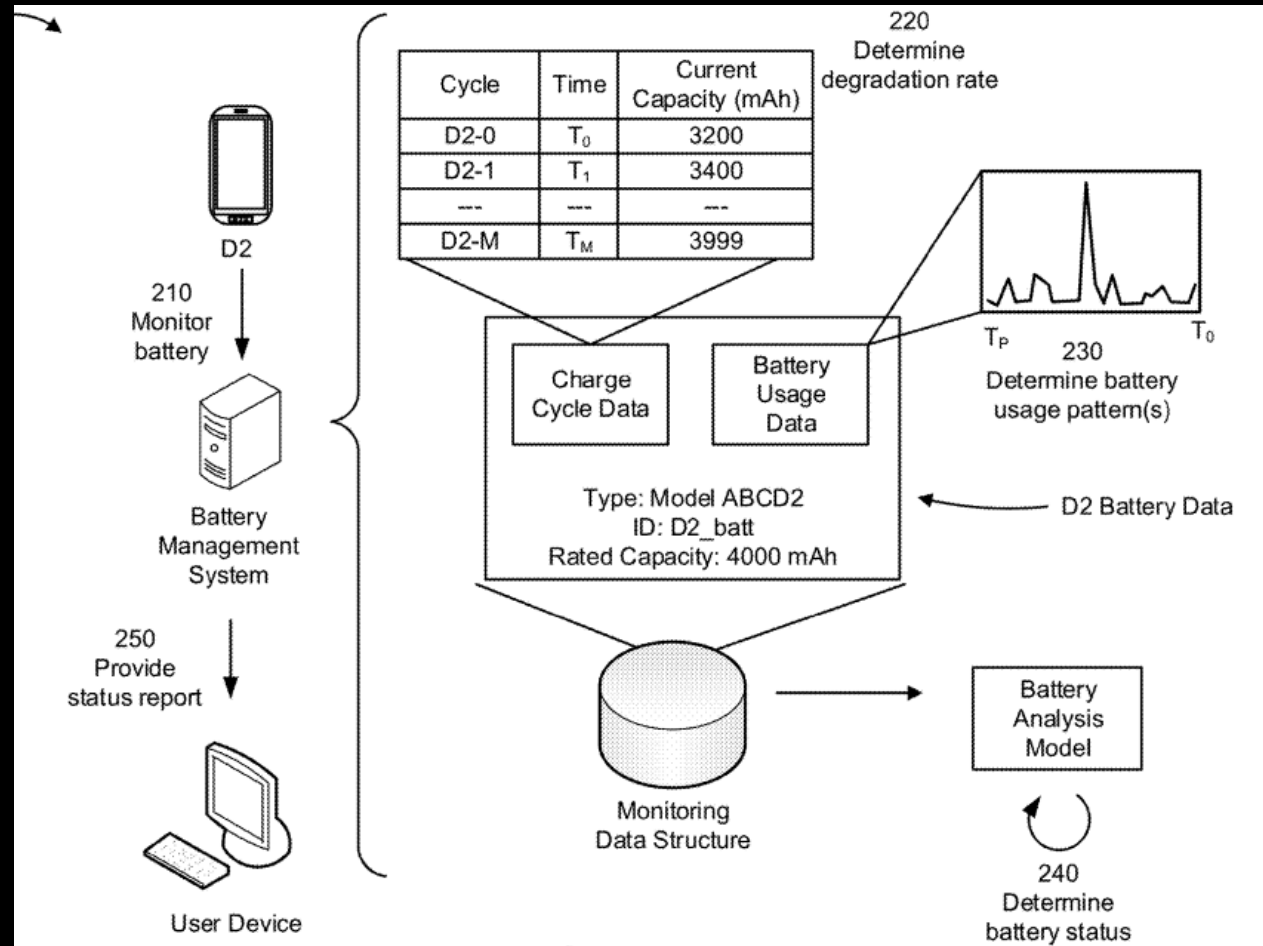


These degradation rates are then converted to number of days based on usage patterns



The degradations are nonlinear due to varying environmental factors and battery cell characteristics

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Reference - <https://patentimages.storage.googleapis.com/65/52/6d/57ad11a747540e/US20220294027A1.pdf>

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Sample Prediction Using the Model Output

Sample Calculation for Remaining Useful Life in Number of days

Assuming the battery spec sheet is as shown below,

cell number BF000315-01 and

Zebra part number asBT-000314-01

Manufacturer Rating	
specified cycle# at RT	specified capacity remaining%
500	80%

Current charge cycle of the battery(cc):	150
Current capacity of the battery(bpc):	3600
Rated capacity of the battery(brc):	4000
Calculated health percentage(bhp):	$3600/4000 = 90$
Charge cycle value obtained from the model where the battery reaches the manufacturer Recommended threshold	270
Predicted Remaining useful charge cycle(cc):	$270 - 150 = 120$
Battery usage pattern(dc/dt):	0.5
Predicted remaining useful life in days	$120/0.5 = 240$ days.

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PBR service



For a Proactive Battery Replacement (PBR) customer, VIQ supporting the PBR service will facilitate battery orders to Siebel based on the current RUL reporting for customer who have batteries with an RUL < 30 days



These batteries which are marked “replacement required” on the PBR report will be shipped to end user customer locations



Other categories could be “Good Battery” or “Excellent Battery”

Application of RUL – Planning Battery Inventory



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Input Factors for Machine learning/simulations for battery

Power Demand Profile - Demands on the battery (indicated by battery discharge in milli Ampere Hours (mAh))

Supply Profile - Present capacity of the battery (mAh that the battery can supply on full supply)

The battery present charge at start of shift (the capacity to which the battery is charged at start of shift)

Number of hours of use. Shift durations, Number of Shifts

Impact of battery degradation over time on battery requirements – RUL helps predict capacity in future quarters

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Natural Degradation of Battery

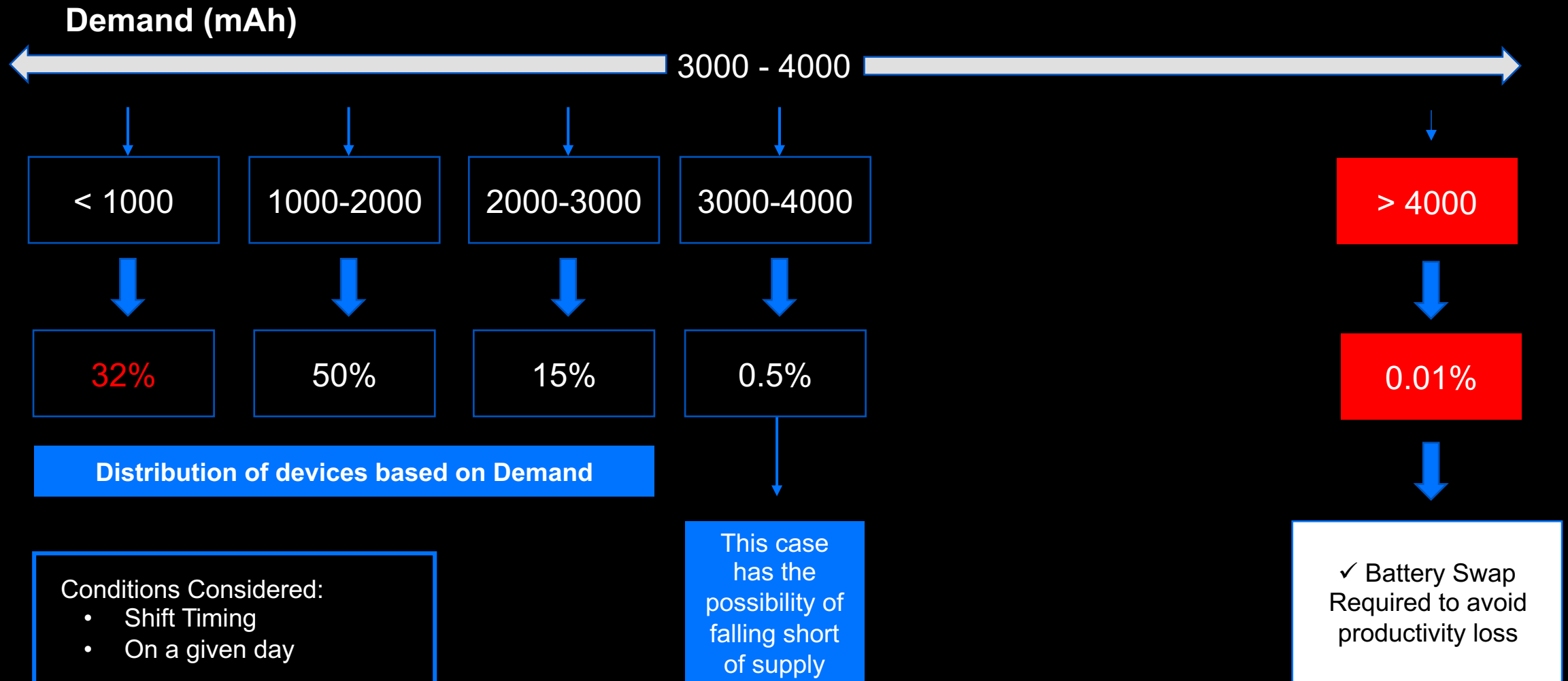


Average Capacity Over Time



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Demand and Supply Profiling Across Sites – A Sample



Sample Recommendations to minimize disruptions

Actions to Minimize Business Disruptions

		Impact			
		Q1	Q2	Q3	Q4
		Potential Productivity Loss on busy days (in hours)			
Forecast if no action is taken and Devices are charged as per associates charging behavior		56	57	58	61
Forecast if no action is taken and Devices are completely charged		0	0	0	0
Estimate if Battery replacement recommendation is followed		0	0	0	0
Estimate if Battery replacement recommendation is followed and Extra Batteries / Mobile Chargers are utilized		0	0	0	0
Recommendations	Number of Batteries to be replaced	0	0	0	0
	Number of devices to carry spare battery OR Mobile / Truck mountable charger	0			



Thank You

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