

Fraud Prevention with Machine Learning

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Outline

- Primer on Credit Card Fraud
- Data Science
 - Challenges
 - Human Intuition
 - Machine Learning
- Data Engineering
 - Real-time Metrics
 - Batch Metrics
 - Transaction Lifecycle

Primer on Credit Card Fraud



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How do you prevent fraud?

Data Science

Challenges

- Data Volume
 - 20M+ transactions per day
 - 200+ added features

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- Data Volume
 - 20M+ transactions per day
 - 200+ added features
- (Really) Unbalanced Problem
 - Positive Class: 0.1% ~ 1.0%
 - Stratified Sampling
 - Undersampling
- Time Series Problem

Human Intuition

- Real-time Features

- Distance between consecutive transactions
- Number of failed attempts in last hour
- Total money spent in last hour by merchant

Human Intuition

- Real-time Features

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- Batch Features

- Total money spent in last month
- Average transaction money by merchant in last month
- Percentage of transactions abroad in last month

Human Intuition

- Real-time Features

- Distance between consecutive transactions
- Number of failed attempts in last hour
- Total money spent in last hour by merchant

- Batch Features

- Total money spent in last month
- Average transaction money by merchant in last month
- Percentage of transactions abroad in last month

- Enrichment Features

Machine Learning

- Literature Algorithms
 - Random Forests
 - XGBoost

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 - Points Of Compromise
 - Cost Sensitive Scoring

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- Others

Data Engineering

Real-time Metrics

- Event Stream Processing
 - Data Studio
 - PQL
 - PKernel

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- Mission-critical Java
 - Zing[®] JVM (by Azul Systems[®])

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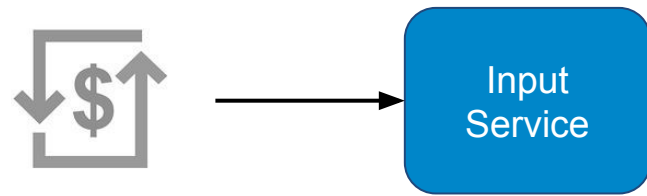
Batch Metrics

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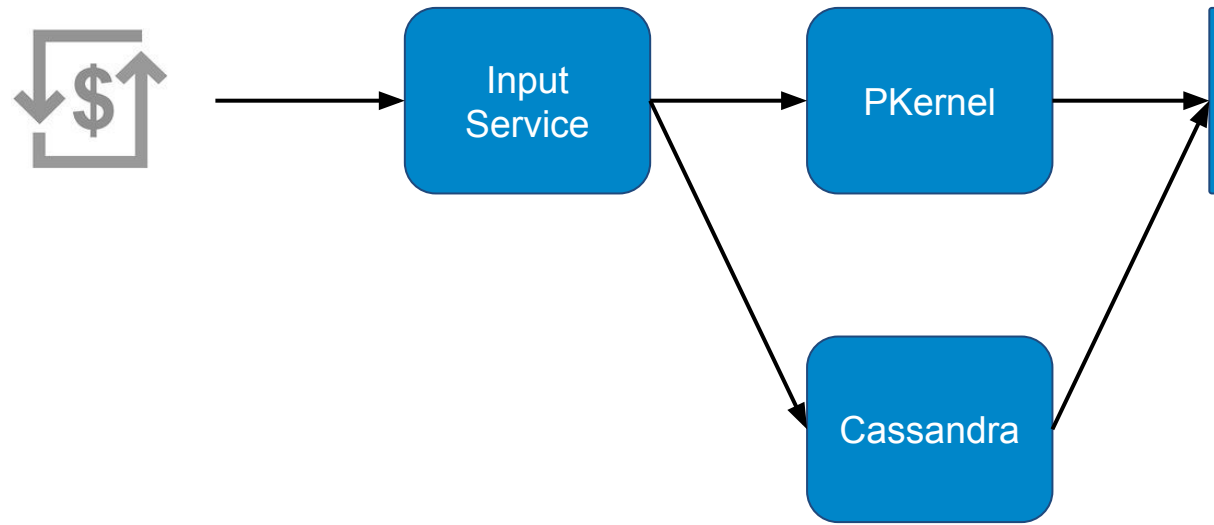
Transaction Lifecycle



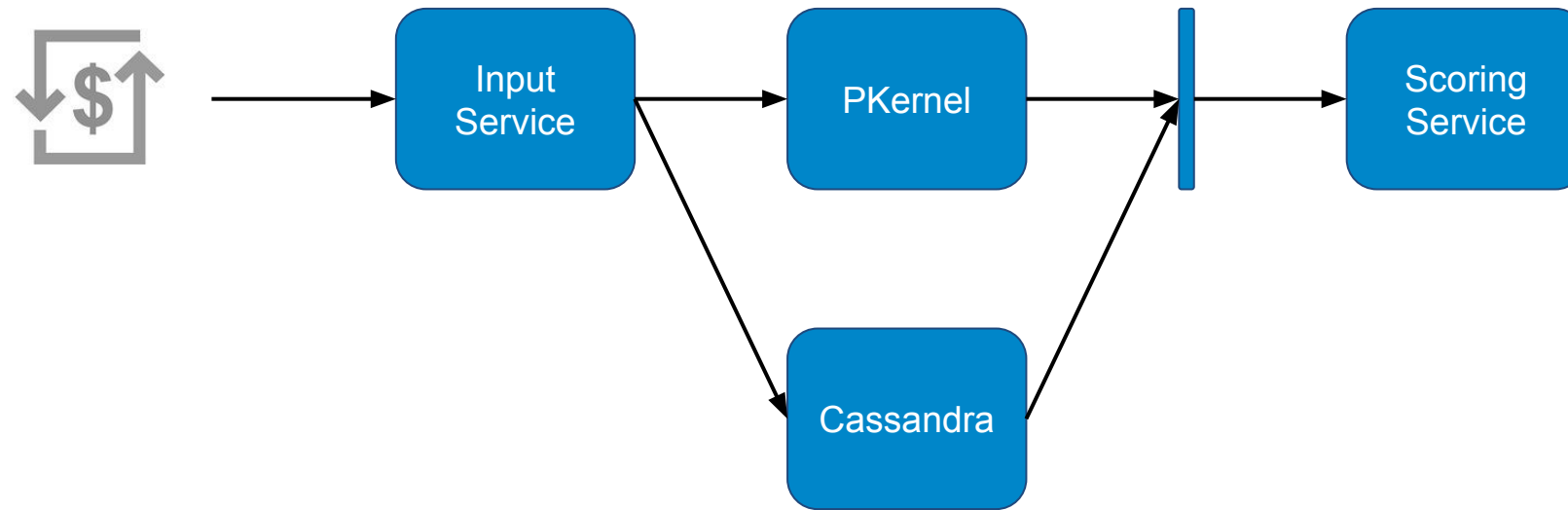
Transaction Lifecycle



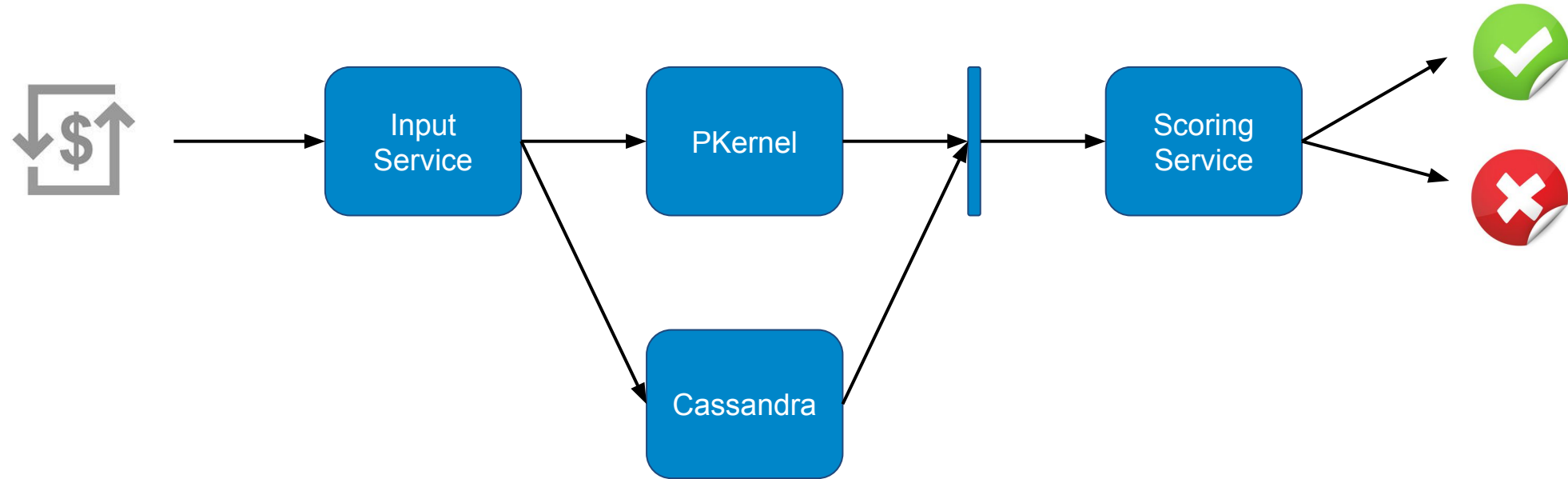
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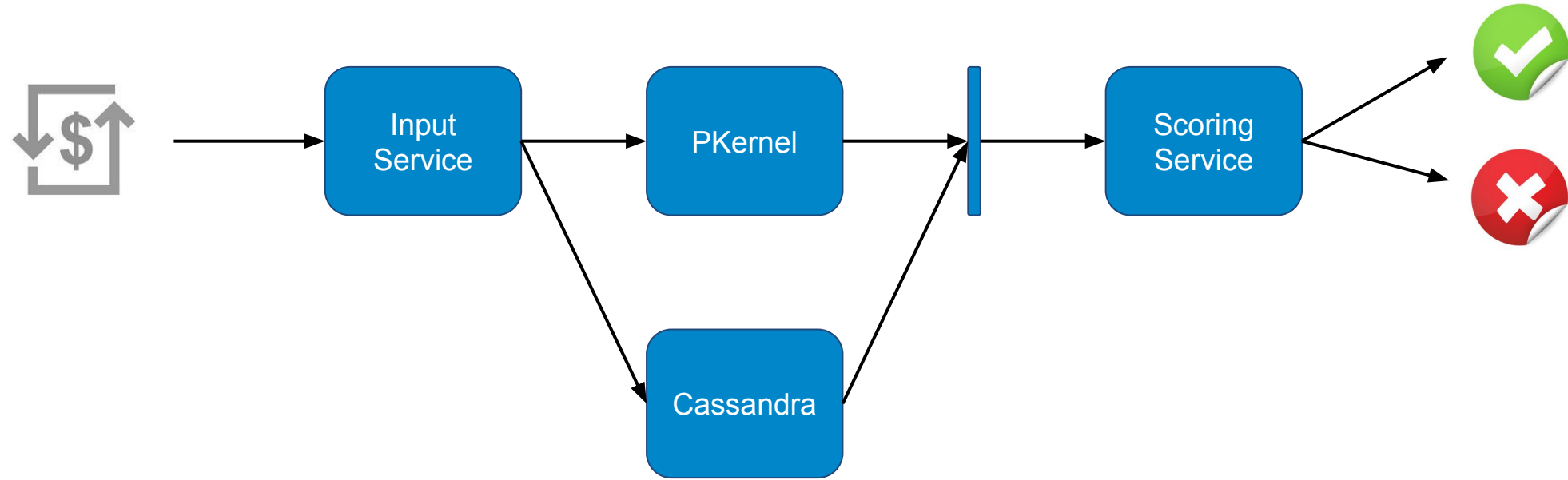
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Processing time ~ 100 ms on 99.999% percentile

Thank you!

