Data Science & Engineering

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November 2018





About Bizzy

Data Case Study

People

Process

Technology

QnA

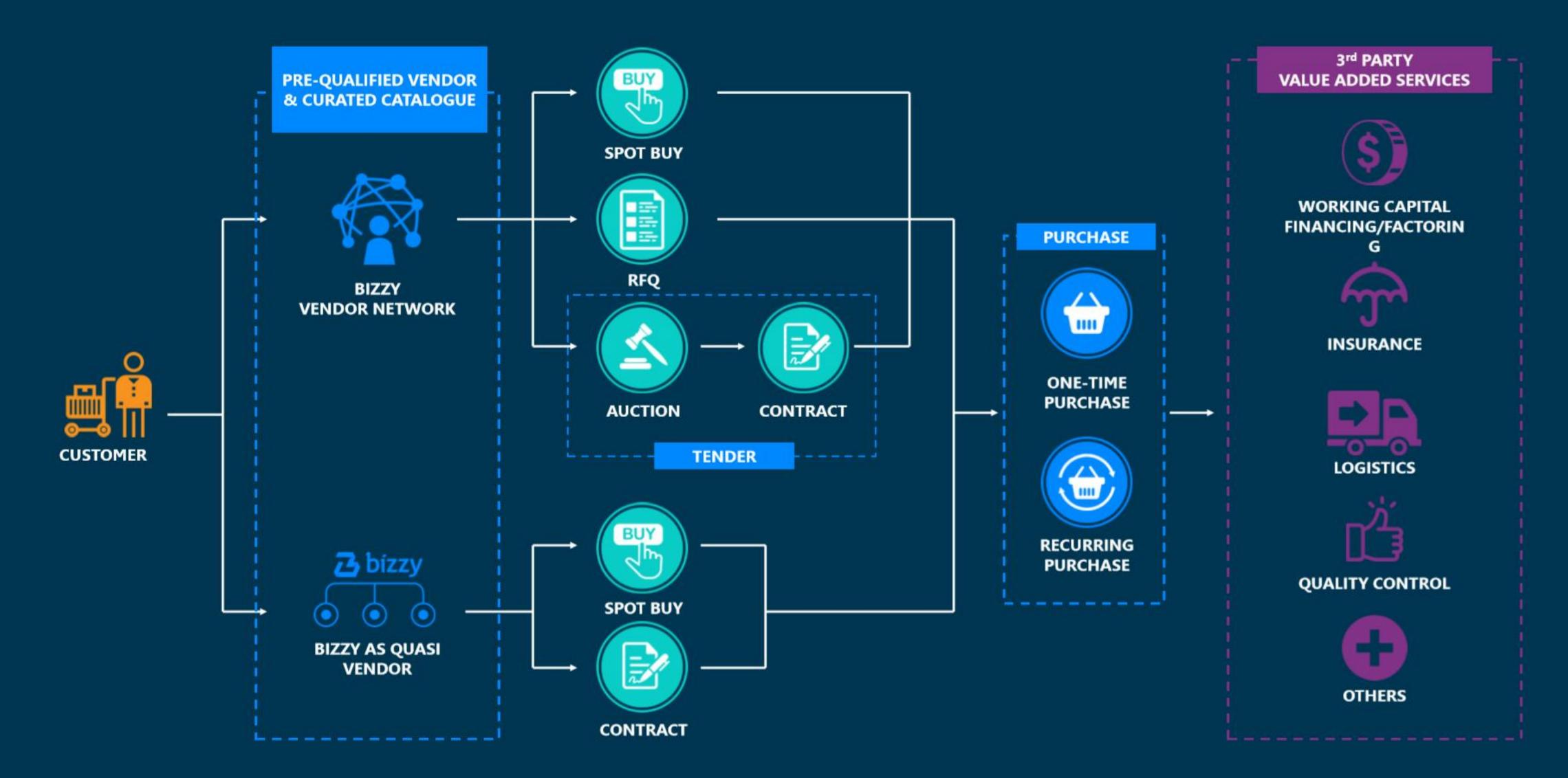


Bizzy has relentless MISSION to accomplish

To underpin a clean economy by powering an inclusive digital business ecosystem that enables transparent, efficient and accountable commerce for all stakeholders

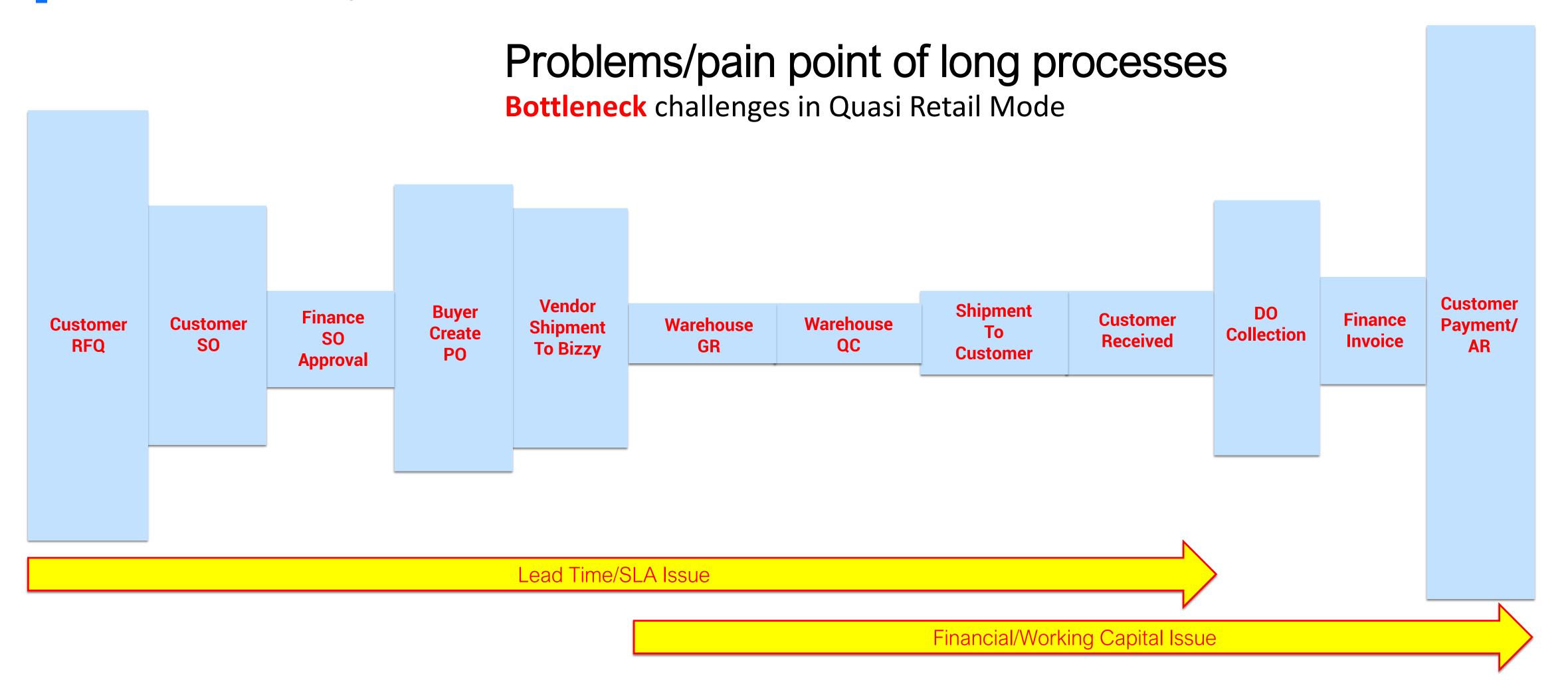


Bizzy enables execution of strategic, tactical and operational procurement activities





Data Case Study





Data Case Study

Daily Notification of end to end process

Addressing bottleneck to increase SLA

Internal Notification:

- Finance Pending Approval
- Buyer Pending PO(Back Order)
- Vendor Pending GR
- Pending Shipment
- Pending DO Collection
- Finance Pending Invoice
- Vendor Pending Payment
- Customer Pending Payment
- Etc

Add Machine Learning prediction to detect potential bottleneck item



Data Case Study

Daily Notification of end to end process

Addressing bottleneck to increase SLA

External Notification:

- Customer AR Statement Reminder
- Vendor Payment Notification
- Customer Weekly Spending Summary



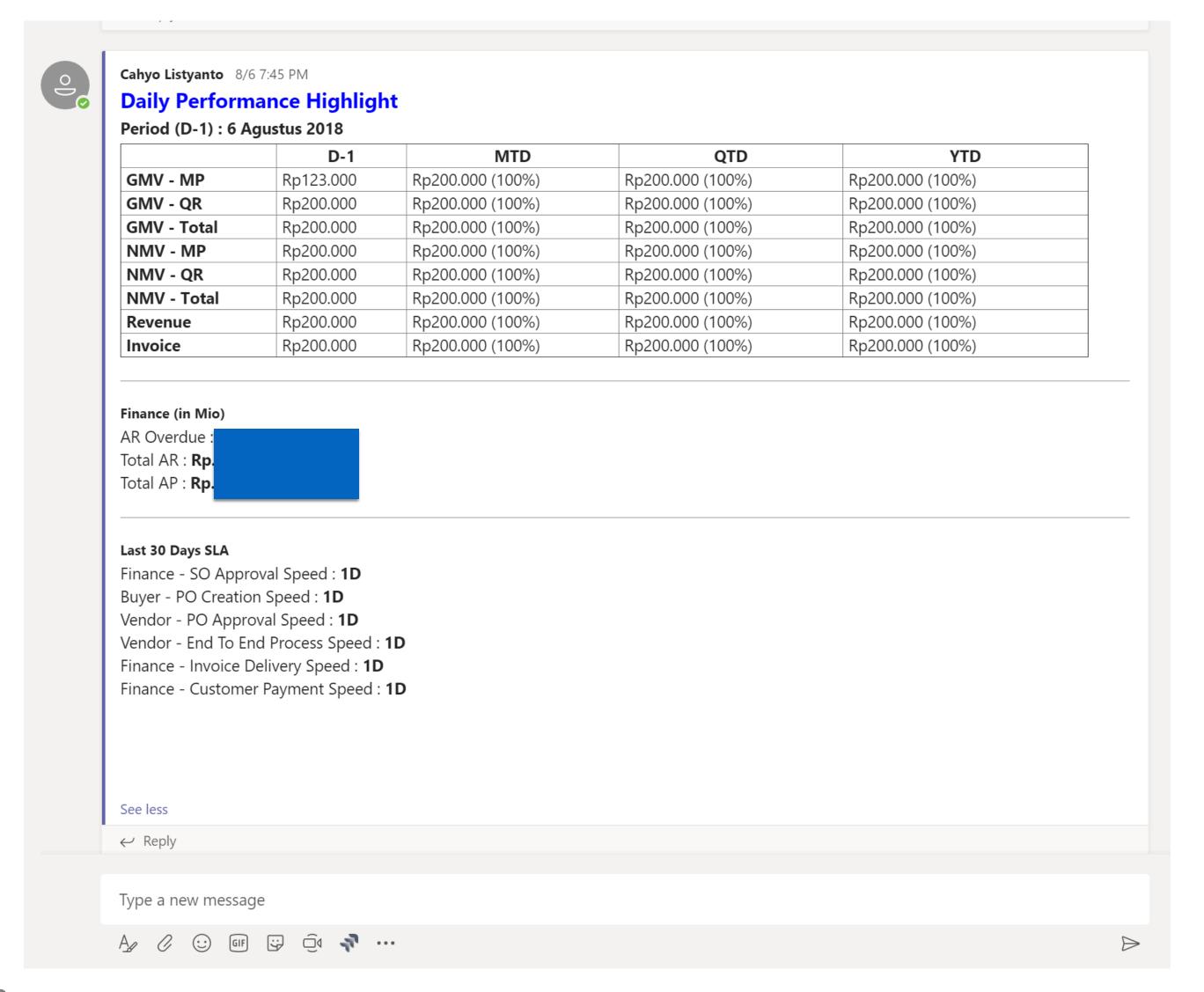
Delivery Channel

- Chat Reporting
- Email Reporting
- Reports Web Portal
- Power BI Dashboard
- Excel Reporting

Leverage channel which the user **used** in day to day operation



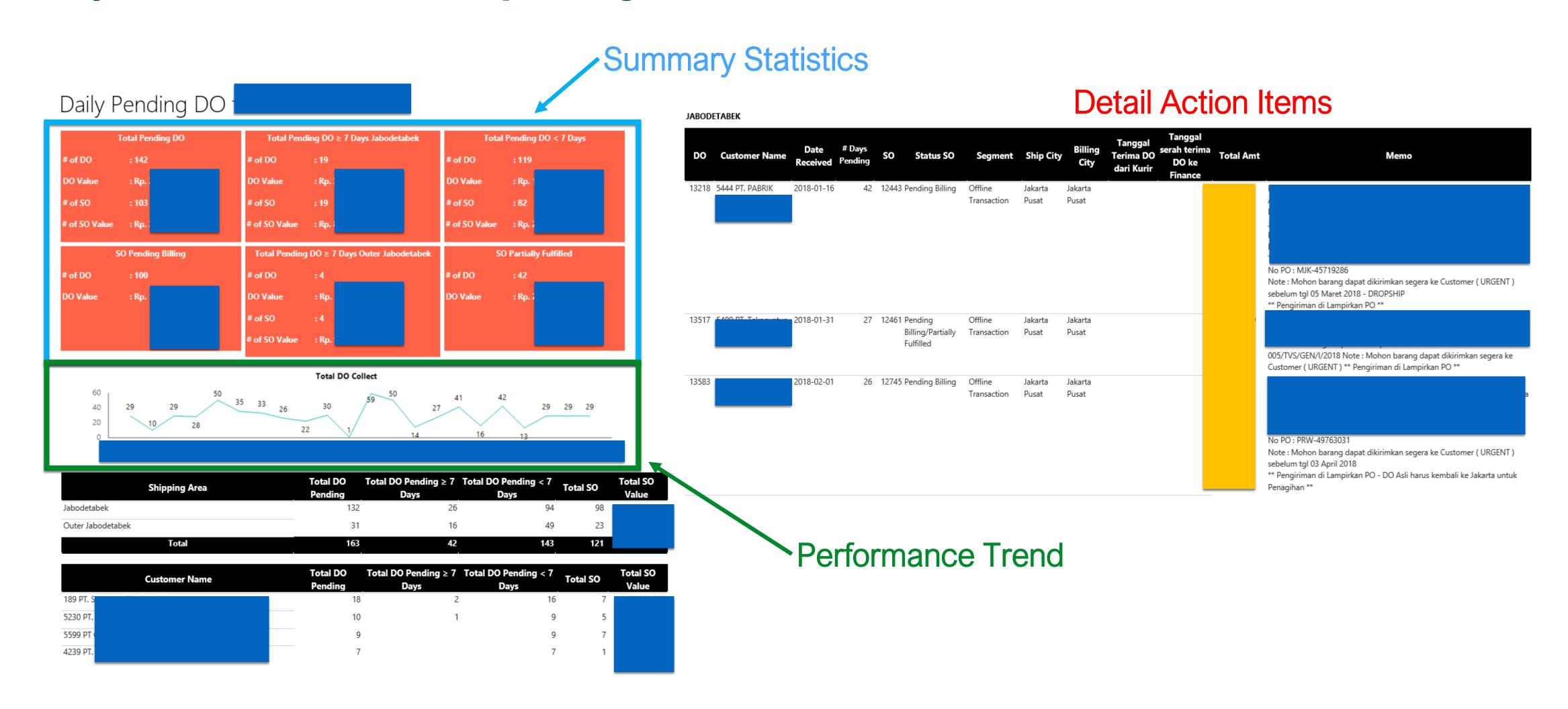
Delivery Channel - Chat/Text Messages





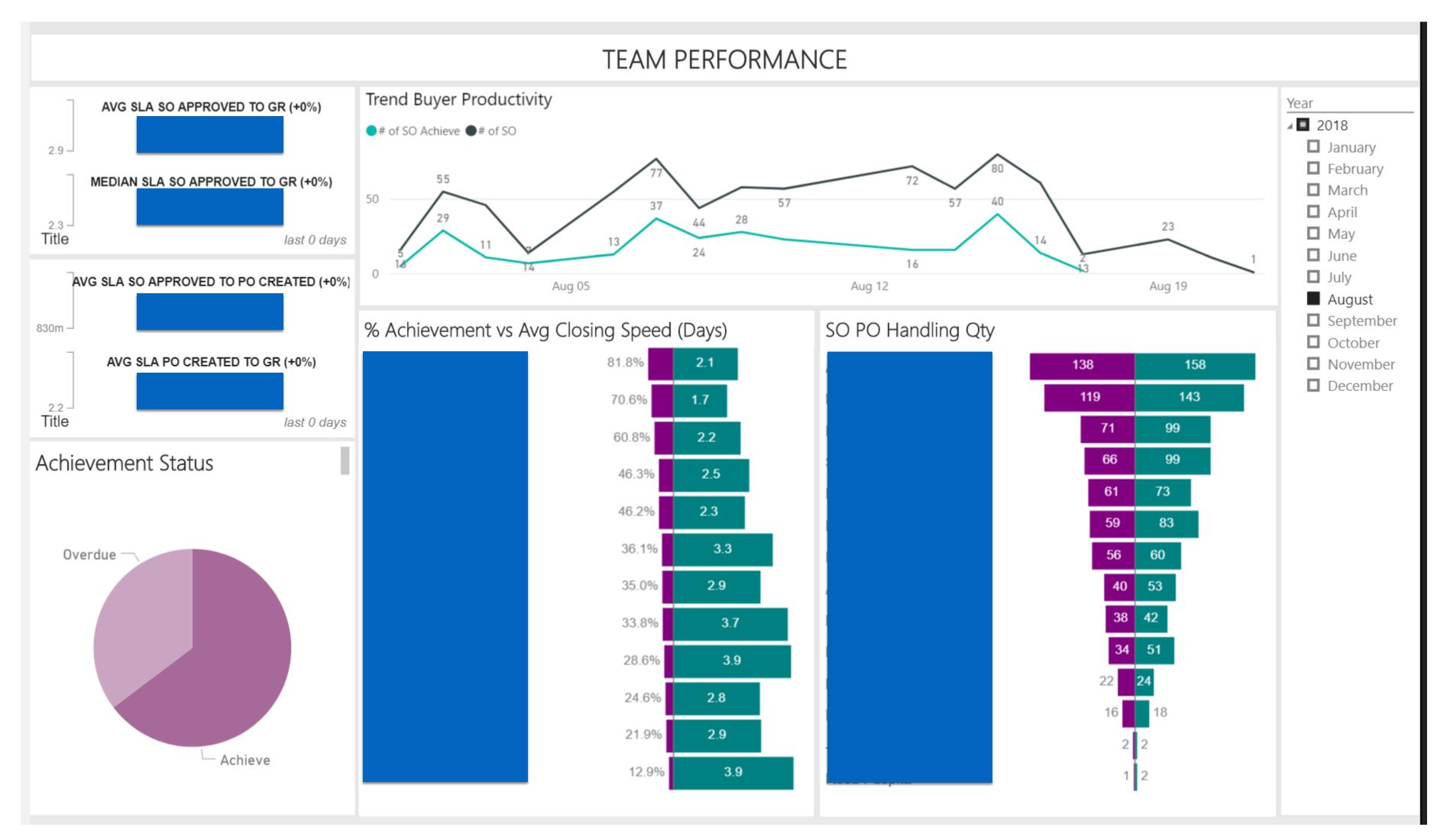


Delivery Channel – Email Reporting



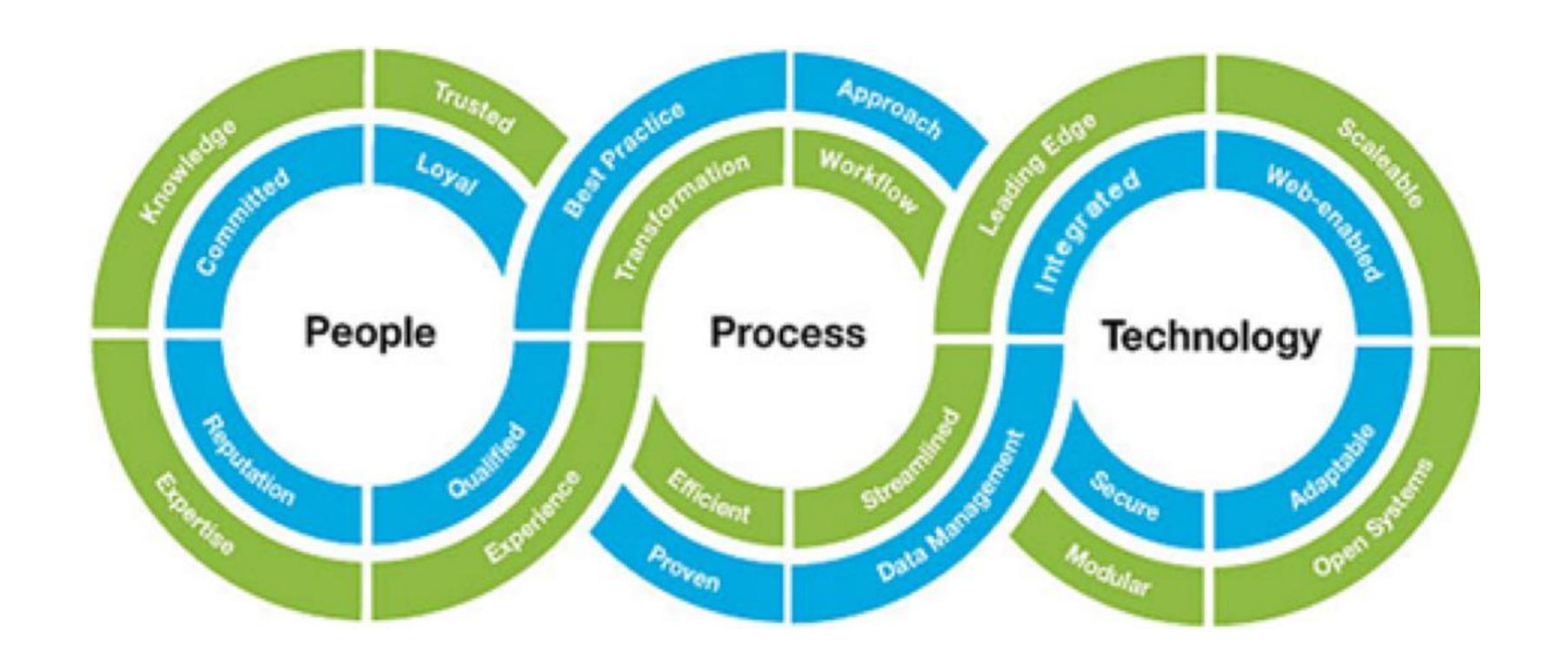


Delivery Channel – Dashboard



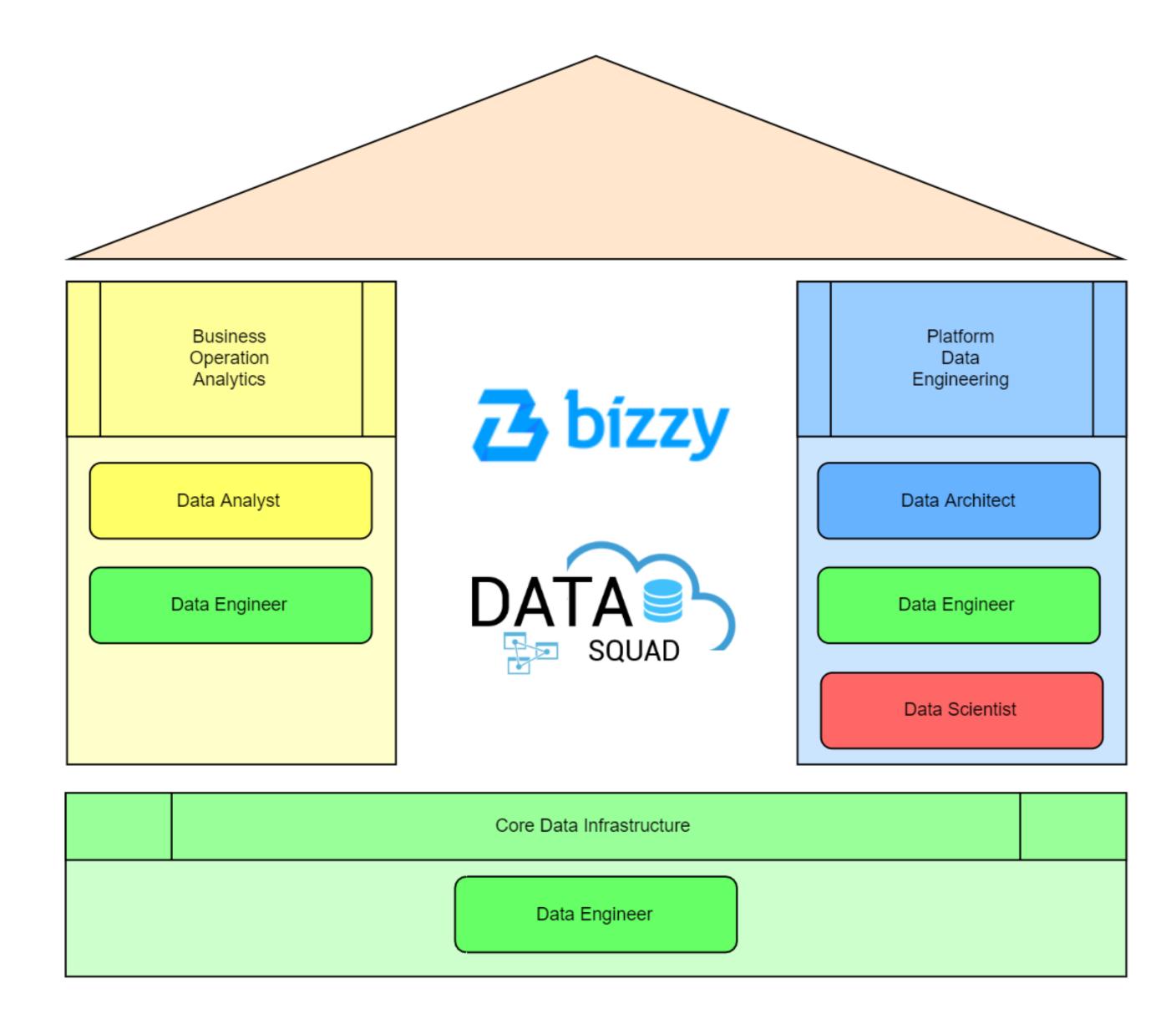


Framework for Business/Digital Transformation





People



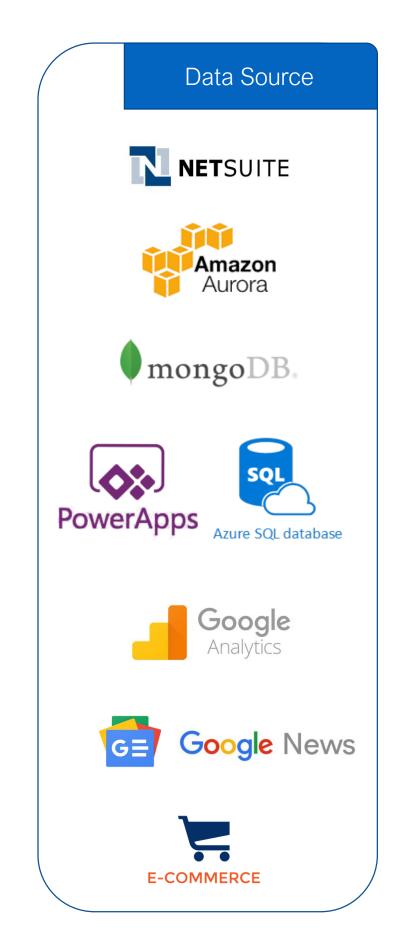


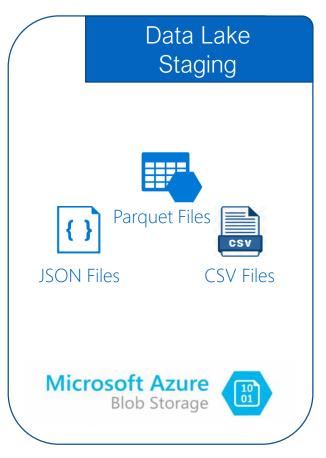
Process

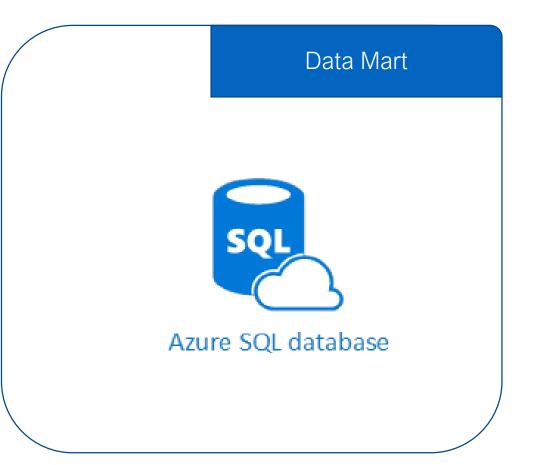
- JIRA (Agile Methodology)
 - Scrum
 - Kanban
- Confluence (Wiki/Documentation)
- Bitbucket (Repository)

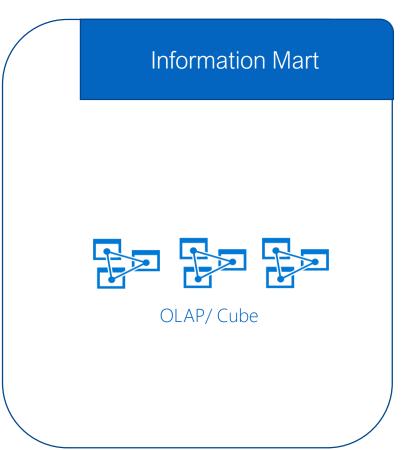


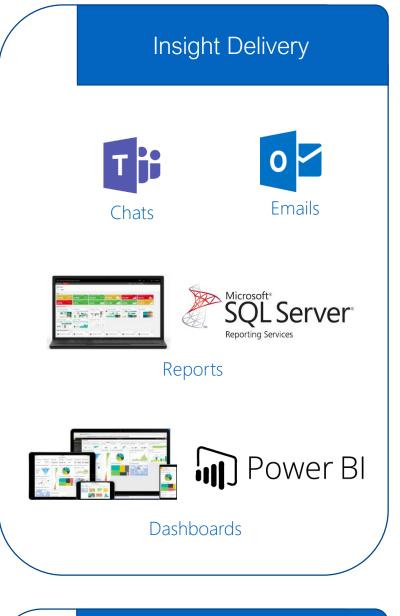
Technology – Data Architecture

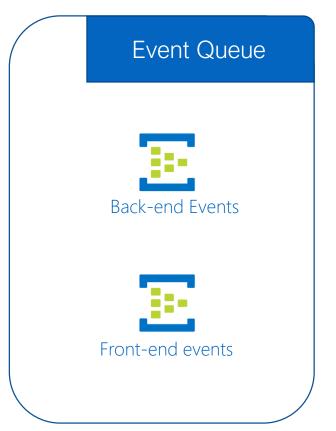


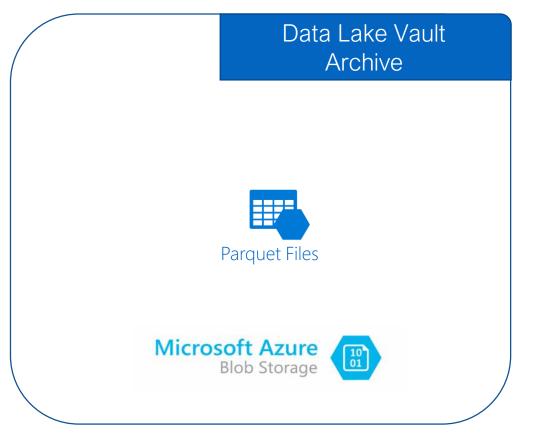


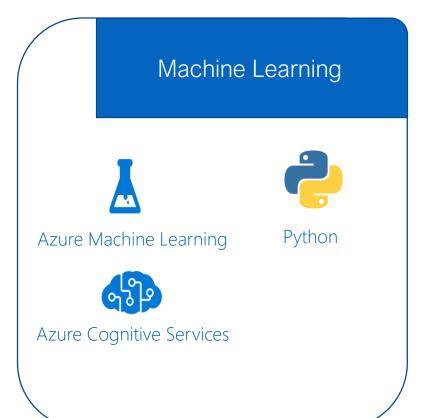


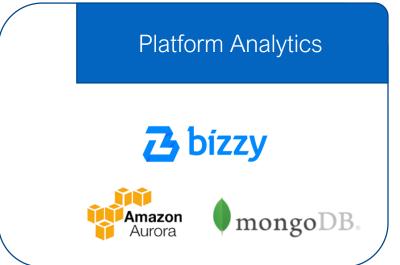


















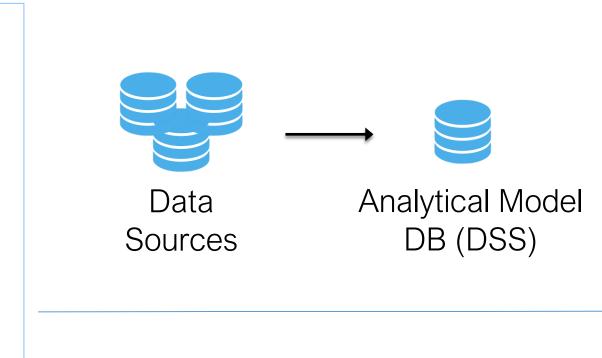
Azure Databricks Spark & Azure Data Factory

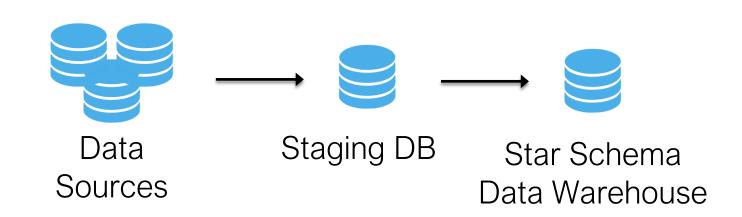


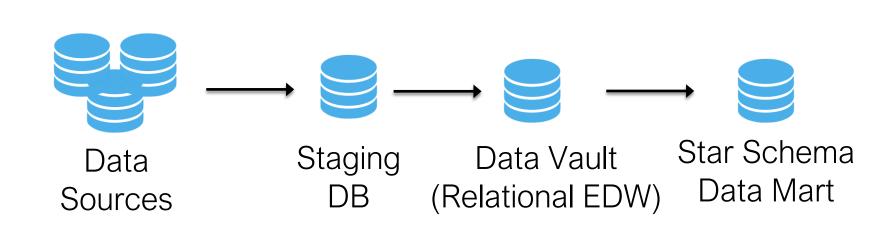


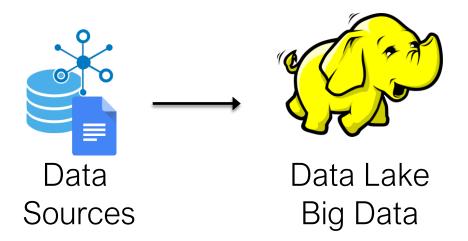


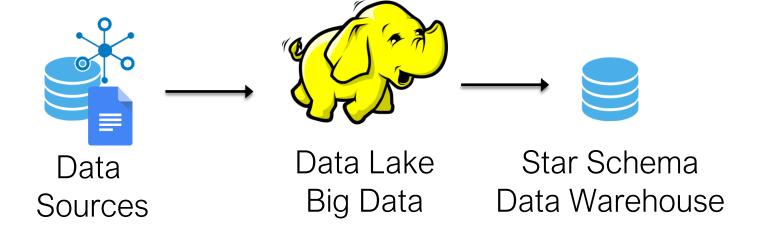
Various Data Pipelines & Technology



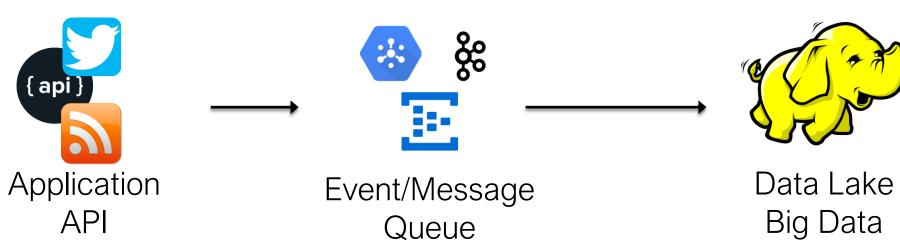














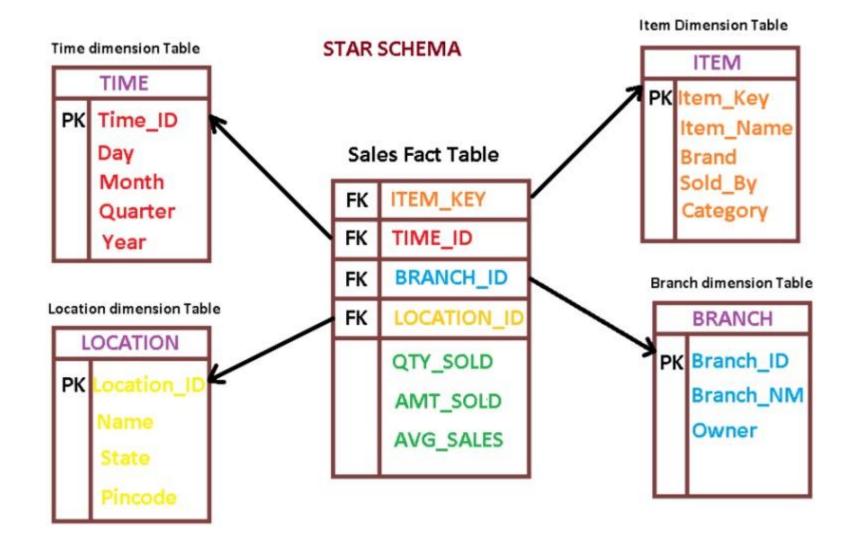
Data Warehouse



Real-time Dashboard



Data Warehouse - Star Schema





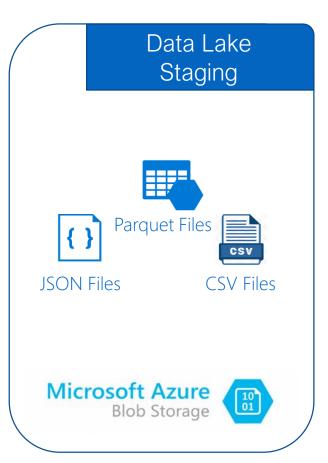
Data Warehouse - Data Vault

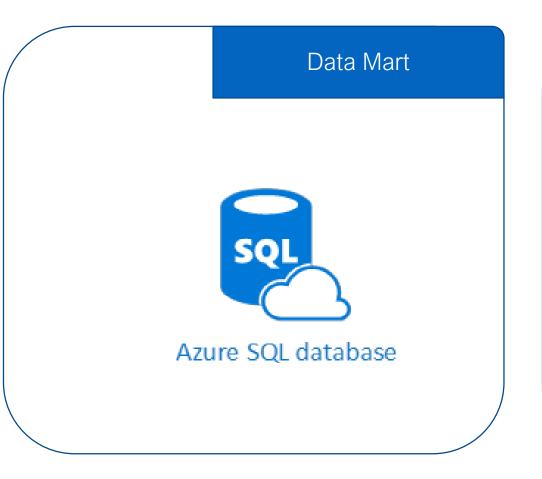
Data Vault – Hubs / Links / Satellites SatCarrierReg

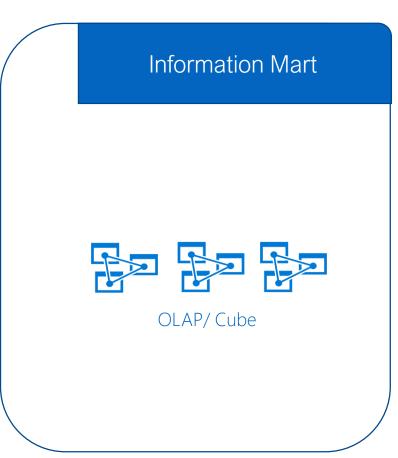


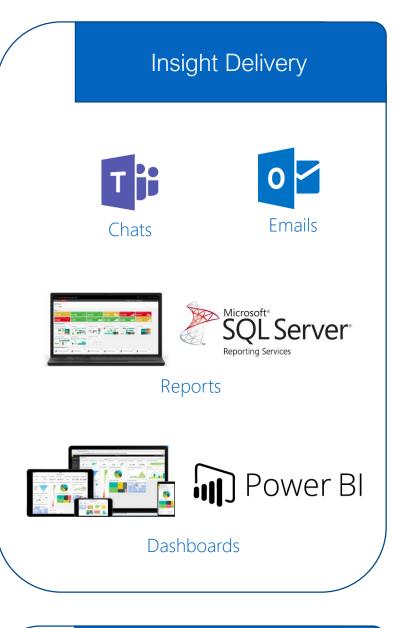
Technology – Data Architecture

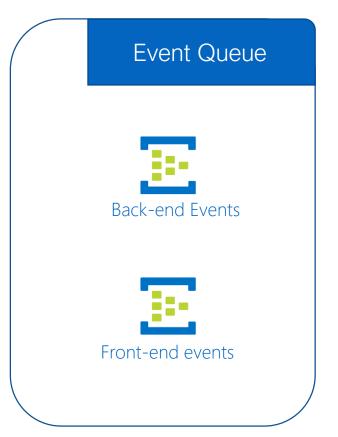


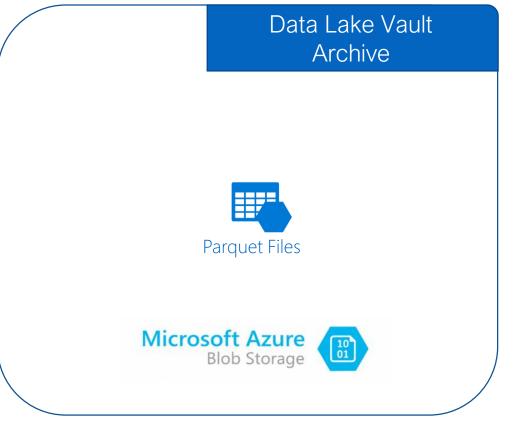


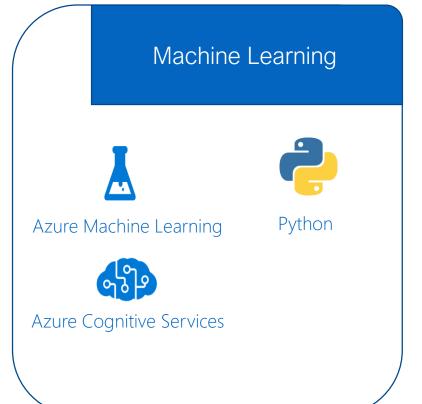


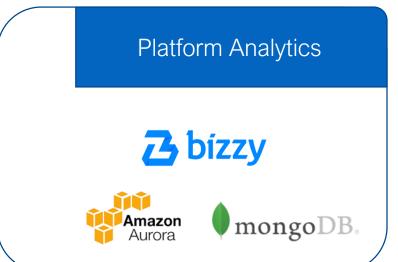


















Azure Databricks Spark & Azure Data Factory







Big Data Infrastucture

Cloud

- Leasing cost model with easier chargeback representation.
- Can be cheaper for low uptime workloads(midnight ETL). Separation of compute and storage.
- Elasticity. For example, the ability to spin up huge numbers of instances to completely run a new release while running the old release at the same time.
- Software based managed services as an option.
- Integrates well with data sources already stored in the cloud.
- No control of hardware, and limited control of software especially if you use a cloud vendor's distribution.
- Latency when interoping with on-premise resources. E.g. warehouse.
- Software level privacy, rather than hardware.
- Lock-in. This can be alleviated some if you use something like Cloudera,
 Hortonworks, or MapR.

On - Premise

- Typically cheaper for non-elastic workloads at the moment.
- Since most Cloud providers use some type of network storage, performance is typically better for a bare-metal based deployment.
- Full control of the Hadoop hardware + software.
- Latency to and from systems you integrate with can be minimized.
- Physical data isolation and privacy.
- Infrastructure managed services. E.g. smart hands to install servers and network.
- Barrier to entry is typically higher since there is commonly "hardware friction" in getting new infrastructure on the floor and operating. (Private Clouds can help reduce "hardware friction" for proof of concepts and development/testing environments.)
- Chargeback models for Hadoop can be complex.



Spark ETL Generator – Source to Staging

```
val srcDb = "catalog"
val tgtDb = "stg_phx_catalog"
val tables = List("attribute_code","attribute_set","attribute_value","brand", "category", "category_temp", "misc", "uom",
"product_group", "product_group_attribute", "product_variant", "product_variant_migration", "product_vendor",
"product_vendor_migration", "stocking_uom")
import com.bizzy.dbrk.etl.JdbcEtlProcess

val notebookPath = dbutils.notebook.getContext.notebookPath.get
val jdbc = new JdbcEtlProcess(notebookPath, spark, jdbcUrl, connectionProperties)
jdbc.process(srcDb, tgtDb, tables, "overwrite", 3, 1000)
```



Spark ETL Generator – Staging to Archive

```
val staging = "stg_phx_catalog"
val archive = "dlv_phx_catalog"
val tables = List("brand", "category", "category_temp", "product_group", "product_variant", "product_vendor")
for(table <- tables ) {
    Archive.archive(s"$staging.$table", s"$archive.$table", s"$tmp.$table", id, date, "phx")
}</pre>
```



THANKS! QnA

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