

Computational Environment

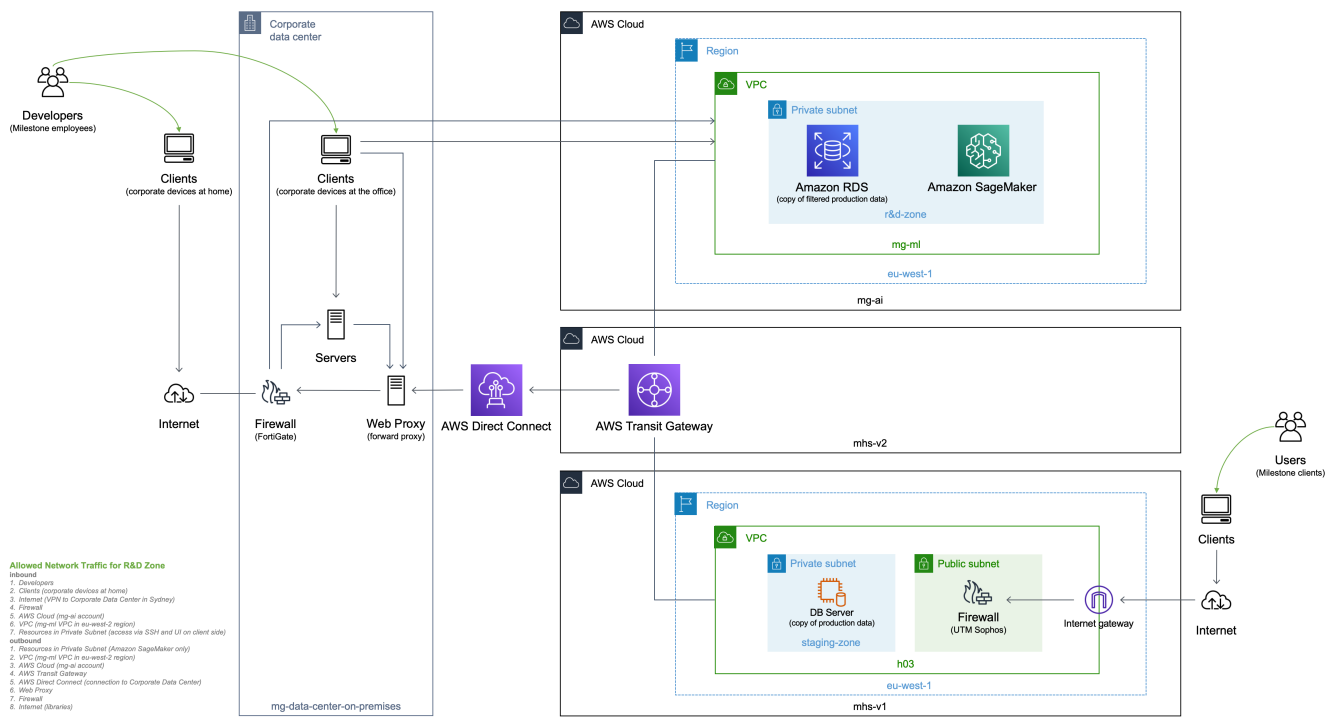
 ... developing machine learning in software development environments

- building infrastructure and pipelines for Data-Driven Science
- this requires architecture that allows to create experimental models (prototypes) and operationalise such innovation into products

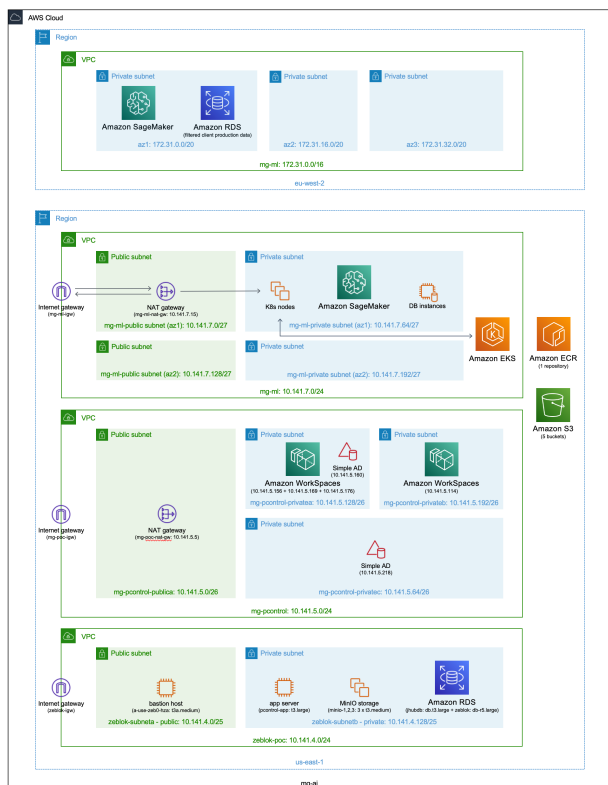
R&D Zone

- AWS account ID: mg-ai

access



resources



MLOps

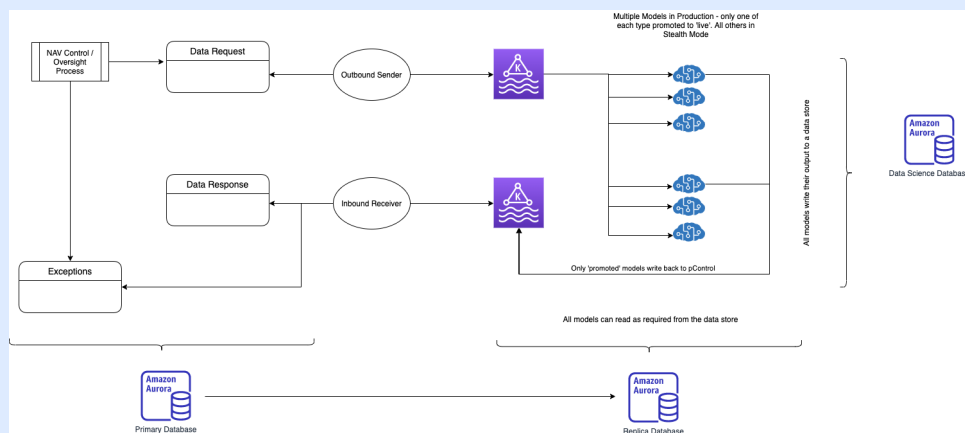
- data pipelines via network traffic between private subnets (staging and R&D zones) and corporate devices for data analysis, model design, and model deployment

pipeline	motivation	approach
data analysis	<i>analysing the quality of production data for design of experimental ML models</i>	<ol style="list-style-type: none"> exporting the copy of production data from clients to their Staging Zone filtering the data in the Staging Zone to select the data for analysis <ul style="list-style-type: none"> data for the client approved list of funds without any any modifications copying the filtered data from the Staging Zone to R&D Zone analysing the data quality for ML experiments in the R&D Zone
model design	<i>designing experimental ML models with production data</i>	<ul style="list-style-type: none"> designing ML models in the R&D zone machine learning workflow <ul style="list-style-type: none"> development via SDKs (Software Development Kits) on corporate devices computation & deployment via programatic, cloud-native approaches <ul style="list-style-type: none"> CaC (Configuration as Code) to manage and configure pipelines IaC (Infrastructure as Code) to manage and provision infrastructure <div> <p>The diagram illustrates the MLOps workflow across three stages:</p> <ul style="list-style-type: none"> development: Data (Amazon RDS, CSV, NEWS) flows into Code (PyTorch, docker). computation: Code flows into Repository (Bitbucket, AWS ECR), then into Workload (Kubeflow, Amazon EC2, kubernetes). deployment: Workload flows into Artifact (Amazon S3), then into Endpoint (TorchServe, KFServing, Istio, Amazon EC2, kubernetes). </div>

model
deployme
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*deploying all ML models in
a single production
environment per region and
continuously testing the
performance of multiple ML
models*

- promoting ML models into production via our ML Data Store
- two modes of deployment into production environments
 1. live mode - ML models are consuming production data from clients and sending the results back to their pControl applications and our ML Data Store
 2. stealth mode - ML models are consuming production data from clients and sending the results to our ML Data Store only (no client exposure)



- this should significantly improve our deployment capabilities, e.g. we will not need to go through multiple sign offs and other lengthy procedures between environments (UAT, prod, etc.) to just test a single model