

#### **ML Metadata with TFX**

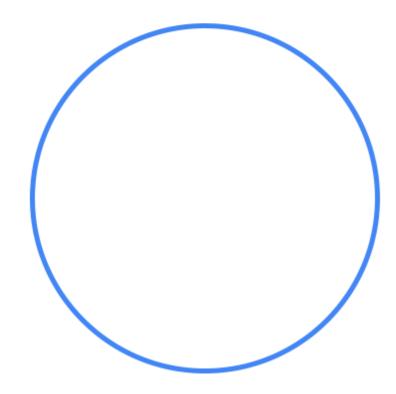
Doug Kelly
ML Solutions Engineer, Google Cloud

# Agenda

TFX pipeline metadata



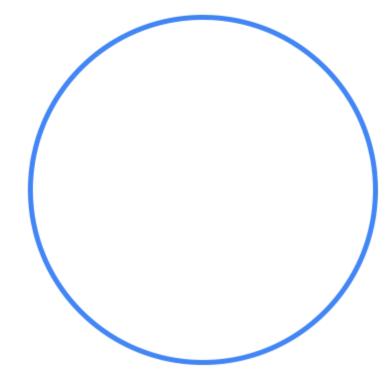
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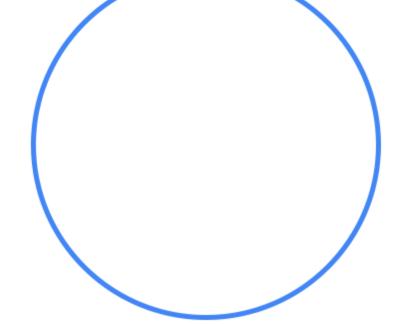




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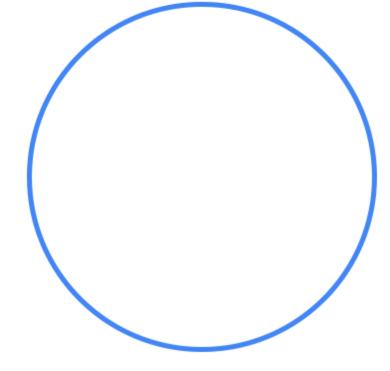


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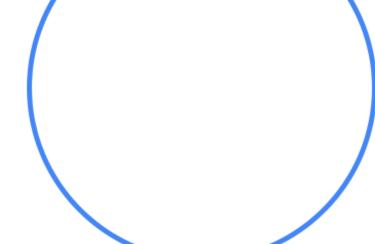
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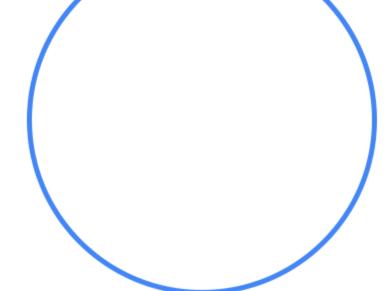
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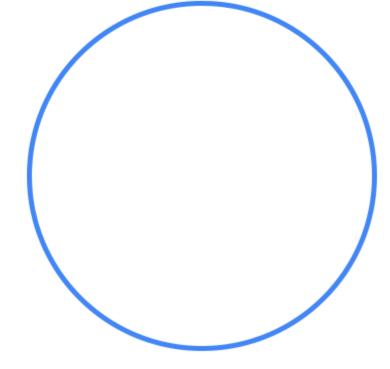
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Where is the model file stored?

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Why was model A preferred over model B?

**How** was the training environment configured?

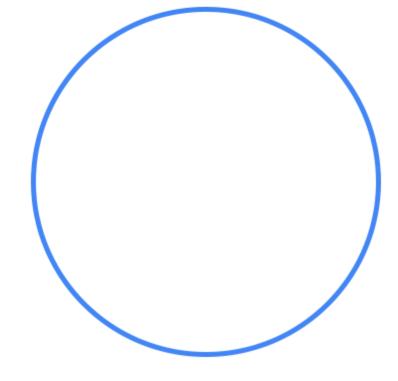




#### What is in TFX's Metadata Store?

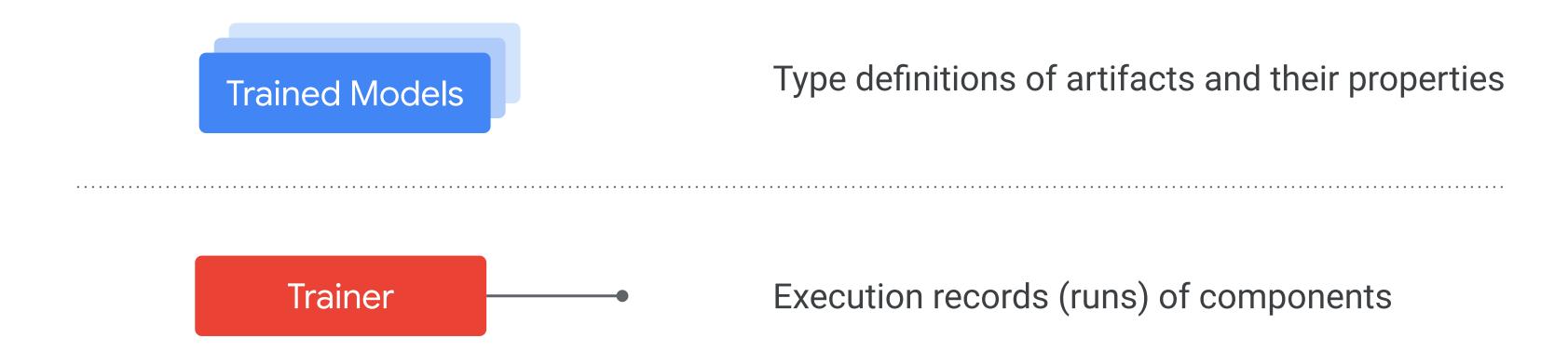
Trained Models

Type definitions of artifacts and their properties



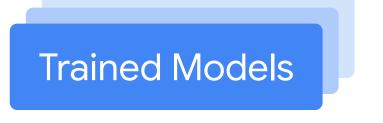


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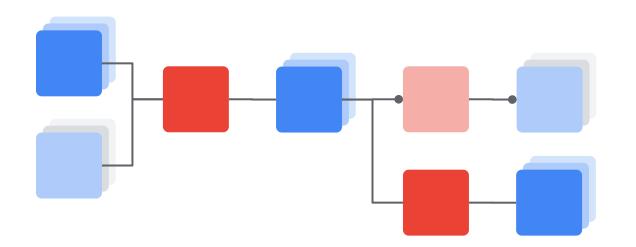
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Type definitions of artifacts and their properties



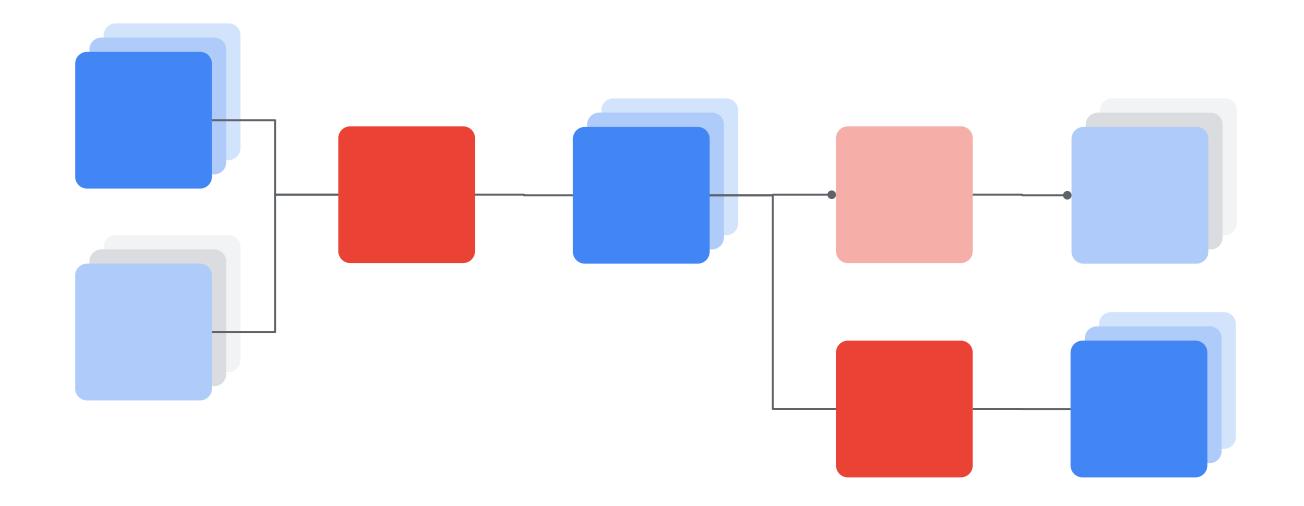
Execution records (runs) of components



Data provenance across all executions

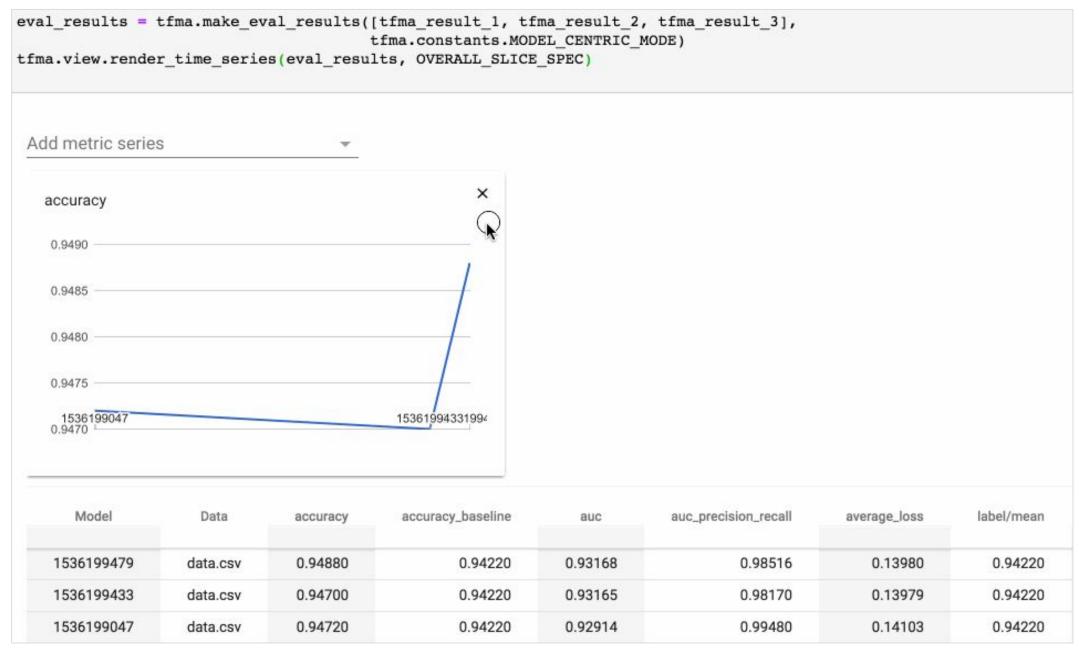


Trace model runs back to the data it was trained on



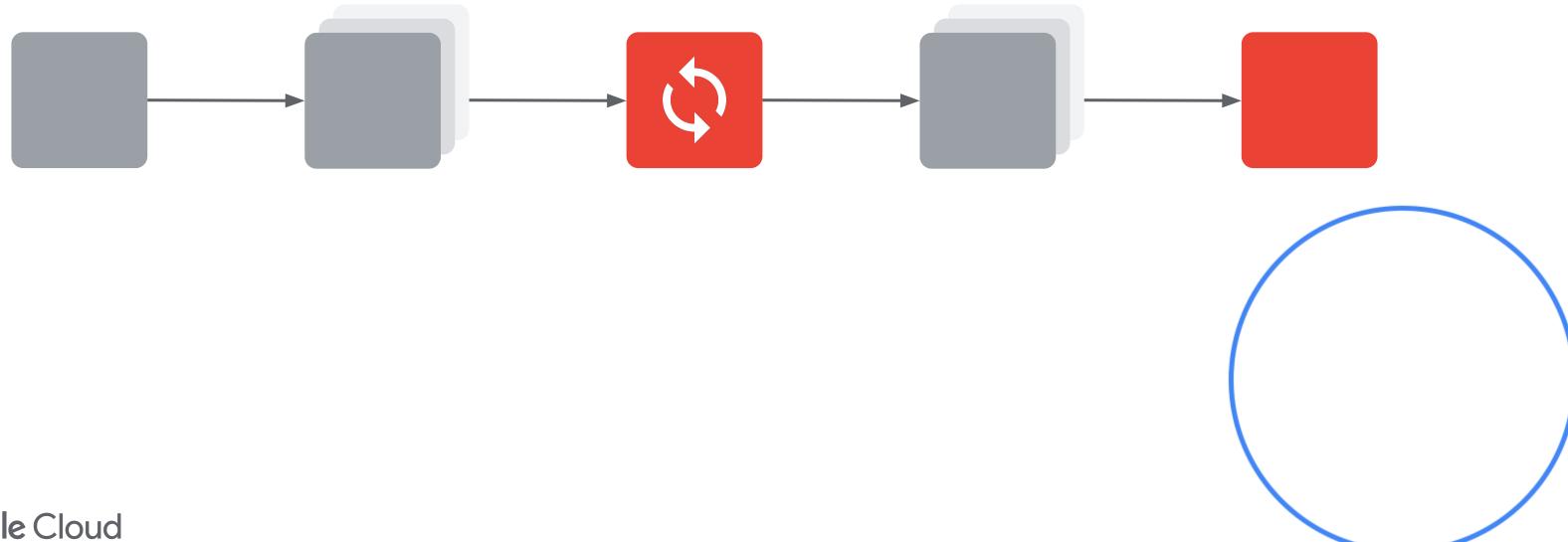


#### Compare previous model runs



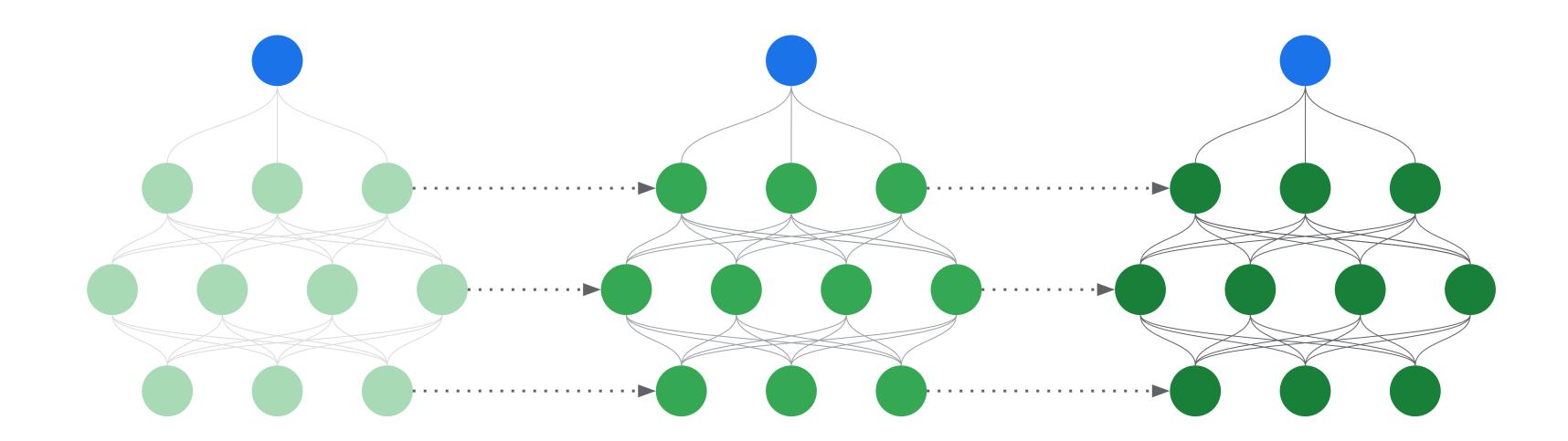


Re-use previously computed outputs





Re-use previously computed outputs





### Advanced TFX use case: pipeline warm starting

```
# Get the latest model so that we can warm start from the model.
 latest_model_resolver = ResolverNode(
     instance_name='latest_model_resolver',
     resolver_class=latest_artifacts_resolver.LatestArtifactsResolver,
     latest_model=Channel(type=Model))
 # Uses user-provided Python function that implements a model using TF-Learn.
 trainer = Trainer(
    module_file=module_file,
     transformed_examples=transform.outputs['transformed_examples'],
     schema=schema_gen.outputs['schema'],
     base_model=latest_model_resolver.outputs['latest_model'],
     transform_graph=transform.outputs['transform_graph'],
     train_args=trainer_pb2.TrainArgs(num_steps=10000),
     eval_args=trainer_pb2.EvalArgs(num_steps=5000))
```

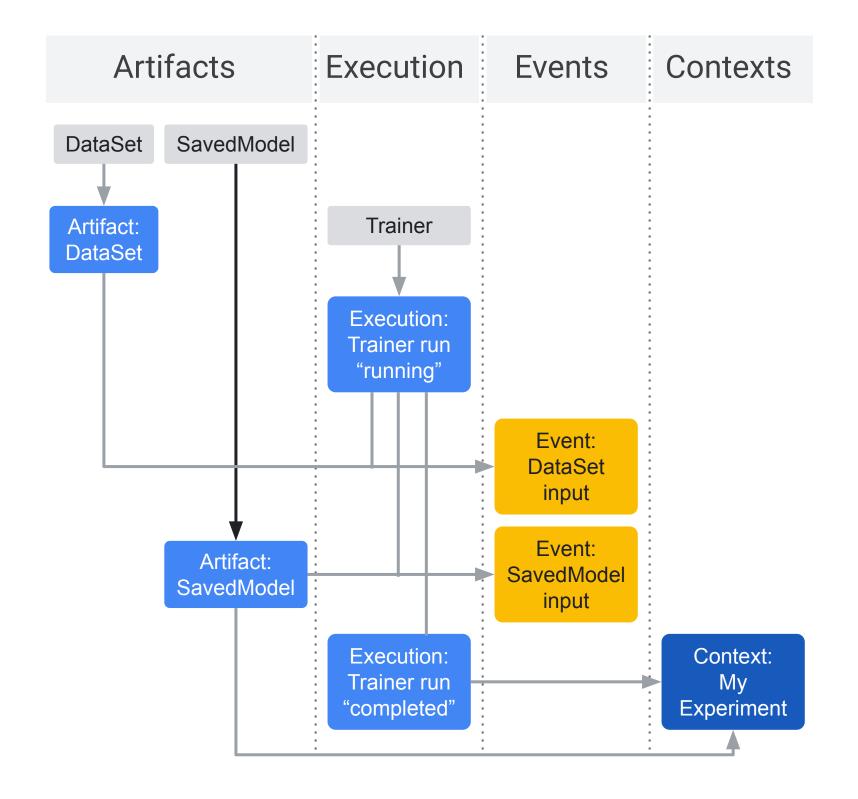


### ML Metadata (MLMD) library data model

ML Platform/Workflow (e.g., TFX) Graphical Pipeline components/steps User MLMD client libraries Interface Metadata and lineage ..... about input/output artifacts Metadata Store Context: ContextType **Association** Event: input **Attribution** Execution: ExecutionType Artifact: ArtifactType Event: output MySQL Source **SQLite Source** SQLite (on disk) MySQL SQLite (in memory)

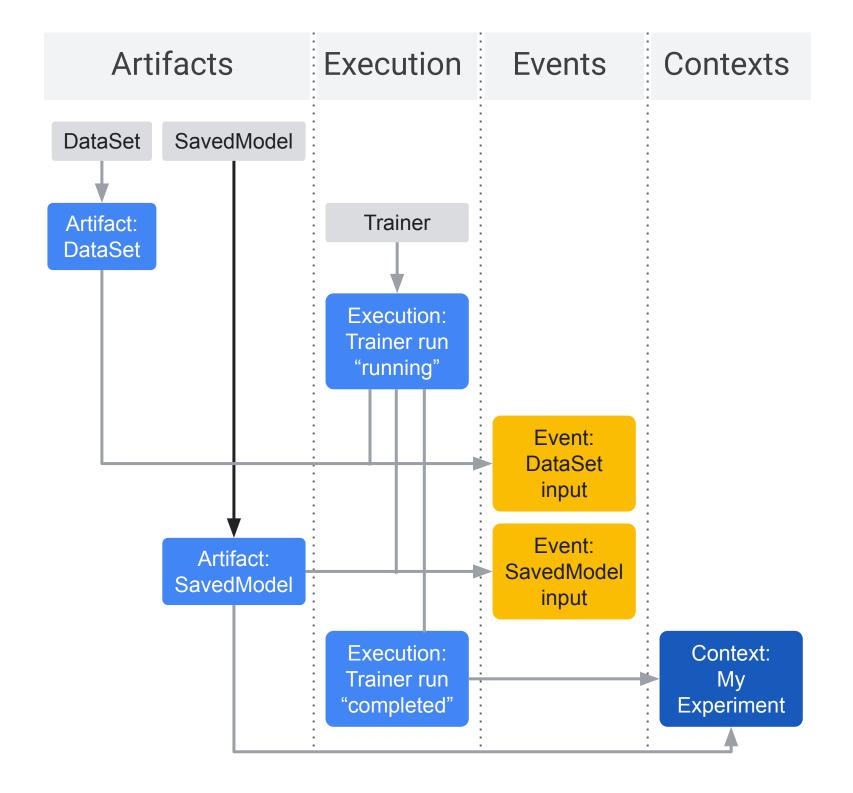


1. Register ArtifactTypes.



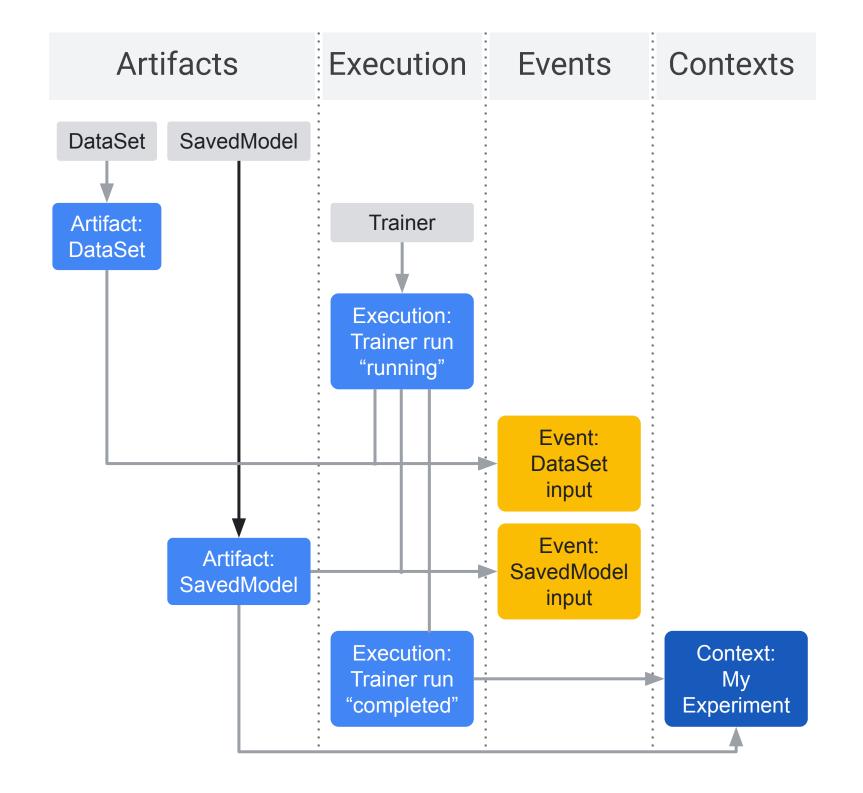


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- 2. Register ExecutionTypes.



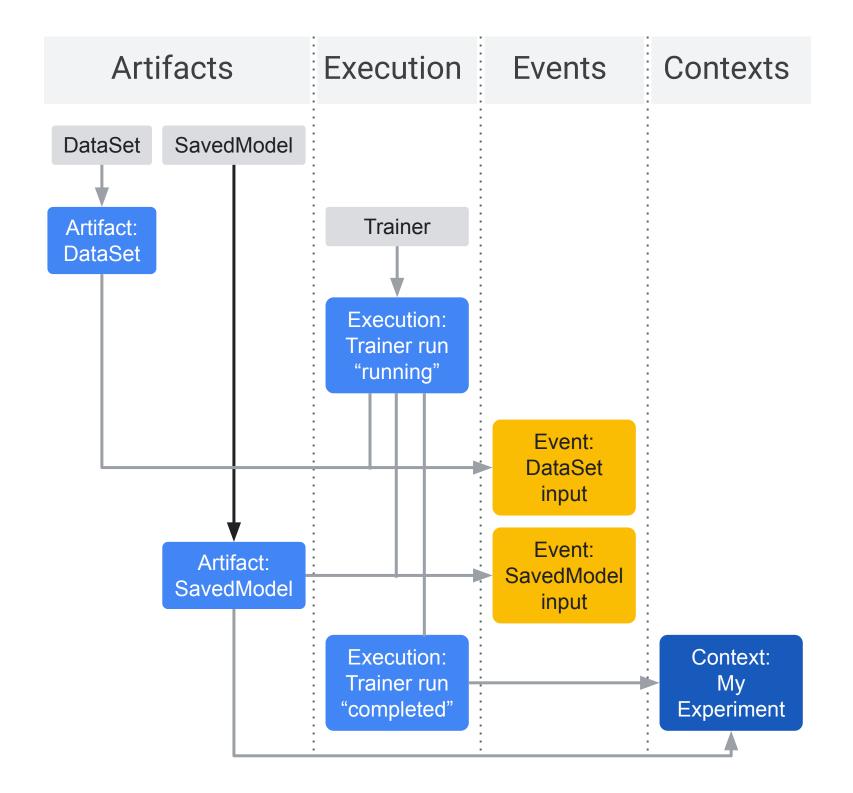


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- 2. Register ExecutionTypes.
- 3. Create DataSet Artifact.



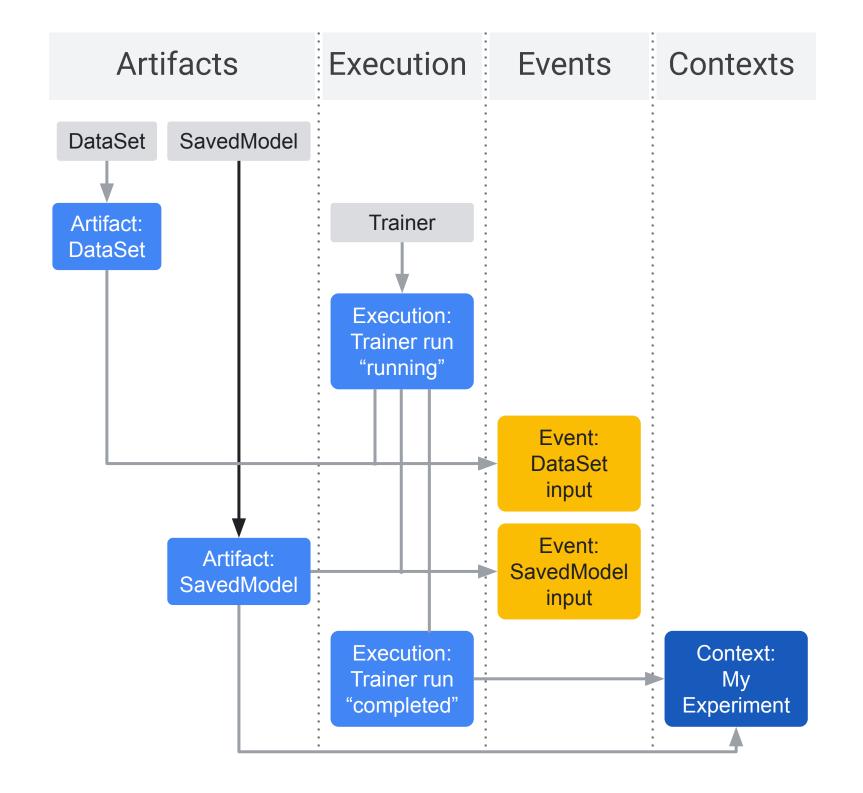


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- 3. Create DataSet Artifact.
- 4. Create Execution for Trainer.



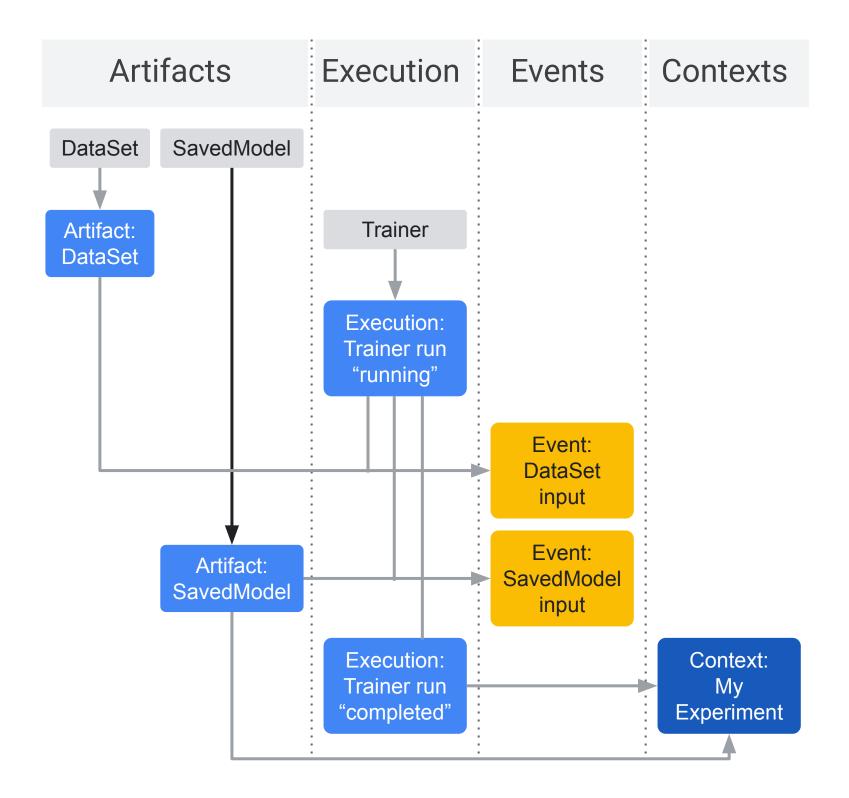


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- 3. Create DataSet Artifact.
- 4. Create Execution for Trainer.
- 5. Read DataSet and record input event.



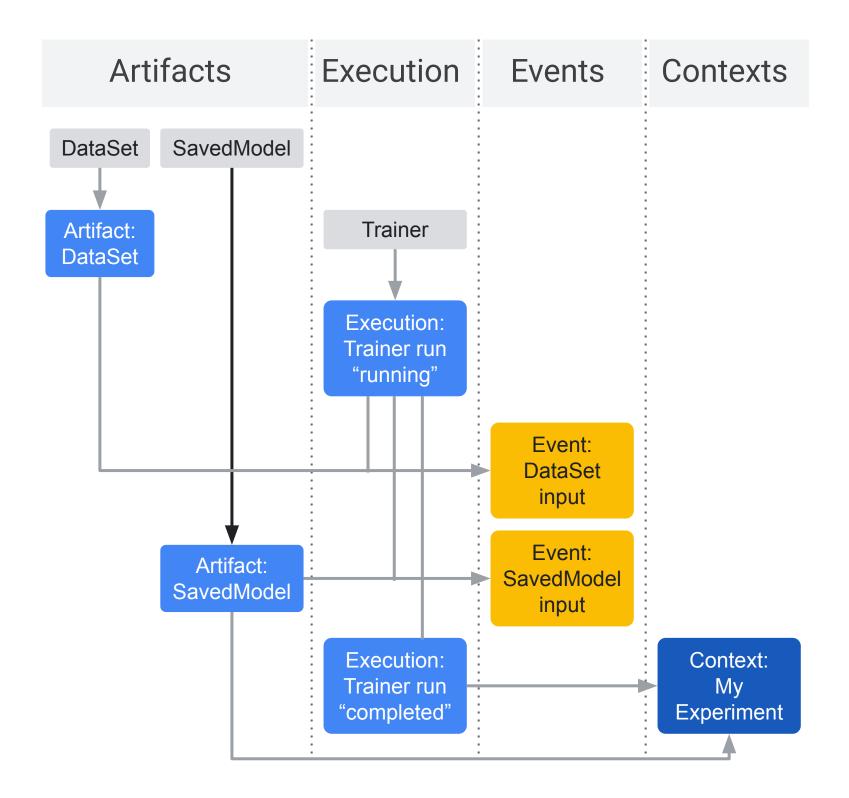


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- 5. Read DataSet and record input event.
- 6. Train Model and Create SavedModel Artifact.



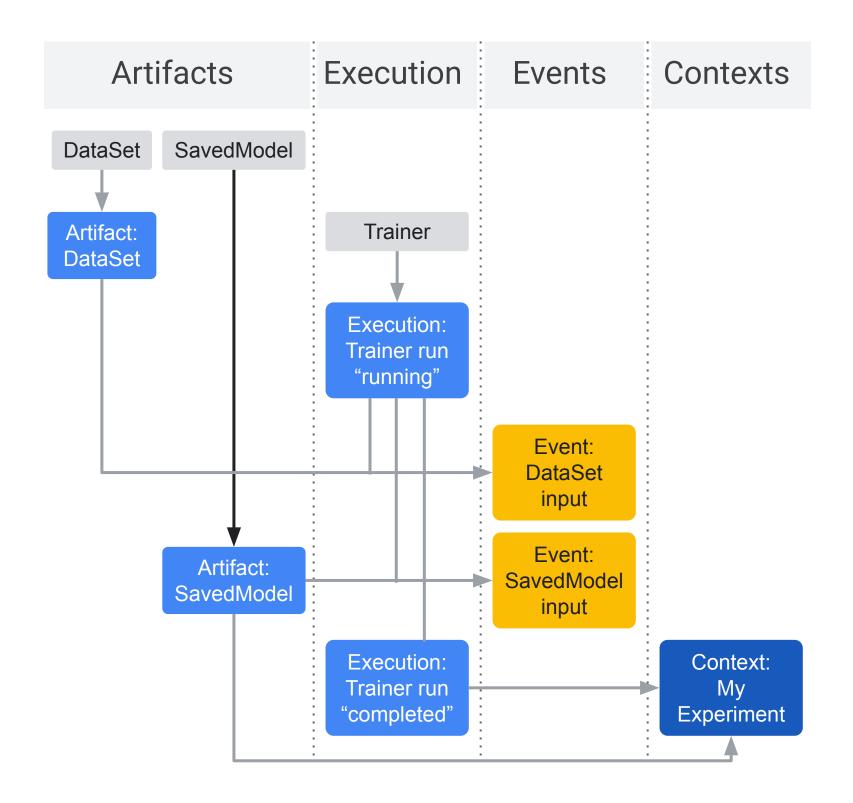


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- 6. Train Model and Create SavedModel Artifact.
- 7. Write SavedModel and record output event.



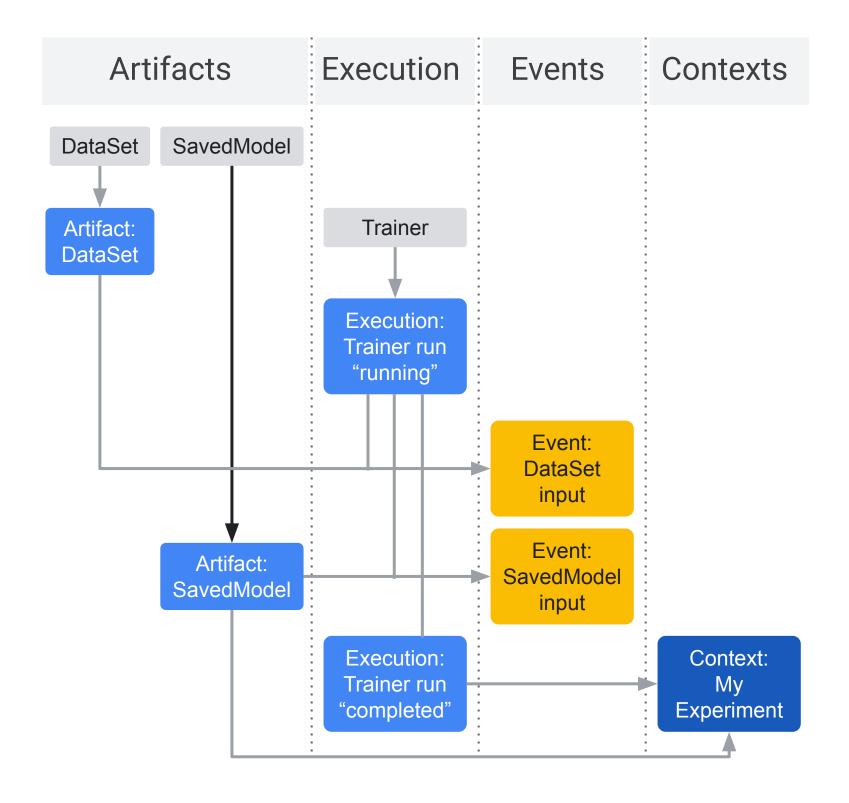


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- 8. Mark Execution completed.



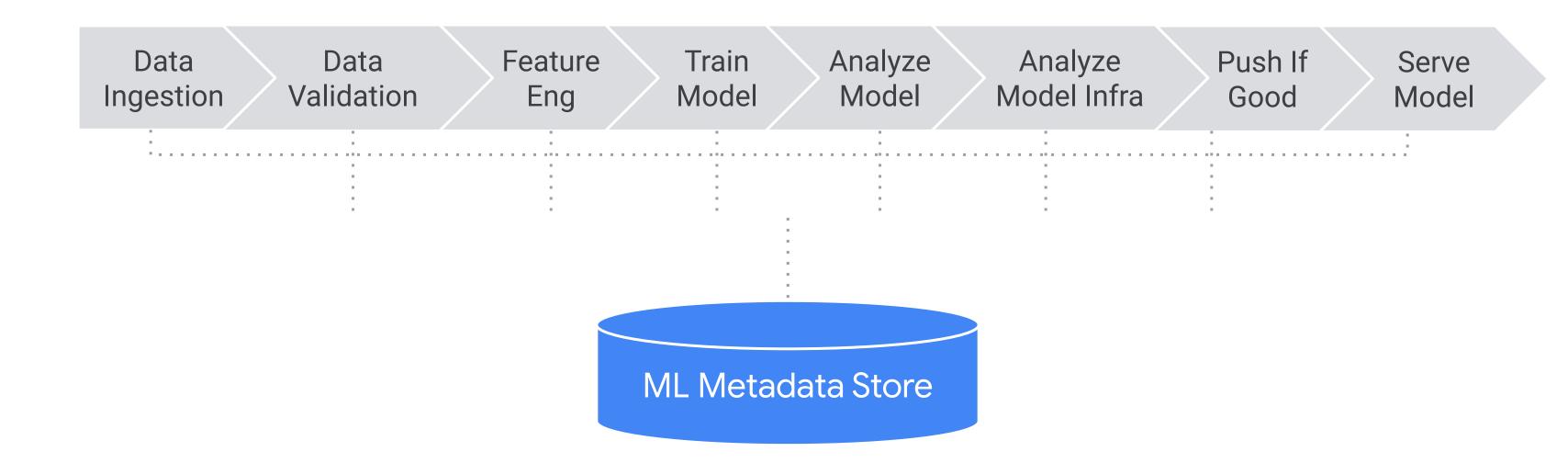


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- 8. Mark Execution completed.
- 9. Annotate the experiment with a Context.





### Metadata enables TFX task and data-aware pipelines





# Lab

TFX Pipeline Metadata



