

1. Static datasets are used for production ML modeling.

1 / 1 punto

☒ False

☐ True

✓ **Correcto**

That's it! Dynamic real-world data is used.

2. In production ML, the design priority is fast training.

1 / 1 punto

☒ No

☐ Yes

✓ **Correcto**

Correct! Fast training and choosing a high-performance algorithm are the design priorities for prototypes or research ML.

3. Developers adhere to modern software development to produce low-maintenance software, and to address project evolution. Select all the key aspects of modern software development (Check all that apply):

1 / 1 punto

☒ Testability

✓ **Correcto**

Yes! The data entering the system is continuously monitored and tested.

☒ Monitoring

✓ **Correcto**

Right on! The deployed model's performance is properly evaluated.

☒ Best practices

✓ **Correcto**

Perfect! Software development best practices must be resolved.

☐ Fast Training

4. Model-performance needs to be continuously monitored, and new data, ingested and re-trained.

1 / 1 punto

☒ Yes

☐ No



**Correcto**

Good job! After deployment, it's necessary to continuously evaluate the model's performance.

5. ML pipeline workflows are almost always DAGs.

1 / 1 punto

☒ True

☐ False



**Correcto**

Well done! The components of an ML pipeline are scheduled based on dependencies defined by a DAG.

6. TensorFlow Extended (TFX) is an end-to-end platform for deploying production ML pipelines.

1 / 1 punto

☐ No

☒ Yes



**Correcto**

You got it right! TFX is used to build and manage ML pipelines in production.

7. Production machine learning combines which two key disciplines?

1 / 1 punto

☐ Feature selection and engineering

☒ Modern software development

☒ **Correcto**

Keep it up! Well-designed software that adheres to best practices is key for the success of a production grade machine learning system.

☒ Machine learning development

☒ **Correcto**

Nice going! ML Development focuses on specific issues related with data and model predictions quality.

☐ Software testing

8. What are the unique challenges to overcome in a production-grade ML system? (Check all that apply)

1 / 1 punto

☒ Handling continuously changing data.

☒ **Correcto**

Indeed! Data will change over the life cycle of a production system, which can harm its performance.

☒ Optimizing computational resources and costs.

☒ **Correcto**

Absolutely! You want your ML system to be as frugal as possible.

☐ Deploying the model to serve requests.

☒ Building integrated ML systems.

☒ **Correcto**

Very well! ML systems perform all operations starting from ingesting the data into the system to deployment.

☒ Continually operating while in production.

☒ **Correcto**

Right on track! ML systems need to be flexible to operate while the system stages or modules are being changed or redesigned.

- ☐ Training the model on real world data.
- ☐ Assessing model performance.

9. **Production grade machine learning** challenges are addressed by implementing an important concept:

1 / 1 punto

- ☒ Machine learning pipelines
- ☐ Directed Acyclic Graphs (DAGs)
- ☐ Orchestrators
- ☐ Tensorflow Extended (TFX)

✓ **Correcto**

Spot on! ML pipelines provide support for automating, monitoring and maintaining a model as you continue to train it over its lifetime.

10. TensorFlow Lite is a deep learning framework to deploy TFX pipelines into:

1 / 1 punto

- ☒ Mobile devices
- ☐ Web browser
- ☐ Servers

✓ **Correcto**

That's it! Tensorflow Lite is the tool for deploying TFX pipeline into mobile and IoT devices.