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# **Analysis on results on Workshop:**

# **Logistic Regression with ADASYN:**

- Suitable for: Medium-Sized Banks.
- **Reason**: High recall (0.894) and ROC AUC score (0.934) make it highly effective for medium-scale operations needing reliable fraud detection.

## **Decision Tree with ADASYN:**

- Suitable for: Large Multinational Banks.
- **Reason**: Both a high recall (0.894) and ROC AUC score (0.934) provide strong fraud detection capabilities necessary for large volumes of transactions.

# Logistic Regression with Oversampling:

- Suitable for: Large Multinational Banks.
- **Reason**: Highest ROC AUC score (0.935) and high recall (0.886), crucial for minimizing false negatives in large-scale fraud detection.

# **Decision Tree with Oversampling:**

- Suitable for: Regional Banks.
- **Reason**: Combines good ROC AUC score (0.910) with enhanced recall (0.829) through oversampling, ideal for regional operations with moderate data volumes.

## **Decision Tree:**

- Suitable for: Local Small Banks.
- **Reason**: Balanced recall (0.821) and ROC AUC score (0.910) make it sufficient for smaller datasets and resource-limited settings.

# **Logistic Regression IMB:**

- Suitable for: Small local operations having conservative approach
- Reason: Even though recall (0.756) is relatively low, ROC AUC score (0.877) makes it
  effective for small operations preferring precision over recall to avoid inconvenience from
  AMEX automatic restrictions