

# RAPIDS

Accelerating ML  
using RAPIDS



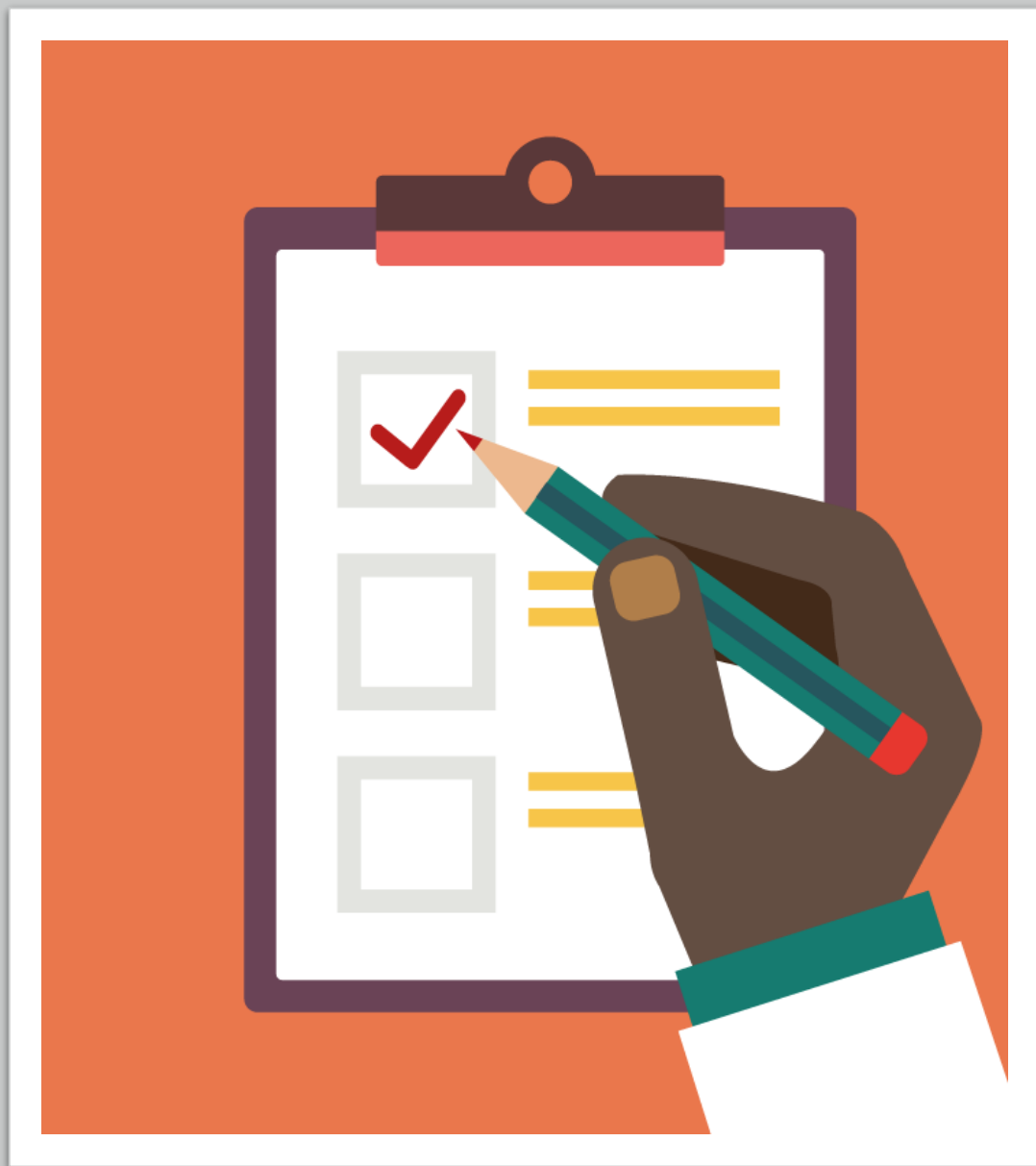
# Agenda

- **Why Rapids?**
- Why GPU?
- Libraries in RAPIDS
- ML Pipeline using RAPIDS
- Pandas vs cuDF
- Scikit-learn vs cuML
- Demo case study

# Why RAPIDS?

- **Rapids** is a suite of libraries designed for accelerating Data Science
- The beauty of Rapids is that it's integrated smoothly with Data Science libraries —things like Pandas data-frames are easily passed through to Rapids for GPU acceleration.



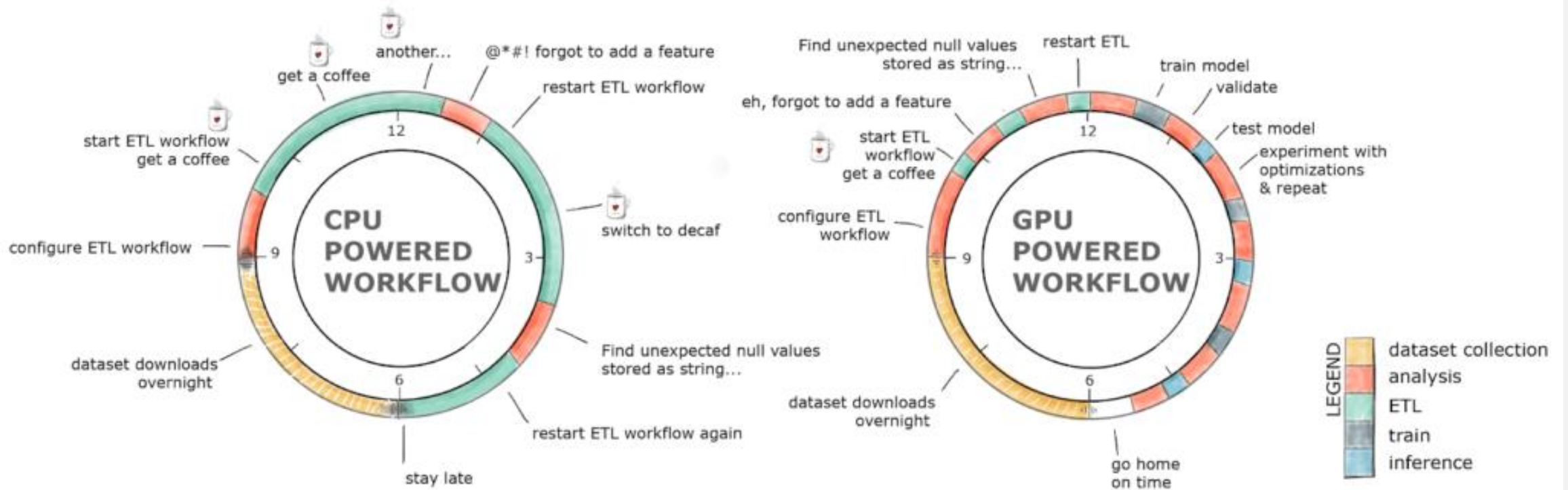


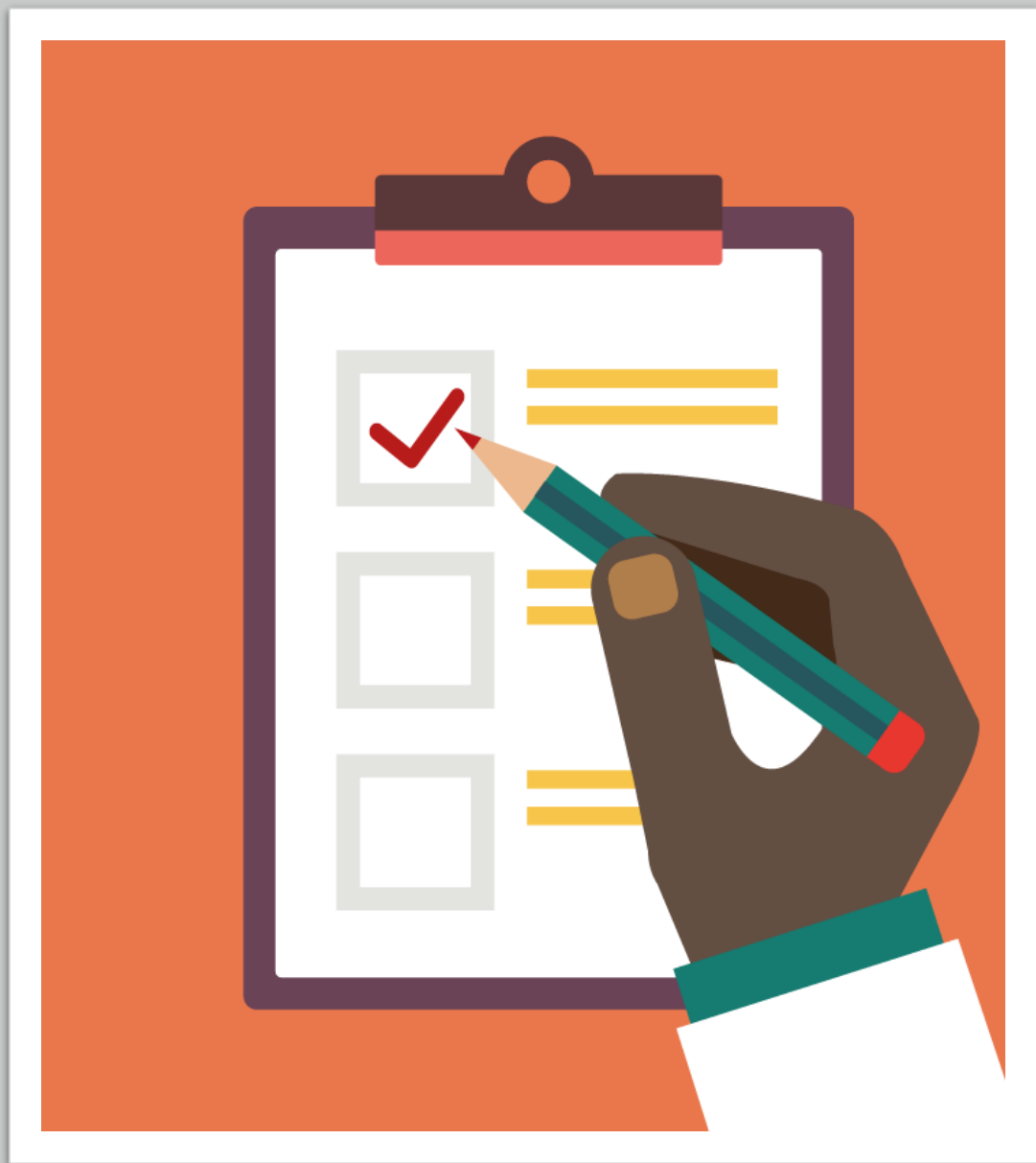
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# Why GPU?

## DAY IN THE LIFE OF A DATA SCIENTIST



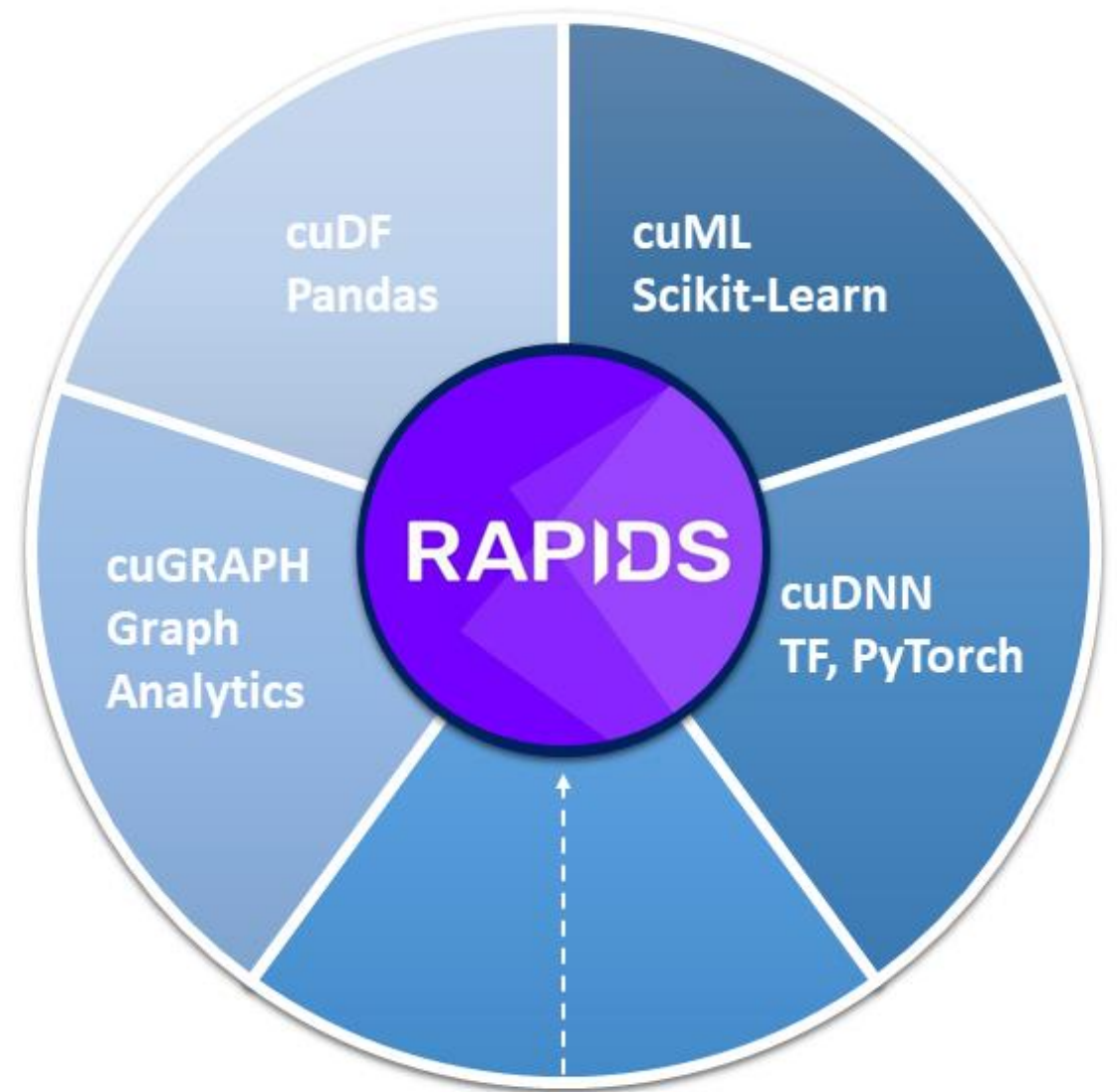


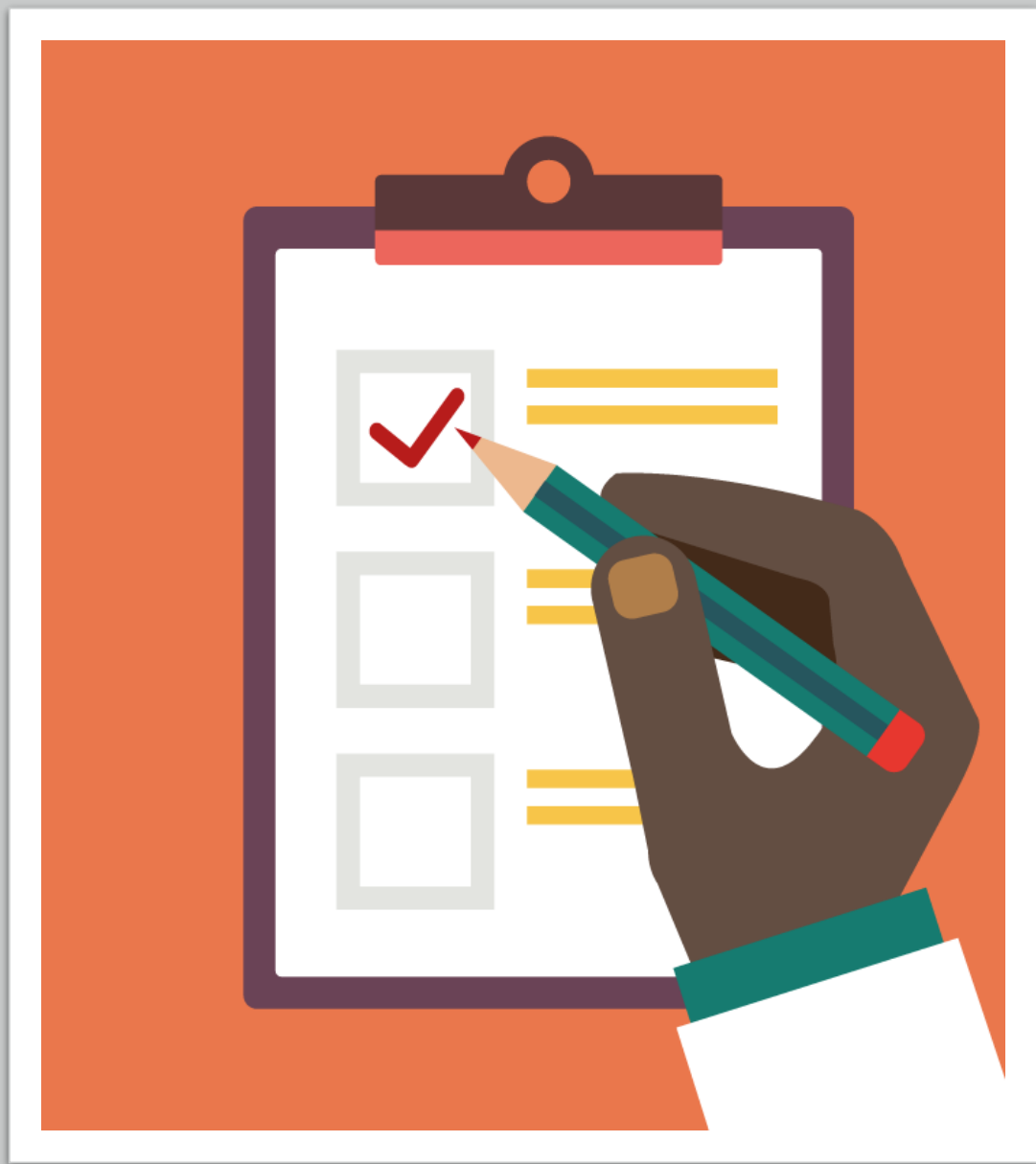
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# Libraries in RAPIDS

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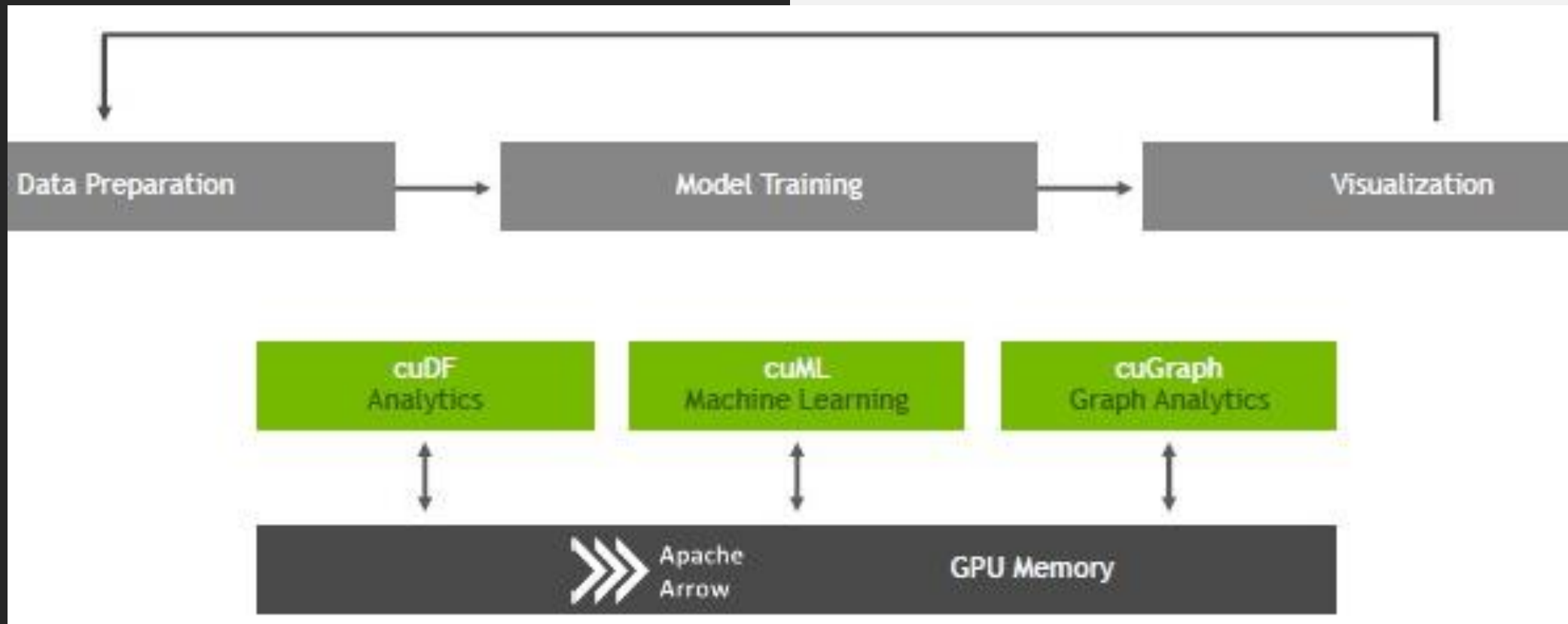


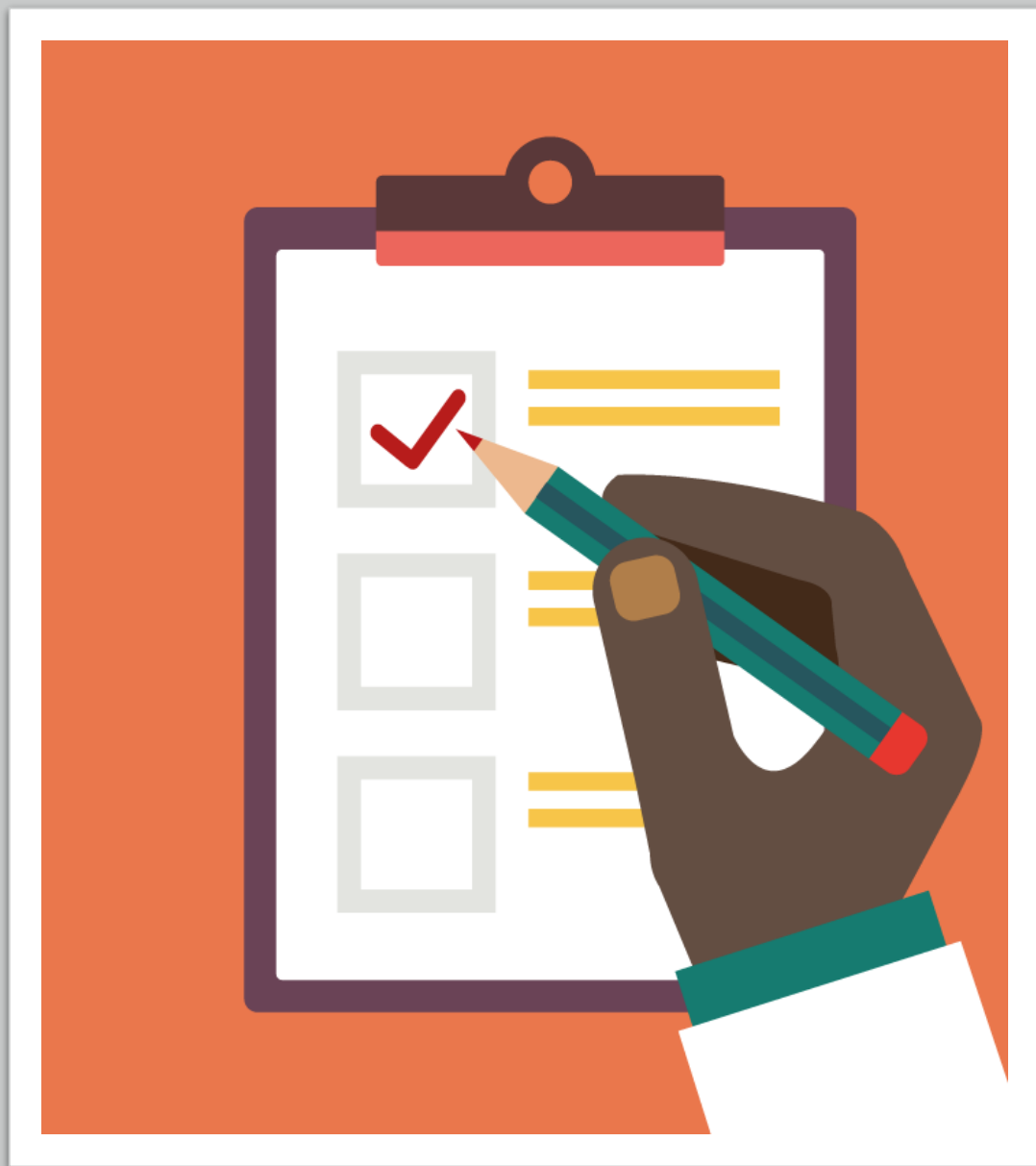
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# ML Pipeline using RAPIDS





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# Pandas vs cuDF

- cuDF always outperforms Pandas while importing a big dataset

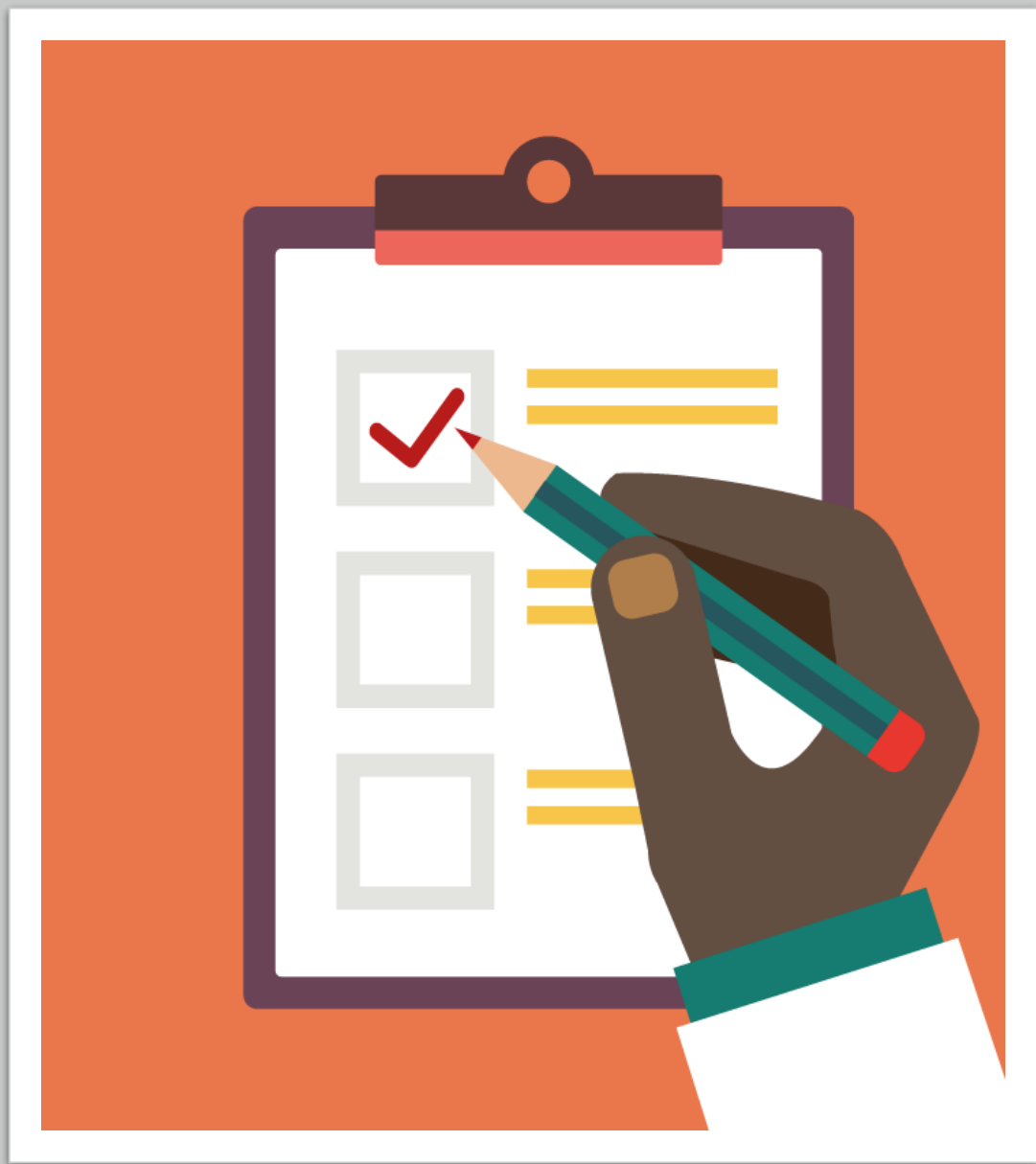
```
[ ] %%time
df=pd.read_csv('https://storage.googleapis.com/industryanalytics/LoanDefaultData.csv')
print(df.shape)
```

```
↳ (887379, 22)
CPU times: user 1.74 s, sys: 475 ms, total: 2.21 s
Wall time: 5.2 s
```

```
[ ] %%time
df1= cu.read_csv('https://storage.googleapis.com/industryanalytics/LoanDefaultData.csv')
print(df1.shape)
```

```
↳ (887379, 22)
CPU times: user 1.31 s, sys: 499 ms, total: 1.81 s
Wall time: 3.15 s
```





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# Scikit-learn vs. cuML

```
[ ] from cuml import RandomForestClassifier as curf
    from cuml import LogisticRegression as lgr
```

```
[ ] import time
    start_time = time.time()
    rfc_gpu = curf(n_estimators = 100, max_depth = 5)
    rfc_gpu.fit(X_train_gdf, y_train_gdf)
    print("GPU Training Time with GPU dataframe: %s seconds" % (str(time.time() - start_time)))
```

```
↳ /usr/local/lib/python3.6/dist-packages/ipykernel_launcher.py:4: UserWarning:
```

To use GPU-based prediction, first train using float 32 data to fit the estimator

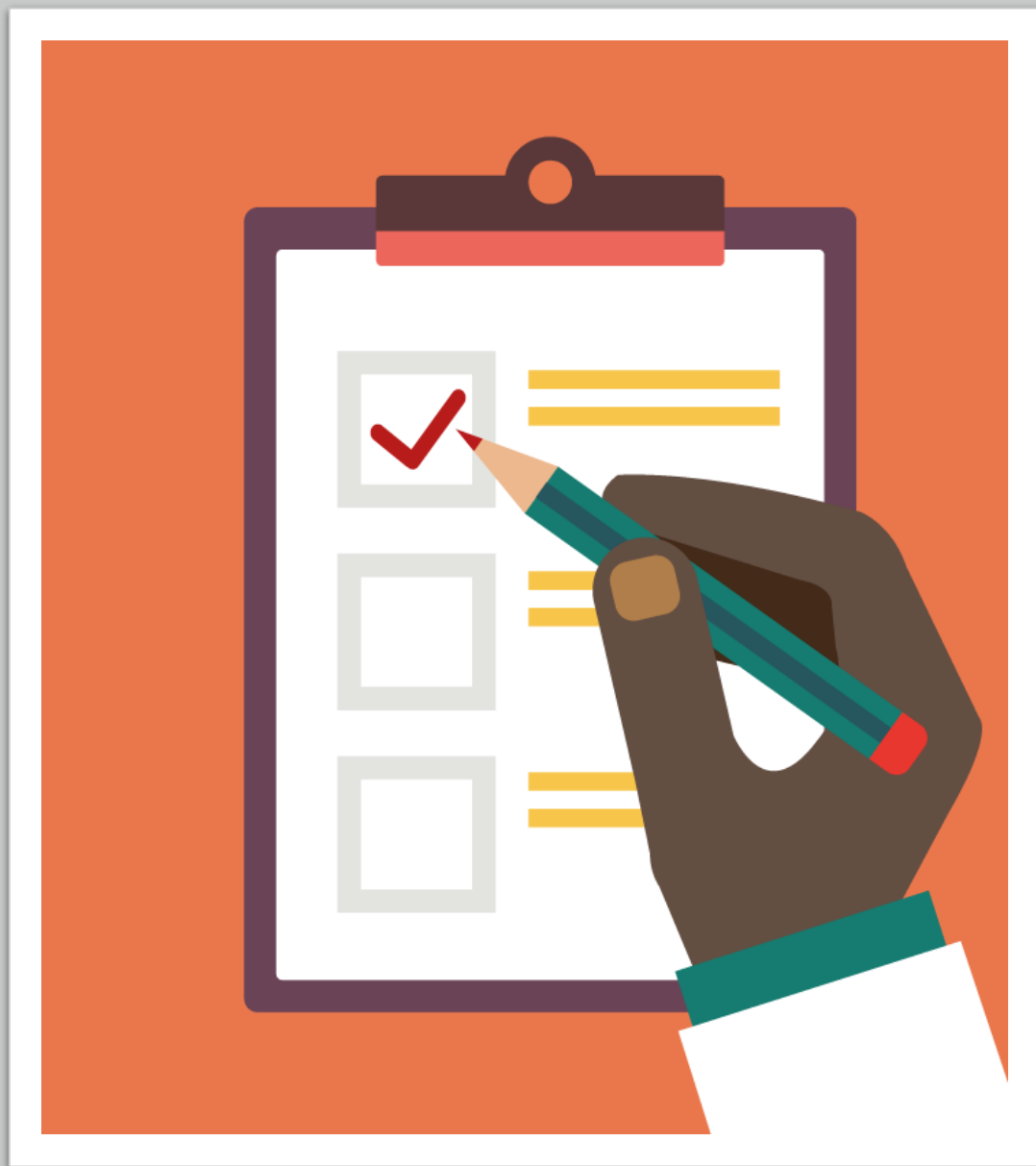
GPU Training Time with GPU dataframe: 0.23898983001708984 seconds

```
[17] from sklearn.ensemble import RandomForestClassifier
```

```
[16] import time
    start_time = time.time()
    rfc = RandomForestClassifier(n_estimators = 100, max_depth = 5)
    rfc.fit(X_train, y_train)
    print("Training Time with Pandas dataframe: %s seconds" % (str(time.time() - start_time)))
```

```
↳ Training Time with Pandas dataframe: 44.850953340530396 seconds
```

- Random-Forest algorithm from cuML outperforms the traditional Scikit-learn Random-Forest algorithm with a speed 44 times higher than the traditional Random forest algorithm



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**THANK  
YOU**

An illustration featuring two hands, each wearing a dark suit sleeve with white cufflinks, holding a rectangular orange banner. The banner is held taut between the hands, which are positioned at the bottom corners. The banner has the words "THANK" and "YOU" written in large, bold, white capital letters, stacked vertically. The background is a solid light blue. The entire illustration is framed by a white border, which is itself set against a yellow background.