

### PHASE 3 Latent Churn & Survival Modeling (Non-Contractual Setting)

# STEP 3.1 – Load Phase 2 Artifact (Immutable)

```
import pandas as pd
```

```
state_df = pd.read_parquet("phase2_customer_state.parquet")
```

```
print(state_df.shape)
```

```
state_df.head()
```

```
(37039, 7)
```

```
{
  "summary": {
    "name": "state_df",
    "rows": 37039,
    "fields": [
      {
        "column": "Customer ID",
        "properties": {
          "dtype": "number",
          "std": 1721.1264193434051,
          "min": 12346.0,
          "max": 18287.0,
          "num_unique_values": 5881,
          "samples": [
            17776.0, 17703.0, 12546.0
          ],
          "semantic_type": "",
          "description": ""
        },
        "column": "InvoiceDate",
        "properties": {
          "dtype": "date",
          "min": "2009-12-01 07:45:00",
          "max": "2011-12-09 12:50:00",
          "num_unique_values": 34591,
          "samples": [
            "2010-09-20 17:32:00", "2011-04-15 08:45:00", "2010-01-20 09:50:00"
          ],
          "semantic_type": "",
          "description": ""
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        "column": "recency_days",
        "properties": {
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          "std": 75.7097797248008,
          "min": 0.0,
          "max": 714.0,
          "num_unique_values": 515,
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            384.0, 682.0, 690.0
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          "semantic_type": "",
          "description": ""
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        "properties": {
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          "num_unique_values": 400,
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          ],
          "semantic_type": "",
          "description": ""
        },
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        "properties": {
          "dtype": "number",
          "std": 777.8735726422763,
          "min": 0.0,
          "max": 84236.25,
          "num_unique_values": 35842,
          "samples": [
            261.8607692307692, 443.9271428571429, 339.4130612244898
          ],
          "semantic_type": "",
          "description": ""
        },
        "column": "delta_revenue",
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          "std": 1635.9432289056133,
          "min": -42409.1,
          "max": 168466.7,
          "num_unique_values": 28759,
          "samples": [
            159.18, 80.27999999999997, 516.1199999999999
          ],
          "semantic_type": ""
        }
      ]
    }
  }
}
```

```

\ "description\": \ "\ "\n      }\n    },\n    {\n      \ "column\":
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618.0,\n      \ "max\": 701.0,\n      \ "num_unique_values\": 796,\n
\ "samples\": [\n      387.0,\n      492.0,\n      -166.0\n
],\n      \ "semantic_type\": \ "\",\n      \ "description\": \ "\ "\n
}\n    }\n  ]\n}\n", "type": "dataframe", "variable_name": "state_df"}

```

*# STEP 3.2 – Define “Observation End” (Critical Concept)*

```

END_DATE = state_df["InvoiceDate"].max()
END_DATE

```

```

Timestamp('2011-12-09 12:50:00')

```

*# STEP 3.3 – Define Inactivity Threshold (Censoring Rule)*

```

INACTIVITY_THRESHOLD_DAYS = 180 # 6 months

```

*# STEP 3.4 – Compute “Time Since Last Purchase”*

```

last_purchase = (
    state_df.groupby("Customer ID")["InvoiceDate"]
        .max()
        .reset_index()
        .rename(columns={"InvoiceDate": "last_invoice_date"})
)

```

```

last_purchase["days_since_last_purchase"] = (
    END_DATE - last_purchase["last_invoice_date"]
).dt.days

```

```

last_purchase.head()

```

```

{"summary": "{\n  \ "name\": \ "last_purchase\","\n  \ "rows\": 5881,\n
\ "fields\": [\n    {\n      \ "column\": \ "Customer ID\","\n
\ "properties\": {\n      \ "dtype\": \ "number\","\n      \ "std\":
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[\n      17776.0,\n      17703.0,\n      12546.0\n
n      ],\n      \ "semantic_type\": \ "\",\n
\ "description\": \ "\ "\n      }\n    },\n    {\n      \ "column\":
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\ "date\","\n      \ "min\": \ "2009-12-01 09:55:00\","\n      \ "max\":
\ "2011-12-09 12:50:00\","\n      \ "num_unique_values\": 5731,\n
\ "samples\": [\n      \ "2011-11-20 10:15:00\","\n      \ "2010-
11-04 09:06:00\","\n      \ "2011-02-04 14:03:00\","\n      ],\n
\ "semantic_type\": \ "\",\n      \ "description\": \ "\ "\n      }\n
n    },\n    {\n      \ "column\": \ "days_since_last_purchase\","\n
\ "properties\": {\n      \ "dtype\": \ "number\","\n      \ "std\":
209,\n      \ "min\": 0,\n      \ "max\": 738,\n

```

```

\ "num_unique_values\ ": 676,\n          \ "samples\ ": [\n          586,\n
492,\n          460\n          ],\n          \ "semantic_type\ ": \ "\",\n
\ "description\ ": \ "\",\n          }\n          }\n          ]\n
n\ },\n          \ "type\ ": "dataframe",\n          \ "variable_name\ ": "last_purchase"}

# STEP 3.5 – Define Survival Status (Alive vs Censored)

last_purchase["is_alive"] = (
    last_purchase["days_since_last_purchase"]
    <= INACTIVITY_THRESHOLD_DAYS
)

last_purchase["is_alive"].value_counts(normalize=True)

is_alive
True      0.591906
False     0.408094
Name: proportion, dtype: float64

# STEP 3.6 – Create Survival Table (Customer-Level)

survival_df["event"] = (
    survival_df["days_since_last_purchase"]
    > INACTIVITY_THRESHOLD_DAYS
).astype(int)

# STEP 3.7 – Merge Survival Target Back to State (Time-Indexed)

state_survival_df = state_df.merge(
    survival_df[["Customer ID", "is_alive", "duration", "event"]],
    on="Customer ID",
    how="left"
)

state_survival_df.head()

{"summary": "{\n  \ "name\ ": \ "state_survival_df\ ",\n  \ "rows\ ": 37039,\n
n  \ "fields\ ": [\n      {\n          \ "column\ ": \ "Customer ID\ ",\n
\ "properties\ ": {\n          \ "dtype\ ": \ "number\ ",\n          \ "std\ ":
1721.1264193434051,\n          \ "min\ ": 12346.0,\n          \ "max\ ":
18287.0,\n          \ "num_unique_values\ ": 5881,\n          \ "samples\ ":
[\n          17776.0,\n          17703.0,\n          12546.0\
n          ],\n          \ "semantic_type\ ": \ "\",\n
\ "description\ ": \ "\",\n          }\n          },\n          {\n          \ "column\ ":
\ "InvoiceDate\ ",\n          \ "properties\ ": {\n          \ "dtype\ ":
\ "date\ ",\n          \ "min\ ": \ "2009-12-01 07:45:00\ ",\n          \ "max\ ":
\ "2011-12-09 12:50:00\ ",\n          \ "num_unique_values\ ": 34591,\n
\ "samples\ ": [\n          \ "2010-09-20 17:32:00\ ",\n          \ "2011-
04-15 08:45:00\ ",\n          \ "2010-01-20 09:50:00\ "\n          ],\n
\ "semantic_type\ ": \ "\",\n          \ "description\ ": \ "\",\n          }\n
n      },\n          {\n          \ "column\ ": \ "recency_days\ ",\n

```

```

\"properties\": {\n          \"dtype\": \"number\", \n          \"std\": 75.7097797248008, \n          \"min\": 0.0, \n          \"max\": 714.0, \n          \"num_unique_values\": 515, \n          \"samples\": [\n            384.0, \n            682.0, \n            690.0\n          ], \n          \"semantic_type\": \"\", \n          \"description\": \"\", \n          \"column\": \"frequency\", \n          \"properties\": {\n            \"dtype\": \"number\", \n            \"std\": 37, \n            \"min\": 0, \n            \"max\": 399, \n            \"num_unique_values\": 400, \n            \"samples\": [\n              209, \n              280, \n              33\n            ], \n            \"semantic_type\": \"\", \n            \"description\": \"\", \n            \"column\": \"monetary_avg\", \n            \"properties\": {\n              \"dtype\": \"number\", \n              \"std\": 777.8735726422763, \n              \"min\": 0.0, \n              \"max\": 84236.25, \n              \"num_unique_values\": 35842, \n              \"samples\": [\n                261.8607692307692, \n                443.9271428571429, \n                339.4130612244898\n              ], \n              \"semantic_type\": \"\", \n              \"description\": \"\", \n              \"column\": \"delta_revenue\", \n              \"properties\": {\n                \"dtype\": \"number\", \n                \"std\": 1635.9432289056133, \n                \"min\": -42409.1, \n                \"max\": 168466.7, \n                \"num_unique_values\": 28759, \n                \"samples\": [\n                  159.18, \n                  80.27999999999997, \n                  -516.1199999999999\n                ], \n                \"semantic_type\": \"\", \n                \"description\": \"\", \n                \"column\": \"delta_recency\", \n                \"properties\": {\n                  \"dtype\": \"number\", \n                  \"std\": 75.6244236542911, \n                  \"min\": -618.0, \n                  \"max\": 701.0, \n                  \"num_unique_values\": 796, \n                  \"samples\": [\n                    387.0, \n                    492.0, \n                    -166.0\n                  ], \n                  \"semantic_type\": \"\", \n                  \"description\": \"\", \n                  \"column\": \"is_alive\", \n                  \"properties\": {\n                    \"dtype\": \"boolean\", \n                    \"num_unique_values\": 2, \n                    \"samples\": [\n                      true, \n                      false\n                    ], \n                    \"semantic_type\": \"\", \n                    \"description\": \"\", \n                    \"column\": \"duration\", \n                    \"properties\": {\n                      \"dtype\": \"number\", \n                      \"std\": 145, \n                      \"min\": 0, \n                      \"max\": 738, \n                      \"num_unique_values\": 676, \n                      \"samples\": [\n                        586, \n                        492\n                      ], \n                      \"semantic_type\": \"\", \n                      \"description\": \"\", \n                      \"column\": \"event\", \n                      \"properties\": {\n                        \"dtype\": \"number\", \n                        \"std\": 0, \n                        \"min\": 0, \n                        \"max\": 1, \n                        \"num_unique_values\": 2, \n                        \"samples\": [\n                          0, \n                          1\n                        ], \n                        \"semantic_type\": \"\", \n                        \"description\": \"\", \n                        \"column\": \"\" \n                      ] \n                    } \n                  } \n                } \n              } \n            } \n          } \n        ], \n        \"type\": \"dataframe\", \"variable_name\": \"state_survival_df\" \n      } \n    } \n  } \n}

```

# STEP 3.8 – Sanity Checks (Must Pass)

```
# No missing survival labels
state_survival_df[["is_alive", "duration", "event"]].isna().sum()
```

```
is_alive    0
duration    0
event       0
dtype: int64
```

```
# Reasonable alive ratio
state_survival_df["is_alive"].mean()
```

```
np.float64(0.8323118874699641)
```

```
# STEP 3.9 – Save Phase 3 Artifact
```

```
state_survival_df.to_parquet(
    "phase3_state_with_survival.parquet",
    index=False
)
```