

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

```

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},
"type": "dataframe",
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}

# 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": {
        "name": "last_purchase",
        "rows": 5881,
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                "column": "Customer ID",
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                        "2011-11-20 10:15:00",
                        "2011-02-04 14:03:00"
                    ],
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                "column": "days_since_last_purchase",
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                    "dtype": "number",
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                    "min": 0,
                    "max": 738
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            }
        ]
    }
}

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    "num_unique_values": 676,\n        "samples": [586,\n492,\n        460],\n        "semantic_type": "\",\n    "description": "\"\\n        }\\n    }\\n ]\\n}\",\n    \"type\": \"dataframe\", \"variable_name\": \"last_purchase\"}\n\n# STEP 3.5 – Define Survival Status (Alive vs Censored)\n\nlast_purchase["is_alive"] = (\n    last_purchase["days_since_last_purchase"]\n    <= INACTIVITY_THRESHOLD_DAYS\n)\n\nlast_purchase["is_alive"].value_counts(normalize=True)\n\nis_alive\nTrue      0.591906\nFalse     0.408094\nName: proportion, dtype: float64\n\n# STEP 3.6 – Create Survival Table (Customer-Level)\n\nsurvival_df["event"] = (\n    survival_df["days_since_last_purchase"]\n    > INACTIVITY_THRESHOLD_DAYS\n).astype(int)\n\n# STEP 3.7 – Merge Survival Target Back to State (Time-Indexed)\n\nstate_survival_df = state_df.merge(\n    survival_df[["Customer ID", "is_alive", "duration", "event"]],\n    on="Customer ID",\n    how="left"
)\n\nstate_survival_df.head()\n\n{"summary": {"name": "state_survival_df",\n    "rows": 37039,\n    "fields": [{"column": "Customer ID",\n        "properties": {"dtype": "number",\n            "std": 1721.1264193434051,\n            "min": 12346.0,\n            "max": 18287.0,\n            "num_unique_values": 5881,\n            "samples": [17776.0, 17703.0, 12546.0],\n            "semantic_type": "\",\n            "description": "\"\\n        }\\n    }\\n ]\\n"},\n        "column": "InvoiceDate",\n        "properties": {"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": ["2010-09-20 17:32:00",\n                "2010-01-20 09:50:00"],\n            "semantic_type": "\",\n            "description": "\"\\n        }\\n    }\\n ]\\n"},\n        "column": "recency_days",\n        "properties": {}}],\n    "semantic_type": "\",\n    "description": "\"\\n        }\\n    }\\n ]\\n"}\n

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        }
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    "type": "dataframe",
    "variable_name": "state_survival_df"
}

# 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
)
```