

## Multi-Event Log in Celonis Process Mining 4.7

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### Motivation for Multi-Event Log

Questions you can answer with Process Mining:

- How much time do you need to sell a product after it was ordered from you?
- If you realise you don't have this product in stock but have to buy it from another vendor, how long does it take until it arrives in your warehouse?
- Are you paying the vendor according to the best possible payment terms?

In the Process Mining world so far, these are three different perspectives on the data that would require three separate Data Models (see below), three different Analyses and a lot of mental overhead to analyze those. Discover in the following how you can combine the three perspectives into one with the new Multi-Eventlog capability.

### Limitations of traditional Process Mining

Traditional Process Mining follows one record type through the system. For this record type, all events related to the record are collected in an activity table and linked together with a case ID.

In general an entry in an Event Log consists of:

- **Case ID**: identifies the record that we follow through the system
- Activity: describes what happens to the record
- **Timestamp**: describes at which point in time something happened to the record

That means each Event Log is **specific to a certain record type**, e.g. there can be one Event Log for everything related to a Purchase Order Items and another Event Log for Sales Order Items.

The Event Log - or rather the record type we chose as Case ID - **defines the perspective and section of data** for the visualization of the as-is process. A Data Model containing only a single Event Log limits the analysis to this one perspective. With such limitations, we can only analyze one process at a time with a certain focus.

For record types in purchase, sales and production the following limitations can be observed when using single-perspective Data Models:



#### **Production Order Items Current Limitation Purchase Orders Items Sales Orders Items** visualization of all all production different data sales order related Purchase purchase and orders models with production orders Requisitions and no insight in different focus placed sales Invoices related to can be included point a Purchase Order no visibility in orders for no cross-filtering no visibility in buy-/make-toadvanced hard to identify "stuck" purchase make-to-stock stock requisitions and bottlenecks no connection planning

no analysis of invoices w/o PO between a customers placed orders and claims

limited to no insight in parallel and sequential production flows

interdependencies across processes and business areas

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Figure 1: Single-perspective Data Models allow for in-depth analysis of single processes but lack inter-process insights

### Opportunities with Multi-Event Log Process Mining

Modern organizations of course run a variety of different processes in order to achieve their desired business outcomes. Those processes do not stand alone but interfere with each other. Possible inefficiencies and their root causes are often correlated among different processes. As an example, an inefficiency in the order process (O2C) may be partially or fully caused by an inefficiency happening in the purchasing process (P2P) inside the same organization.

So, while analyzing one process helps to find inefficiencies within that particular process, it does not provide insights into how inefficiencies correlate with events in related processes. Analyzing multiple process perspectives at a time bears the opportunity to capture such dependencies and remove operational friction more holistically.



### Multi-Event Log in a nutshell

The Multi-Event Log technology provides the capability to analyze end-to-end business processes and their interdependencies within one analysis. It allows to connect directly linked processes and to bring indirectly linked processes into context.

The Multi-Event Log technology allows you to include multiple Event Logs with their respective case tables in one semantic Data Model. All Event Logs can be linked to each other as well as to master data via joins/foreign key relationships. A holistic analysis can then access all this data on top.

This capability adds value for various use cases. In general there are four high level use cases where Multi-Event Log can be used to simplify or extend the analytical capabilities:

- Use Case 1: Put independent processes into context (filter across processes)
- Use Case 2: Analyze parallel processes by linking multiple hierarchical Event Logs
- Use Case 3: Simplified scripts (no-join transformations) by merging Event Logs
- Use Case 4: Visualize end-to-end processes by linking multiple Event Logs\*

By using existing components and process comprehensive filtering you can then analyze how KPIs behave across multiple processes. This enables you to identify inefficiencies and their sources that can be spread over several processes.

\* This functionality is only available on the Execution Management System Platform (Cloud deployment) with the new visualisation tech-stack of "Views".



### Set up of a Multi-Event Log Data Model

### Data Transformation

Writing transformation scripts with Multi-Event Log enabled data models simplifies the transformations as well as can speed up the data transformation (executing the script) significantly.

Instead of having one case key and mapping all activities to this perspective, Multi-Event Log allows to keep different perspectives (that means case definitions) for different processes. For each process, users can simply create a separate activity table with the corresponding case key of that process. This takes away the burden to create complicated joins between source tables for each activity definition. Instead, the link between processes is shifted from the transformation to the Data Model. Their relationship is simply modelled via join key relationships in the visual Data Model Editor.

Let's look at an example, the definition of the 'Clear Invoice' activity in the SAP Purchase-To-Pay process. Usually, this activity is defined in the Purchase-To-Pay process with the following SQL statement:

```
SELECT DISTINCT
E._CASE_KEY AS "_CASE_KEY"
,E.MANDT AS "MANDT"
,E.EBELN AS "EBELN"
,E.EBELP AS "EBELP"
,'Gleiche Rechnung aus' AS "ACTIVITY_DE"
,'Clear Invoice' AS "ACTIVITY_EN"
,CASE
    WHEN CAST(BKPF_Z."CPUDT" AS DATE) = CAST(B."AUGDT" AS DATE)
    THEN CAST(BKPF_Z."CPUDT" AS DATE) + CAST(BKPF_Z."CPUTM" as TIME)
    ELSE CAST(B."AUGDT" AS DATE) + CAST(BKPF_Z."CPUTM" as TIME)
    ELSE CAST(B."AUGDT" AS DATE) + CAST('23:59:59' as TIME)
END AS "EVENTTIME"
,2500 AS "_SORTING"
,BKPF_Z.USNAM AS "USER_NAME"
,USR02.USTYP AS "USER_TYPE"
```



```
, BKPF Z.TCODE AS "TRANSACTION CODE"
FROM
   RSEG AS RSEG
   INNER JOIN TMP P2P EKKO EKPO AS E ON 1=1
         AND RSEG.MANDT = E.MANDT
        AND RSEG.EBELN = E.EBELN
         AND RSEG.EBELP = E.EBELP
   INNER JOIN TMP P2P BKPF BSEG AS B ON 1=1
        AND B.MANDT = RSEG.MANDT
       AND SUBSTRING (B.AWKEY, 1, 14) = RSEG.BELNR || CAST (RSEG.GJAHR AS
VARCHAR(4))
   LEFT JOIN BKPF AS BKPF Z ON 1=1
       AND B.MANDT = BKPF Z.MANDT
       AND B.BUKRS = BKPF Z.BUKRS
       AND B.AUGBL = BKPF Z.BELNR
       AND B.AUGGJ = BKPF Z.GJAHR
   LEFT JOIN USR02 AS USR02 ON 1=1
       AND BKPF Z.MANDT = USR02.MANDT
       AND BKPF Z.USNAM = USR02.BNAME
WHERE
   B.AUGBL IS NOT NULL AND
   BKPF Z.CPUDT IS NOT NULL;
```

The details like the 'User Name' for the 'Clear Invoice' activity are not stored in the Purchase Order table (EKPO). They are stored in the Accounting Document Header table (BKPF). To add this information, it is necessary to create this lengthy join starting from EKKO/EKPO to RSEG to BKPF/BSEG to the clearing document in BKPF (declared as BKPF\_Z).

With the Multi-Event Log approach, users can save lines of code and associated headache. The clearing activity can be written together with all other invoice related activities into a separate activity table with the case key of the corresponding invoice document. This way, the table join is much shorter. The link to a purchase document is not necessary at this point. The SQL statement can be reduced to the following expression:



```
SELECT DISTINCT
  B." CASE KEY" AS " CASE KEY"
   , 'Clear Invoice' AS "ACTIVITY EN"
   , CASE
       WHEN CAST (BKPF Z. "CPUDT" AS DATE) = CAST (B. "AUGDT" AS DATE)
       THEN CAST (BKPF Z."CPUDT" AS DATE) + CAST (BKPF Z."CPUTM" as TIME)
       ELSE CAST(B. "AUGDT" AS DATE) + CAST('23:59:59' as TIME)
  END AS "EVENTTIME"
   , BKPF Z. "USNAM" AS "USER NAME"
    ,USR02."USTYP" AS "USER TYPE"
    ,900 AS " SORTING"
    ,B."MANDT" AS "MANDT"
    ,B."BUKRS" AS "BUKRS"
    ,B."BELNR" AS "BELNR"
    ,B."GJAHR" AS "GJAHR"
    ,B."BUZEI" AS "BUZEI"
   , BKPF Z. "TCODE" AS "TRANSACTION CODE"
FROM
   "TMP AP BKPF BSEG" AS B
   LEFT JOIN "BKPF" AS BKPF Z ON 1=1
       AND B. "MANDT" = BKPF Z. "MANDT"
       AND B. "BUKRS" = BKPF Z. "BUKRS"
       AND B. "AUGBL" = BKPF Z. "BELNR"
        AND B. "AUGGJ" = BKPF Z. "GJAHR"
    LEFT JOIN "USR02" AS USR02 ON 1=1
       AND BKPF Z. "MANDT" = USR02. "MANDT"
        AND BKPF Z."USNAM" = USR02."BNAME"
WHERE
  B.AUGBL IS NOT NULL AND
  BKPF Z.CPUDT IS NOT NULL;
```

The example above contains two instead of the previous four table joins. But writing scripts this way, not only simplifies the necessary joins. Using a separate activity table for invoice documents allows users to additionally include invoices in the Data Model that do not have a link to a purchase document. For the SAP example, invoices in the Data Model are not restricted



to MM related invoices (link to Purchase Order). FI invoices are also part of the Data Model which would otherwise only be visible in the Accounts Payable process.



### Data Model

The setup of a Multi-Event Log Data Model is similar to the setup of a Data Model with one Event Log. First, all tables of a Data Model need to be added. Afterwards, the foreign key relationships between tables need to be defined. Now, every table that serves as an Event Log can be configured as an Activity table by choosing the role "Activities table" shown in the following screenshot.

	General Columns Process
General	
Technical name	
_MEL_EKPO_ACTIVITIES	
Short Name	
_MEL_EKPO_ACTIVITIES	
Description	
Description	
DB Connection	
Choose connection	
Source Tabelle	
Choose table	
Table Loader Pool	
Table Loader Pool	
Table is part of a partial load	
Hide table in analysis	
Table role	
No Role @	
Activities table	
Activities table	

After choosing the option, the process configuration has to be performed and Case ID, Activity Name and Timestamp columns can be defined as usual. Additionally, the corresponding case table has to be set in this configuration step (see figure 3). One Activity Table can be defined as the default Activity Table.

Afterwards, the configured activity table is marked with an **A** in the Data Model. The default Activity Table is marked as **DA**. Case tables are indicated with a **C**.



MEL_EKPO_ACTIVITIES Options		×
	General Columns Process	
Process configuration		
Case ID column 🔞		
_CASE_KEY		~
Activity column 🔞		
ACTIVITY_EN		~
Timestamp column 🔞		
EVENTTIME		~
End timestamp column 🔞		
		~
Related Case table 🔞		
EKPO		~
Default Activity table		
gure 3: Configuration of Activity	y table - assigning the Case table	



### Set up of a Multi-Event Log Analysis

### Discover multiple processes

Once the Data Model is loaded, the analysis can be set up to discover all processes (Event Logs) in context with each other (Use case 1).

### Configure Process Visualization in Process Explorer and Variant Explorer

Within the analysis, all process related components like the Process Explorer, Variant Explorer or Case Explorer refer to the Activity table marked as the Default Activity table in the Data Model (see previous chapter). However, the Variant Explorer as well as the Process Explorer allow to dynamically choose the Event Log that is visualized by selecting in the settings a "Custom dimension". This dimension refers by default to the specified activity column of the default activity table. By ticking the "Custom dimension" option it can be changed in the PQL Editor to any column of any Activity table in the Data Model.

- Zoom + 🔅	Activities List view	General Options	\$
m	65.1% of activities Reset Less - More +	PROCESS NODES Custom dimension Custom di	f(x)

Figure 4: Configuring the "Custom dimension" of Process/Variant Explorer to display different Event Log than from the default activity table



In the below example, the two Process Explorers have custom dimensions, one referencing the activity column of the Order-To-Cash activity table and one referencing the activity column of the Purchase-To-Pay activity table.

"_CEL_02C_ACTIVITIES" . "ACTIVITY_EN" Order To Cash Process	"_MEL_EKPO_ACTIVITIES"."ACTIVITY_EN" Purchase To Pay Process
<ul> <li>Create Sales Order Item</li> <li>Zoom +</li> <li>1,78,997</li> <li>Create Delivery</li> <li>1,767,538</li> <li>Record Goods Issue</li> <li>1,768,725</li> <li>1,768,725</li> <li>1,769,950</li> <li>1,769,950</li> <li>1,765,926</li> <li>93,1% - + <sup>T</sup>nd 79,5% - +</li> </ul>	604.928 - Zoom + Create Purchase Order Item 128.225 411.143 Change Price PO Item 106.790 Record Goods Receipt 106.790 Record Invoice 413.766 Process End 732.855 93.7%
Figure 5: Two processes in Process Explorers in the used to select the Order-To-Cash and the Purchase-Twas used to distinguish the two processes visually.	

Furthermore, users can now even combine two Event Logs into one Event Log with the PQL Function MERGE\_EVENTLOG. Users that split for example the Purchase Requisition activities from the Purchase Order activities into two Event Logs, to save time writing lengthy transformations per activity, can now combine the two activities into one Event Log by merging them together. For further details on this function, please visit the PQL Documentation on https://help.celonis.de/



### Create KPI Calculations across processes

Of course it is also possible to

- create KPIs for different processes within one analysis,
- filter across multiple processes, and
- create KPIs across multiple processes.

#### Metrics from different Event Logs in one Analysis

Here is an example of metrics of two processes in the same analysis:

	sh			b Dav	
Activity	Eventtime	Case Key	Activity	Eventtime	Case Key
Create Sales Ord	2017-12-31 05:20	110000153837	Create Purchase	2019-05-10 15:03	110430004980
Create Delivery	2018-01-11 05:51	110000153837	Change Price PO	2019-05-13 15:03	110430004980
Record Goods Iss	2018-01-14 11:24	110000153837	Record Goods Re		110430004980
Create Invoice	2018-01-17 11:24	110000153837	Record Invoice	2019-05-17 09:15	110430004980
Clear Invoice	2018-03-18 11:24	110000153837	Create Purchase	2019-05-10 14:54	110430004980
Create Sales Ord	2017-12-31 05:48	110000153837	Change Price PO	2019-05-13 14:54	110430004980
Change Price	2018-01-02 11:51	110000153837	Record Goods Re		110430004980
Create Delivery	2018-01-07 12:33	110000153837	Record Invoice	2019-06-01 14:19	110430004980
Create Delivery	2018-01-08 07:05	110000153837	Create Purchase	2019-05-10 14:54	110430004980
Record Goods Iss	2018-01-08 12:34	110000153837	Change Price PO	2019-05-13 14:54	110430004980
Record Goods Iss	2018-01-09 07:06	110000153837	Record Goods Re	2019-05-29 14:19	110430004980
Create Invoice	2018-01-11 12:34	110000153837	Record Invoice	2019-06-01 14:19	110430004980
Create Invoice	2018-01-12 07:06	110000153837	Create Purchase	2018-05-03 12:45	110430004980
Clear Invoice	2018-03-12 12:34	110000153837	Record Goods Re	2018-05-09 17:36	110430004980

Figure 6: KPIs for different processes can be calculated in the same analysis. Here an example for average throughput times for the Order-To-Cash process and the Purchase-To-Pay process. The components just reference different Event Logs in the same Data Model.



#### Filter across multiple processes

When the cases of different processes are connected, users can filter and discover dependencies across processes.

The example below shows that by filtering on one Purchase Order Item also the Order-To-Cash Process is filtered and the related Sales Order Item to this Purchase Order Item is shown.

	b da			day		
Order To Ca			Purchase To			
Activity	Eventtime	Case Key	Activity	Eventtime	Case Key	1
Create Sales Ord	2019-05-09 14:48	110000183916	Create Purchase	2019-05-10 15:03	110430004980	
Create Delivery	2019-05-15 15:10	110000183916	Change Price PO		110430004980	
Record Goods Iss	2019-05-17 06:23	110000183916	Record Goods Re	2019-05-14 09:15	110430004980	
Create Delivery	2019-05-19 09:38	110000183916	Record Invoice	2019-05-17 09:15	110430004980	
Create Invoice	2019-05-20 06:23	110000183916				
Record Goods Iss		110000183916				
Create Invoice	2019-05-24 08:46	110000183916				
Clear Invoice	2019-07-19 06:23	110000183916				
Clear Invoice	2019-07-23 08:46	110000183916				

#### Metrics across different Event Logs

There are no limitations to calculate cross-process statistics. Typical questions could be:

- How do KPIs between different processes correlate? How often for example are complaints in the service process filed for parts produced in the manufacturing process?
- How do process KPIs like Throughput Time correlate across processes?



- How much time passes from one process to another?

As we have seen in the filtering example, the Celonis PQL behaviour is the same for Multi-Event Log as before. To compare the throughput times, one could simply calculate the throughput times per process and then choose from the over 150 PQL functions to run statistical analysis on both data sets.

### Discover parallel processes

One main advantage of the Multi-Event Log technology is the ability to model and work with parallel processes (Use case 2). Parallel processes can always be looked at in a hierarchical way. An example for a hierarchical process is this car manufacturing process.

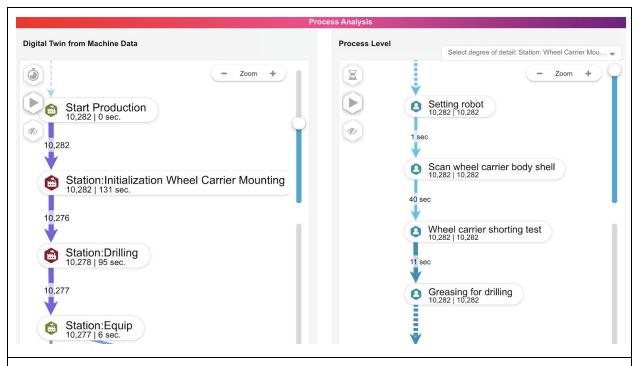


Figure 8: High-level process on the left. Granular process for each station on the right. The activities on the right are all taking place inside the activity "Station: Initialization Wheel Carrier Mounting" of the high-level process.

The top hierarchy of the process is modeled on station level. Each station is represented by one activity. Inside each station are multiple activities happening. This more granular process is in the Data Model stored in a separate Activity table and represented on the right in a separate Process Explorer.



This way of modelling can be exploited for data exploration of parallel processes. If there are parallel streams on the granular process level, they can now be visualized natively in this granular Process Explorer. As we use the granular Event Log for this Process Explorer, multiple cases can occur in parallel for each high level case. An example of this is shown below for a review process that includes two reviewers:

High-level Process	Granular Process
O	Zoom +
Author Submits Article	Invite Reviewer
Assign Editor	1 days     4 days     Reviewer Accepts Invitation     Reviewer Declines Invitation
2 days	1 3 days 41 days
Editor Selects Reviewers	Peer Review Complete
Editor Rejects Submission	Process End
● 100% - +) ↓ 100% - +)	● 100% - + ↓ 100% - +
	t has one case per reviewed paper. The granular reviewing ewer. Note that for one high-level case, multiple granular case
n occur in parallel. The Multi-Event Log	technology allows to visualize and calculate KPIs on parallel

To set up such a hierarchical Data Model that allows to analyze parallel processes, the Event Logs of different granularity can be linked through the Case tables. The number of levels of granularity is not limited. It would for example be possible to have four or five levels of granularity.



a,	Reviewer Case	×	Q.	Paper Case	×
REV	IEWER - CASE TABI C 🔅	- F	PAF	PER - CASE TABLE C	@ →
2.	Reviewer Case	×	Q_	Paper Case	×
2,	Paper Case	×	a,	Paper Case	×

# Limitations of the current Analysis regarding Multi-Event Log usage

Existing components and visualizations in the Analysis UI can be used for Multi-Event Log analyses to a great extent already. Some components are more applicable to single Event Log use cases and always access the default Activity table/Case table, those are: Throughput Time Search, Activity Explorer, Case Explorer, Selection Views (except Attribute Selection), Conformance Checker.

Some best practices to leverage the existing Analysis capabilities for MEL analyses:

- **Throughout Time Search**: can be composed with variables, button drilldowns, single KPI and histogram to calculate throughput times for a specific or across Event Logs
- Process Explorer:
  - Use Case 1: Put independent processes into context (filter across processes) Use one process explorer for each Event Log and define "Custom dimension" accordingly
  - Use Case 2: Use MERGE\_EVENTLOG functionality in "Custom dimension"
- **Case Explorer**: can be composed for each Case & Activity Table on a sheet with two OLAP tables, one holding the case information and one holding the activity information