Expanding window features: part 1

Window features

Rolling vs expanding features

Rolling window

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Expanding window

- Uses only some of the previous values at any step.
- Useful to capture recent behaviors of the time series.
- Example: the average sales over the past one month.

- Uses all previous values at any step.
- Useful when you need access to the entire time series.
- Examples:
 - aggregating a variable which has a cumulative effect.
 - target encoding.

Date	Sales
2020-02-12	35
2020-02-13	30
2020-02-14	23
2020-02-15	30
2020-02-16	34
2020-02-17	Ś

Apply a window to the time series.

Date	Sales
2020-02-12	35
2020-02-13	30
2020-02-14	23
2020-02-15	30
2020-02-16	34
2020-02-17	Ś

- Apply a window to the time series.
- Compute statistics from data inside the window (e.g., mean, min).
- To avoid data leakage, assign the statistics to timestamp after window.

Date	Sales	Sales mean	Sales min
2020-02-12	35		
2020-02-13	30		
2020-02-14	23		
2020-02-15	30		
2020-02-16	34		
2020-02-17	Ś		

- Apply a window to the time series.
- Compute statistics from data inside the window (e.g., mean, min).
- To avoid data leakage, assign the statistics to timestamp after window.

Date	Sales	Sales mean	Sales min
2020-02-12	35		
2020-02-13	30	35.0	35
2020-02-14	23		
2020-02-15	30		
2020-02-16	34		
2020-02-17	Ś		

- Apply a window to the time series.
- Compute statistics from data inside the window (e.g., mean, min).
- To avoid data leakage, assign the statistics to timestamp after window.

Date	Sales	Sales mean	Sales min
2020-02-12	35		
2020-02-13	30	35.0	35
2020-02-14	23		
2020-02-15	30		
2020-02-16	34		
2020-02-17	Ś		

- Apply a window to the time series.
- Compute statistics from data inside the window (e.g., mean, min).
- To avoid data leakage, assign the statistics to timestamp after window.
- Expand window and iterate across the time series.

Date	Sales	Sales mean	Sales min
2020-02-12	35		
2020-02-13	30	35.0	35
2020-02-14	23	32.5	30
2020-02-15	30		
2020-02-16	34		
2020-02-17	Ś		

- Apply a window to the time series.
- Compute statistics from data inside the window (e.g., mean, min).
- To avoid data leakage, assign the statistics to timestamp after window.
- Expand window and iterate across the time series.

Date	Sales	Sales mean	Sales min
2020-02-12	35		
2020-02-13	30	35.0	35
2020-02-14	23	32.5	30
2020-02-15	30	29.3	23
2020-02-16	34		
2020-02-17	Ś		

- Apply a window to the time series.
- Compute statistics from data inside the window (e.g., mean, min).
- To avoid data leakage, assign the statistics to timestamp after window.
- Expand window and iterate across the time series.

Date	Sales	Sales mean	Sales min
2020-02-12	35		
2020-02-13	30	35.0	35
2020-02-14	23	32.5	30
2020-02-15	30	29.3	23
2020-02-16	34	29.5	23
2020-02-17	Ś		

- Apply a window to the time series.
- Compute statistics from data inside the window (e.g., mean, min).
- To avoid data leakage, assign the statistics to timestamp after window.
- Expand window and iterate across the time series.

Date	Sales	Sales mean	Sales min
2020-02-12	35		
2020-02-13	30	35.0	35
2020-02-14	23	32.5	30
2020-02-15	30	29.3	23
2020-02-16	34	29.5	23
2020-02-17	Ś	30.4	23

- Apply a window to the time series.
- Compute statistics from data inside the window (e.g., mean, min).
- To avoid data leakage, assign the statistics to timestamp after window.
- Expand window and iterate across the time series.

Date	Sales	Sales mean	Sales min
2020-02-12	35	NaN	NaN
2020-02-13	30	35.0	35
2020-02-14	23	32.5	30
2020-02-15	30	29.3	23
2020-02-16	34	29.5	23
2020-02-17	Ś	30.4	23

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- Compute statistics from data inside the window (e.g., mean, min).
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- Expand window and iterate across the time series.