Rolling window features: part 1

Window features

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

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2020-02-12	35
2020-02-13	30
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2020-02-17	Ś

Apply a window to the time series.

Date	Sales	Sales rolling mean	Sales rolling min
2020-02-12	35		
2020-02-13	30		
2020-02-14	23		
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).

Date	Sales	Sales rolling mean	Sales rolling min
2020-02-12	35		
2020-02-13	30		
2020-02-14	23		
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.

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

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2020-02-12	35		
2020-02-13	30		
2020-02-14	23		
2020-02-15	30	29.3	23
2020-02-16	34		
2020-02-17	Ś		

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- 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 rolling mean	Sales rolling min
2020-02-12	35		
2020-02-13	30		
2020-02-14	23		
2020-02-15	30	29.3	23
2020-02-16	34	27.7	30
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
- Move window and iterate (i.e., roll) across the time series

Date	Sales	Sales rolling mean	Sales rolling min
2020-02-12	35		
2020-02-13	30		
2020-02-14	23		
2020-02-15	30	29.3	23
2020-02-16	34	27.7	30
2020-02-17	Ś	29	30

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

Date	Sales	Sales rolling mean	Sales rolling min
2020-02-12	35		
2020-02-13	30		
2020-02-14	23		
2020-02-15	30	29.3	23
2020-02-16	34	27.7	30
2020-02-17	Ś	29	30

• Option 1: Treat as missing data.

Date	Sales	Sales rolling mean	Sales rolling min
2020-02-12	35	NaN	NaN
2020-02-13	30	NaN	NaN
2020-02-14	23	NaN	NaN
2020-02-15	30	29.3	23
2020-02-16	34	27.7	30
2020-02-17	Ś	29	30

- Option 1: Treat as missing data.
- Drop the rows with missing data.
- Impute the missing data.
- Pros:
 - All rolling statistics have the same window size.
 - Simple.
- Cons:
 - Reduces the amount of data if dropping rows.
 - Edge cases could be quite different to rest of data.

Date	Sales	Sales rolling mean	Sales rolling min
2020-02-12	35		
2020-02-13	30		
2020-02-14	23		
2020-02-15	30	29.3	23
2020-02-16	34	27.7	30
2020-02-17	Ś	29	30

Date	Sales	Sales rolling mean	Sales rolling min
2020-02-12	35		
2020-02-13	30	35	35
2020-02-14	23		
2020-02-15	30	29.3	23
2020-02-16	34	27.7	30
2020-02-17	Ś	29	30

Date	Sales	Sales rolling mean	Sales rolling min
2020-02-12	35		
2020-02-13	30	35	35
2020-02-14	23	32.5	30
2020-02-15	30	29.3	23
2020-02-16	34	27.7	30
2020-02-17	Ś	29	30

Date	Sales	Sales rolling mean	Sales rolling min
2020-02-12	35		
2020-02-13	30	35	35
2020-02-14	23	32.5	30
2020-02-15	30	29.3	23
2020-02-16	34	27.7	30
2020-02-17	Ś	29	30

Date	Sales	Sales rolling mean	Sales rolling min
2020-02-12	35	NaN	NaN
2020-02-13	30	35	35
2020-02-14	23	32.5	30
2020-02-15	30	29.3	23
2020-02-16	34	27.7	30
2020-02-17	Ś	29	30

- Option 2: Use smaller window sizes at the edges
- Drop the rows with missing data
- Impute the missing data
- Pros:
 - Less missing data
 - Simple
- Cons:
 - Statistics at edges are based on smaller window sizes