

# Forecasting: principles and practice

Rob J Hyndman

2.4 Non-seasonal ARIMA models

# #Autoregressive models

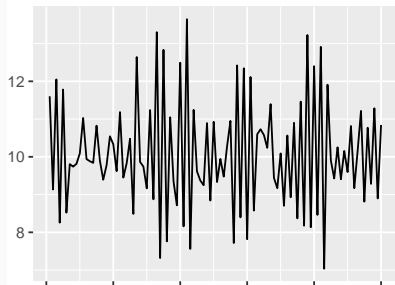
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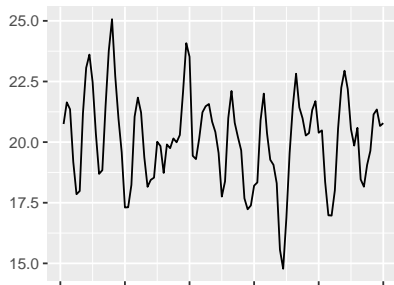
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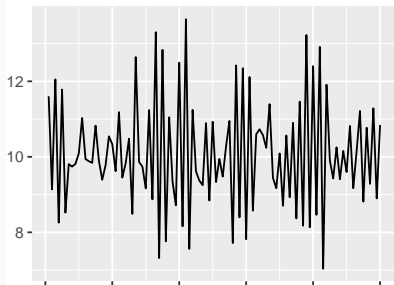
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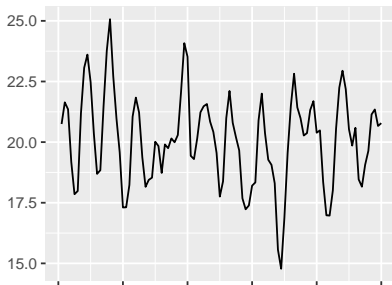
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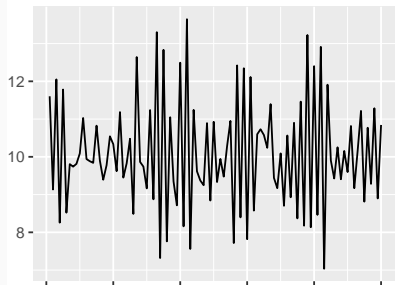
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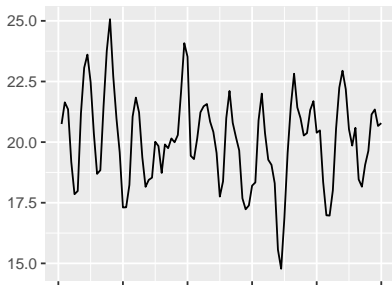
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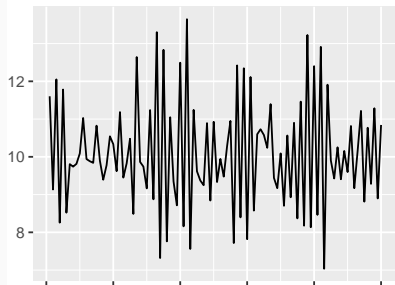
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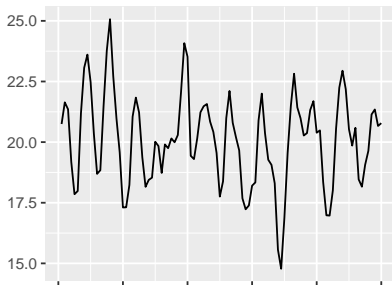
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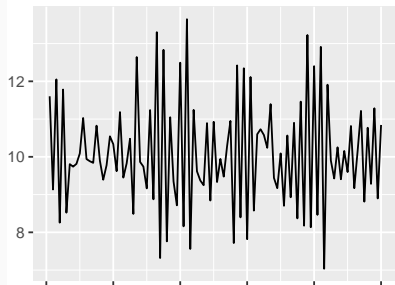
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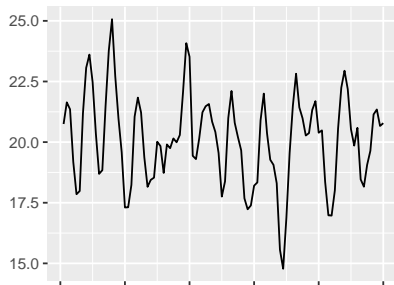
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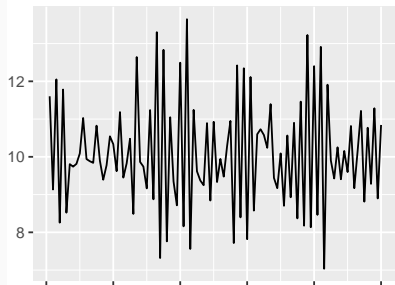
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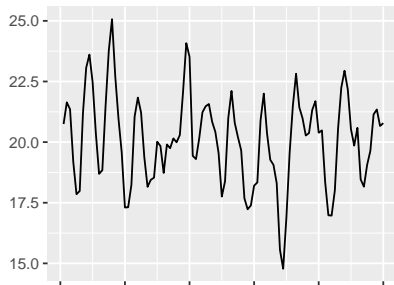
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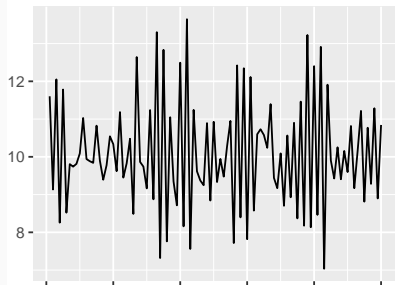
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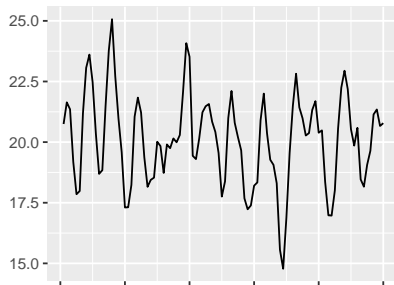
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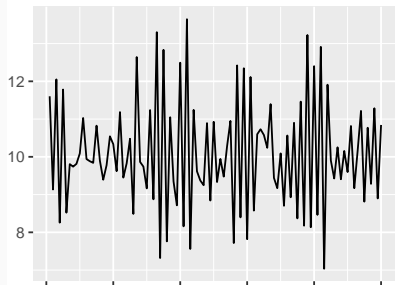
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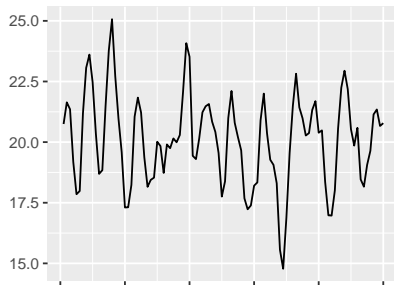
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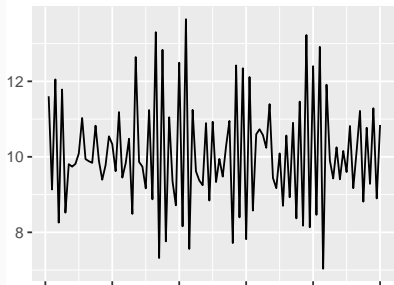
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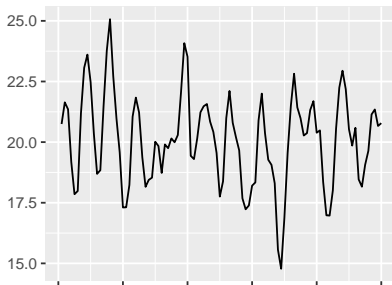
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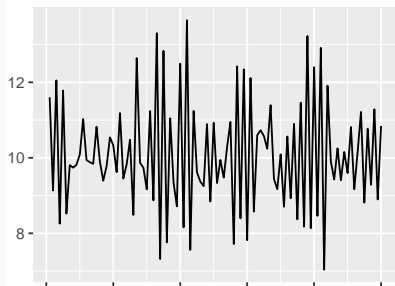
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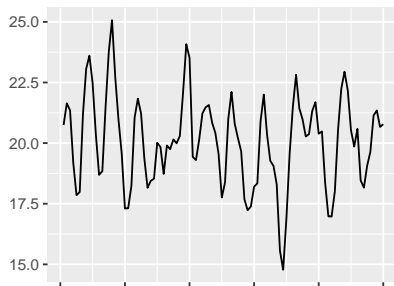
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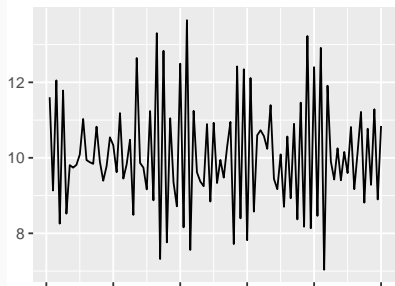
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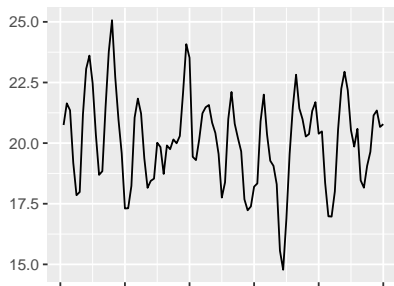
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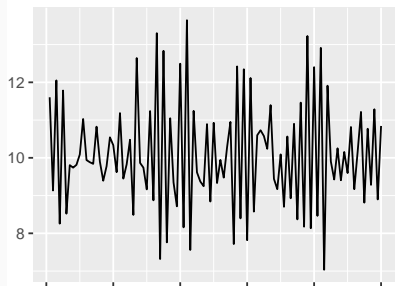
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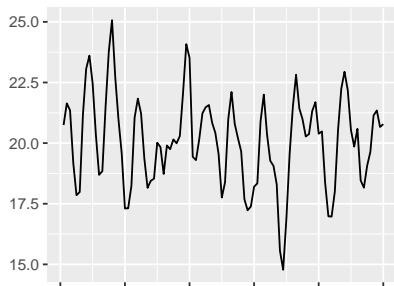
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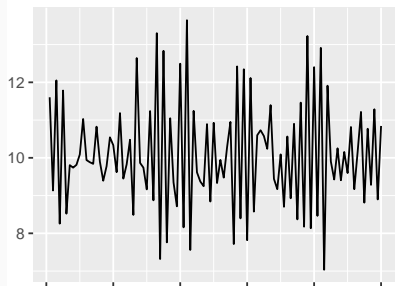
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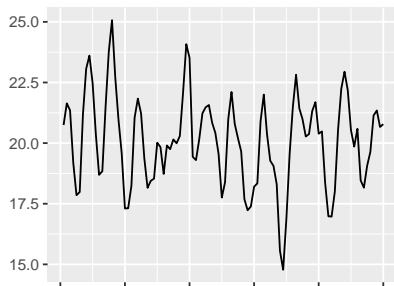
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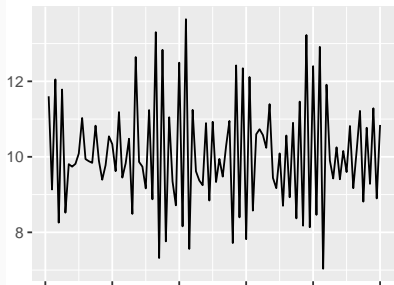
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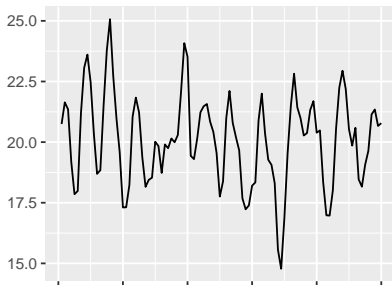
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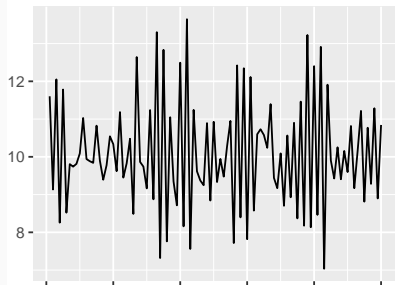
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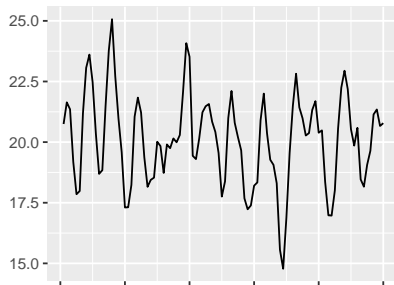
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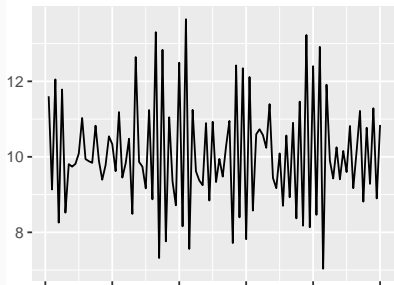
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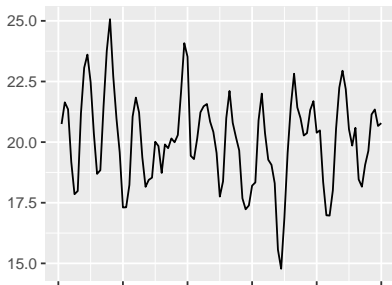
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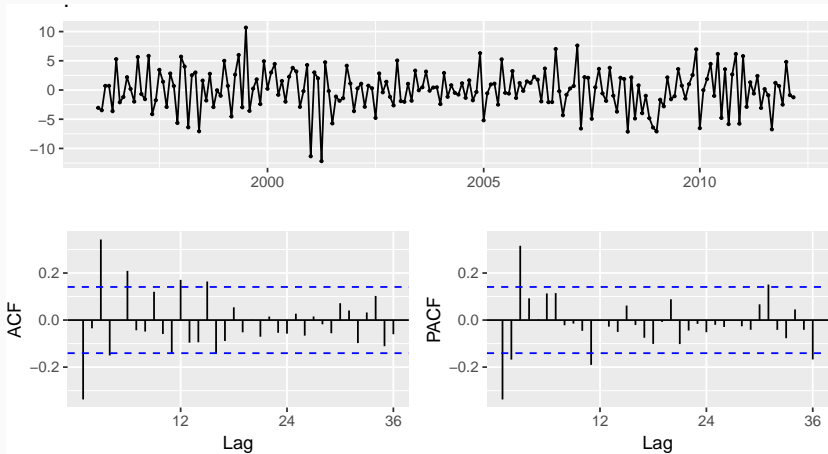


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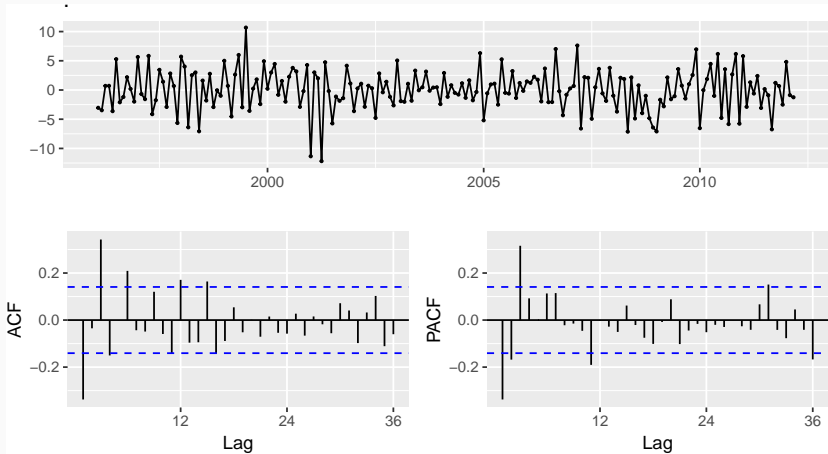
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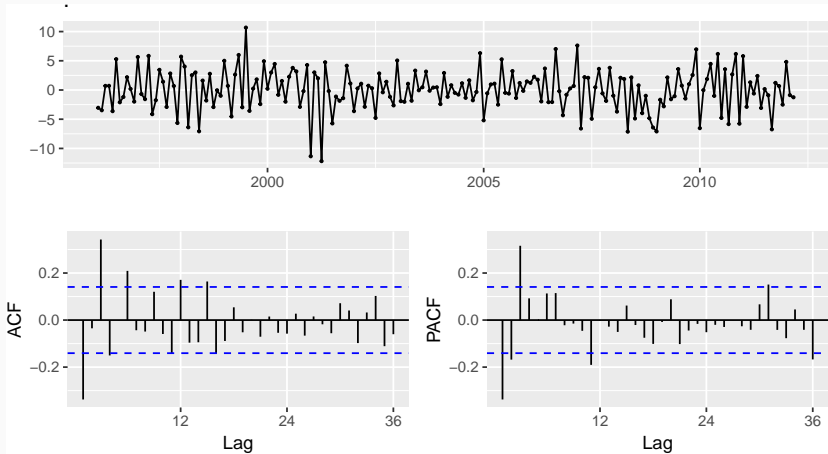
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