

# Forecasting: principles and practice

**Rob J Hyndman**

2.4 Non-seasonal ARIMA models

#Autoregressive models

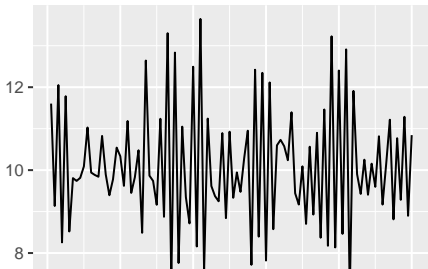
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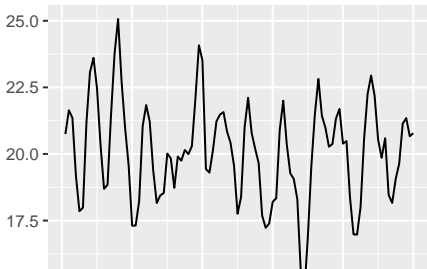
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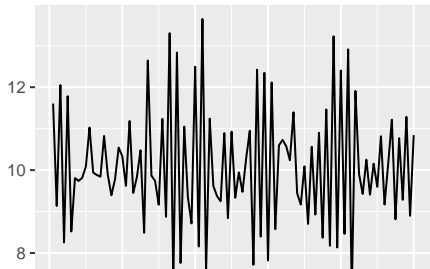
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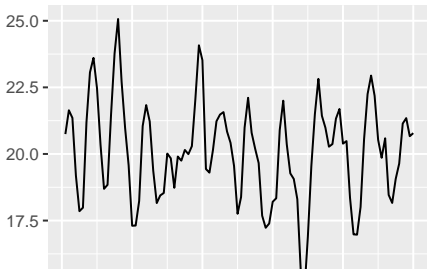
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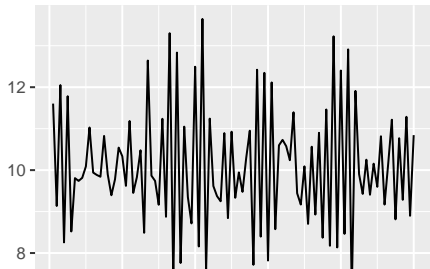
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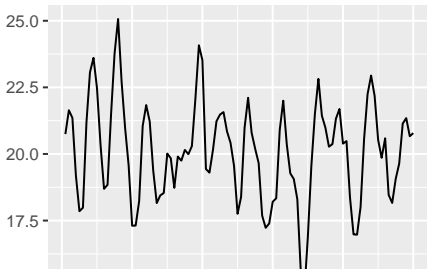
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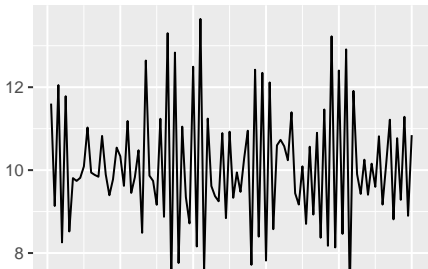
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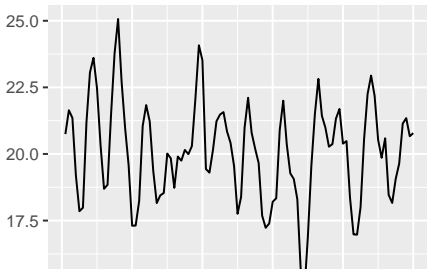
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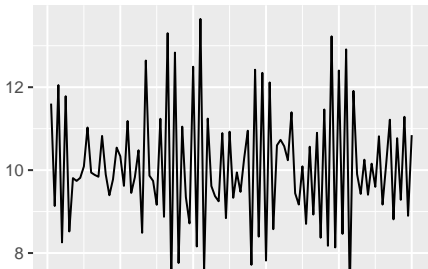
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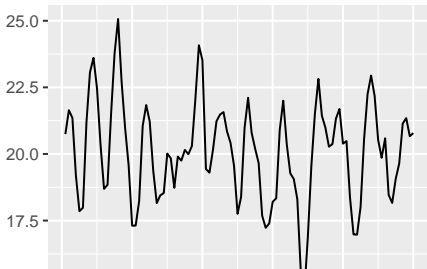
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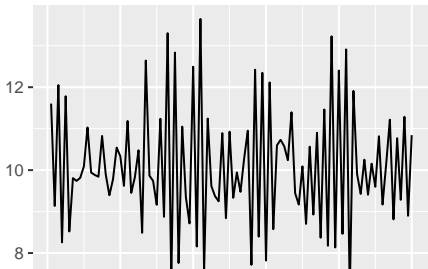
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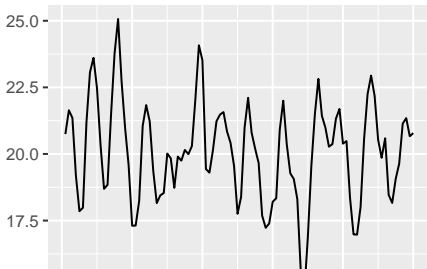
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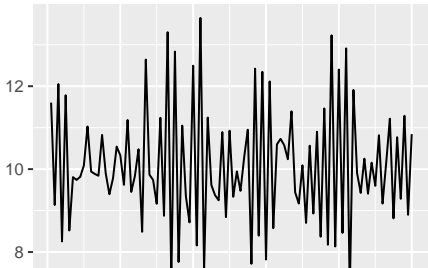
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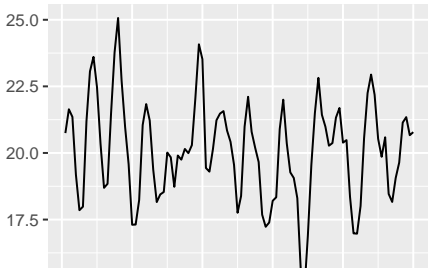
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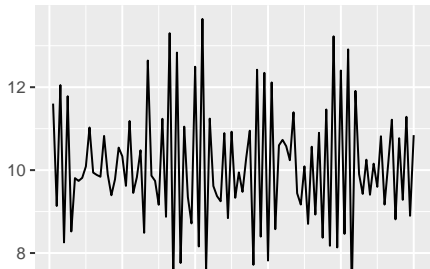
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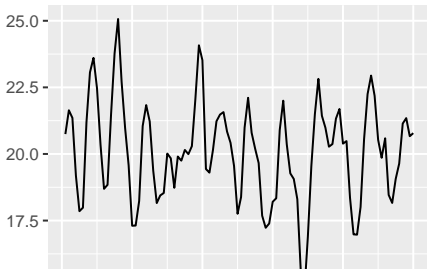
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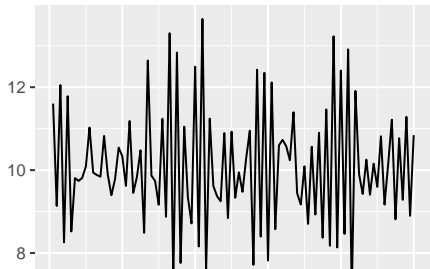
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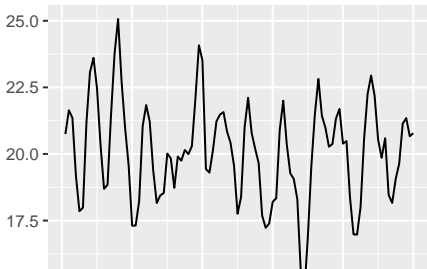
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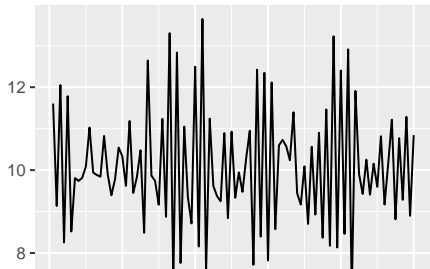
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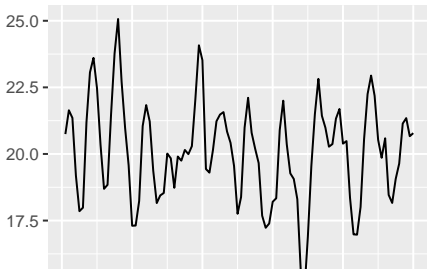
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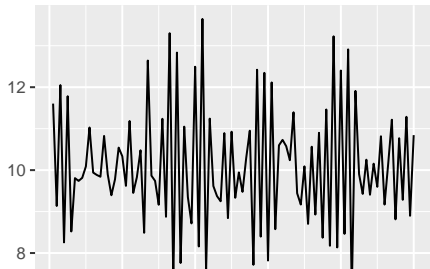
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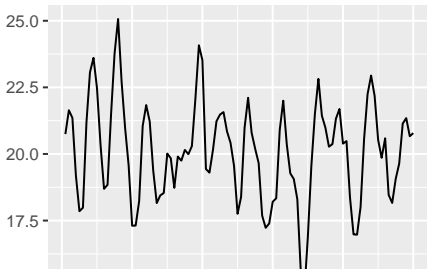
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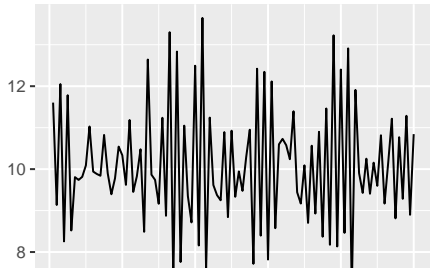
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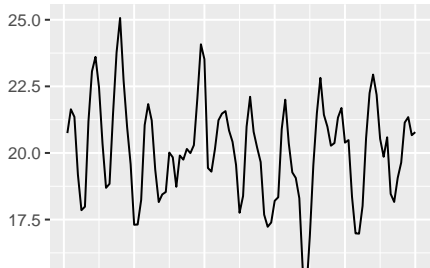
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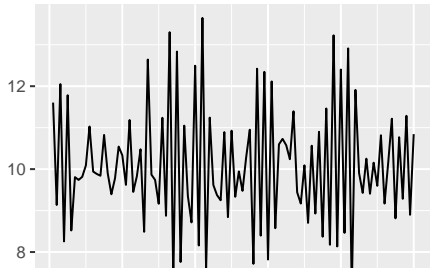
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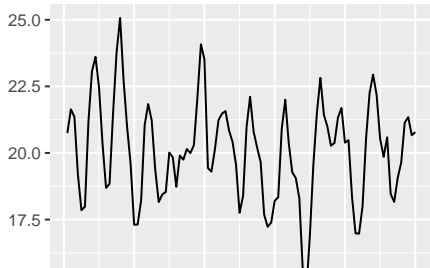
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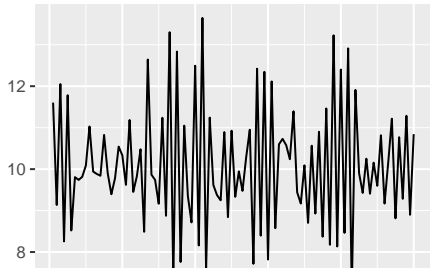
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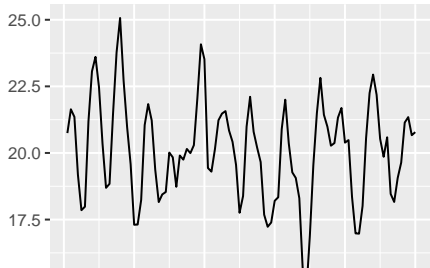
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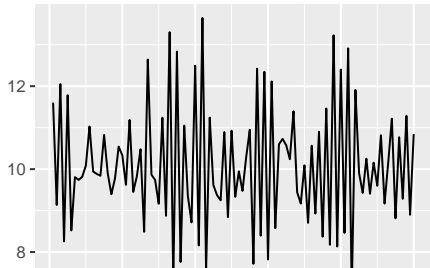
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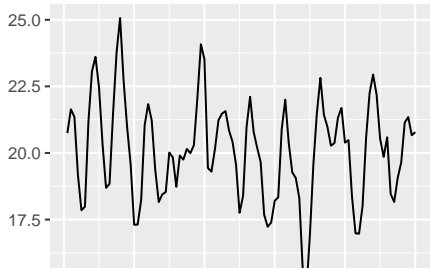
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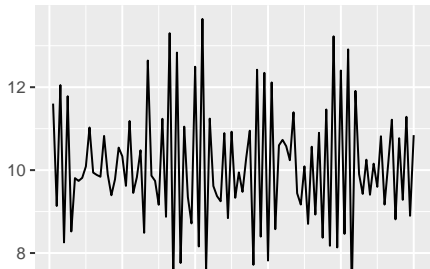
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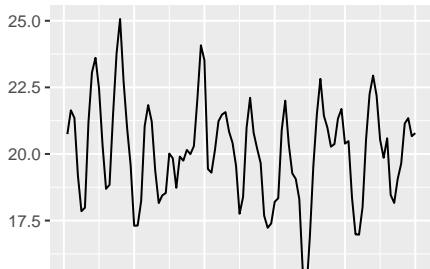
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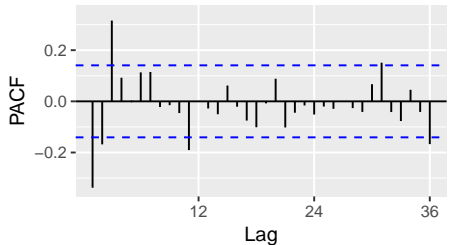
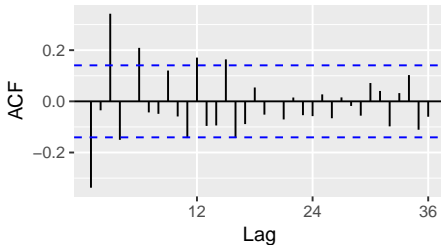
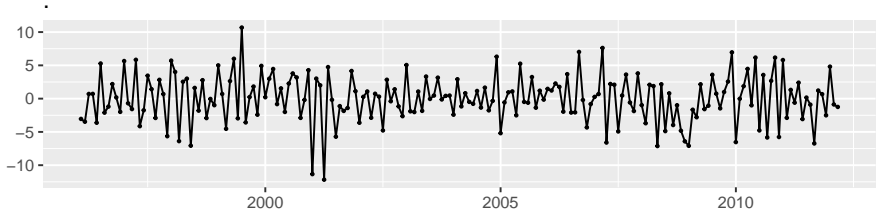


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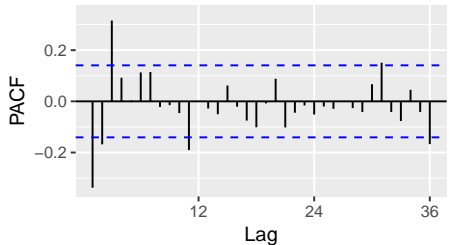
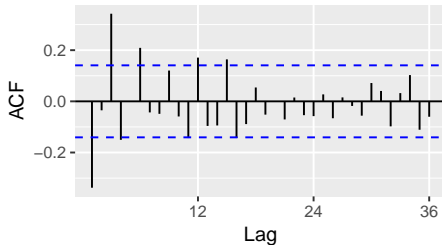
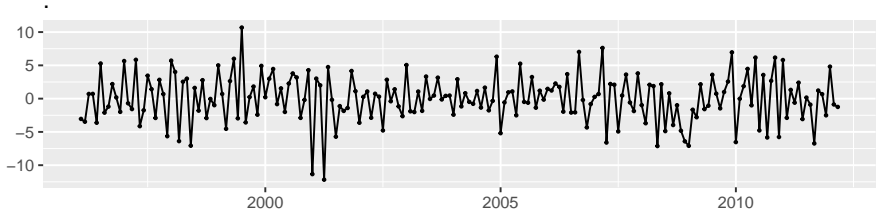
# equipment

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