

Improving quality of care in total knee arthroplasty using risk prediction: a narrative review of predictive models and factors associated with their implementation in clinical practice

Appendix A:

References

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Appendix C:

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