



UNIVERSITY OF LEEDS

This is a repository copy of *Level of accuracy of diagnoses recorded in discharge summaries: A cohort study in three respiratory wards*.

White Rose Research Online URL for this paper:
<http://eprints.whiterose.ac.uk/137734/>

Version: Accepted Version

Article:

Tsopra, R, Wyatt, JC, Beirne, P et al. (6 more authors) (2019) Level of accuracy of diagnoses recorded in discharge summaries: A cohort study in three respiratory wards. *Journal of Evaluation in Clinical Practice*, 25 (1). pp. 36-43. ISSN 1356-1294

<https://doi.org/10.1111/jep.13020>

© 2018 John Wiley & Sons, Ltd. This is the peer reviewed version of the following article: Tsopra R, Wyatt JC, Beirne P, et al. Level of accuracy of diagnoses recorded in discharge summaries: A cohort study in three respiratory wards. *J Eval Clin Pract*. 2018;1–8, which has been published in final form at <https://doi.org/doi.org/10.1111/jep.13020>. This article may be used for non-commercial purposes in accordance with Wiley Terms and Conditions for Self-Archiving. Uploaded in accordance with the publisher's self-archiving policy.

Reuse

Items deposited in White Rose Research Online are protected by copyright, with all rights reserved unless indicated otherwise. They may be downloaded and/or printed for private study, or other acts as permitted by national copyright laws. The publisher or other rights holders may allow further reproduction and re-use of the full text version. This is indicated by the licence information on the White Rose Research Online record for the item.

Takedown

If you consider content in White Rose Research Online to be in breach of UK law, please notify us by emailing eprints@whiterose.ac.uk including the URL of the record and the reason for the withdrawal request.



eprints@whiterose.ac.uk
<https://eprints.whiterose.ac.uk/>

TITLE PAGE

TITLE OF THE ARTICLE

Level of accuracy of diagnoses recorded in discharge summaries: a cohort study in three respiratory wards.

AUTHORS' NAMES

Rosy TSOPRA, Jeremy C WYATT, Paul BEIRNE, Kirsty RODGER, Matthew CALLISTER, Dipansu GHOSH,
Ian J CLIFTON, Paul WHITAKER, Daniel PECKHAM

Rosy TSOPRA

MD, PhD, Postdoctoral Fellow

Leeds Centre for Respiratory Medicine, St James's University Hospital, Beckett Street, Leeds
LS9 7TF, United Kingdom

Leeds Institute of Health Sciences, 101 Clarendon Rd, Leeds LS2 9LJ, United Kingdom

Université Paris 13, , F93000 Bobigny, France

AP-HP, Paris, France

Jeremy C WYATT

Professor of Digital Healthcare

Wessex Institute of Health & Research, Faculty of Medicine, University of Southampton,
SO16 7NS, United Kingdom

Paul BEIRNE

Consultant Respiratory Physician

Leeds Centre for Respiratory Medicine, St James's University Hospital, Beckett Street, Leeds

LS9 7TF, United Kingdom

Kirsty RODGER

Consultant Respiratory Physician

Leeds Centre for Respiratory Medicine, St James's University Hospital, Beckett Street, Leeds

LS9 7TF, United Kingdom

Matthew CALLISTER

Consultant Respiratory Physician

Leeds Centre for Respiratory Medicine, St James's University Hospital, Beckett Street, Leeds

LS9 7TF, United Kingdom

Dipansu GHOSH

Consultant Respiratory Physician

Leeds Centre for Respiratory Medicine, St James's University Hospital, Beckett Street, Leeds

LS9 7TF, United Kingdom

Ian J CLIFTON

Consultant Respiratory Physician

Leeds Centre for Respiratory Medicine, St James's University Hospital, Beckett Street, Leeds

LS9 7TF, United Kingdom

Paul WHITAKER

Consultant Respiratory Physician

Leeds Centre for Respiratory Medicine, St James's University Hospital, Beckett Street, Leeds

LS9 7TF, United Kingdom

Daniel PECKHAM

Clinical Associate Professor

Leeds Centre for Respiratory Medicine, St James's University Hospital, Beckett Street, Leeds

LS9 7TF, United Kingdom

Leeds Institute of Biomedical & Clinical Sciences, St James's University Hospital, Beckett Street, Leeds LS9 7TF, United Kingdom

INSTITUTIONS AT WHICH THE WORK WAS CARRIED OUT

Leeds Centre for Respiratory Medicine, St James's University Hospital, Beckett Street, Leeds LS9 7TF,
United Kingdom

CORRESPONDING AUTHOR

Dr Rosy TSOPRA

Postal address:

Université Paris 13, F93000 Bobigny, France

AP-HP, Paris, France

e-mail: rosy.tsopra@aphp.fr

Telephone number: +33 1 48 38 73 34

SUGGESTED RUNNING TITLE OF MORE THAN FIFTY CHARACTERS

Accuracy of discharge summaries

KEYWORDS

Patient discharge summaries, Diagnosis, Data Accuracy, Quality of health care, ICD 10 coding, Coding

WORD COUNT (*Introduction, methods, results, discussion, conclusion*): 2768 words

NUMBER OF FIGURES: 4

NUMBER OF TABLES: 2

NUMBER OF SUPPLEMENTARY FILES FOR ONLINE PUBLICATION: 0

NUMBER OF REFERENCES: 38

ABSTRACT

Rationale: One of the key functions of the discharge summary is to convey accurate diagnostic description of patients. Inaccurate or missing diagnoses may result in a false clinical picture, inappropriate management, poor quality of care and a higher risk of re-admission. While several studies have investigated the presence or absence of diagnoses within discharge summaries, there are very few published studies assessing the accuracy of these diagnoses. The aim of this study was to measure the accuracy of diagnoses recorded in sample summaries, and to determine if it was correlated with the type of diagnoses (e.g. “respiratory” diagnoses), the number of diagnoses or the length of patient stay.

Methods: A prospective cohort study was conducted in 3 respiratory wards in a large UK NHS Teaching Hospital. We determined the reference list of diagnoses (the closest to the true state of the patient based on consultant knowledge, patient records, and laboratory investigations) for comparison with the diagnoses recorded in a discharge summary. To enable objective comparison, all patient diagnoses were encoded using a standardized terminology (ICD-10). Inaccuracy of the primary diagnosis alone and all diagnoses in discharge summaries was measured, and then correlated with type of diseases, number of diagnoses and length of patient stay.

Results: 107 of 110 consecutive discharge summaries were analyzed. The mean inaccuracy rate per discharge summary was 55% [95% CI 52 to 58%]. Primary diagnoses were wrong, inaccurate, missing or mis-recorded as a secondary diagnosis in half the summaries. The inaccuracy rate was correlated with the type of disease but not with number of diagnoses nor length of patient stay.

Conclusion: Our study showed that diagnoses were not accurately recorded in discharge summaries, highlighting the need to measure and improve discharge summary quality.

MAIN TEXT

INTRODUCTION

The discharge summary is an important communication tool for promoting the quality, safety and continuity of care ^{1,2}. It enables the efficient transfer of information between health care professionals in primary and secondary care settings ^{3,4}, and forms part of the permanent patient record ⁴. One of the key functions of the discharge summary is to convey accurate diagnostic information by identifying the primary reason for admission (primary diagnosis), and other relevant diagnoses including co-morbidities (secondary diagnoses). Inaccurate or missing diagnoses may result in a false clinical picture, inappropriate management ⁵, poor quality of care ⁵ and a higher risk of re-admission ⁶. Inaccuracies in the list of diagnoses recorded in a discharge summary will also have wider impact on key processes including clinical coding ^{7,8}, research ⁷, surveillance ^{9,10}, clinical audit ⁷, quality improvement and financial remuneration of the hospital ⁷.

Despite abundant literature ^{5,11-13} highlighting the importance of accurate and complete diagnosis capture, diagnoses are often missing in discharge summaries. A review of the literature by Kripalani et al. ⁵ found that diagnoses were recorded in only 17.5% and 28% of discharge summaries respectively. While several studies have investigated the presence or absence of diagnoses ^{5,12,14-16}, there are very few published studies assessing the accuracy of diagnoses within discharge summaries.

The aim of this article is therefore to measure the accuracy of diagnoses recorded in a sample of discharge summaries, and to determine if this was correlated with the type of diagnoses (e.g. “respiratory” diagnoses), the number of diagnoses or the length of patient stay.

METHODS

To evaluate the accuracy of discharge summary diagnoses, we designed the following 5-step method (Figure 1):

- Step 1: Patient selection
- Step 2: Determination of the reference list of diagnoses for each patient
- Step 3: Determination of the list of diagnoses recorded in discharge summary for each patient
- Step 4: Comparison of the list of discharge summary diagnoses versus the reference list of diagnoses for each patient
- Step 5: Calculating the inaccuracy rate of diagnoses recorded in the discharge summaries

To allow objective comparison of the list of discharge summary diagnoses versus the reference list of diagnoses for each patient, we choose to standardize the medical vocabulary using a medical terminology. Each diagnosis was translated into an encoded diagnosis according to ICD 10 terminology, the most widely used medical terminology to code diagnoses¹⁷. Encoded diagnoses in ICD 10 begin with a letter followed by alphanumeric characters (e.g. J15.7 “Pneumonia due to *Mycoplasma pneumoniae*”). Each diagnosis was coded with the greatest possible level of accuracy, i.e. using the third, fourth or fifth character.

Step 1: Patient selection

We selected all consecutive adult patients discharged from three respiratory wards at St James’s University Hospital Leeds, in March 2015. Exclusion criteria included a non-respiratory primary diagnosis or a missing discharge summary from the case notes.

Step 2: Determination of the reference list of diagnoses for each patient

To obtain a reference list of diagnoses for each patient that was as close as possible to the true state of the patient, we recruited the consultant responsible for that patient's care episode and a specialist respiratory coder. At point of the discharge, the consultant determined the reference diagnoses for the patient. He/she identified the primary diagnosis and any other diagnoses for each patient using his/her knowledge about the patient, the patient notes and the test results. The coder helped him/her to standardize diagnoses by translating them into ICD 10 codes (e.g. "Acute severe asthma" was translated in "J46 Status asthmaticus"). This resulted in a reference list of encoded diagnoses for each patient. Both consultant and coder were blinded to the content of the discharge summary during step 2.

Step 3: Determination of the list of diagnoses recorded in discharge summary for each patient

In our teaching hospital, discharge summaries are written in free text by junior doctors using a basic electronic template. To obtain the list of diagnoses recorded in the discharge summaries for each patient, we recruited a consultant naive to the clinical case and a specialist coder. At least 3 weeks after the discharge, they determined and coded the patient's discharge summary diagnoses together. This resulted in a list of encoded diagnoses recorded in the discharge summaries for each patient.

Both consultant and coder were blinded to the case notes during step 3.

To reduce bias, the discharge summary diagnoses were extracted: (i) by a consultant naïve to the case, to guarantee that the diagnoses were extracted from the discharge summary and not influenced by prior knowledge about the patient, (ii) by the same coder involved in the determination of the reference diagnoses, to guarantee that the translation of the diagnoses into

encoded diagnoses was consistent; (iv) at least three weeks after the determination of the reference list of diagnoses, to reduce memory effects in the coder.

Step 4: Comparison of the list of discharge summary diagnoses versus the reference list of diagnoses for each patient

For each patient, a third person not involved in the generation of the two diagnostic lists (RT) compared the encoded discharge summary diagnoses to the encoded reference list of diagnoses. We followed the ICD 10 structure, and thus distinguished exact ICD-10 diagnosis matches at 3 or 4 character levels from matches at block title level only or matches at clinical connection level only (i.e. medical meaning is similar but there is no code similarity). This matching process generated 5 well-defined accuracy categories:

- Accurate diagnosis: same ICD10 for at least the first 3 characters
- Partial inaccurate diagnosis: same ICD10 block titles only
- Serious inaccurate diagnosis: clinical similarity only
- Missing diagnosis: “reference” diagnosis not present in the discharge summary
- Wrong diagnosis: discharge summary diagnosis not present in the reference list of diagnoses for that patient

Examples of each category of accuracy are shown in Figure 2.

Step 5: Calculating the inaccuracy rate of diagnoses recorded in discharge summaries

The inaccuracy rate per discharge summary, p , was defined as follows:

$$p\% = \frac{\text{Partial inaccurate diagnoses} + \text{Serious inaccurate diagnoses} + \text{Missing diagnoses} + \text{Wrong diagnoses}}{\text{Same diagnoses} + \text{Partial inaccurate diagnoses} + \text{Serious inaccurate diagnoses} + \text{Missing diagnoses} + \text{Wrong diagnoses}}$$

The inaccuracy rate was measured separately for all diagnoses (i.e. primary and secondary diagnoses), and for the primary diagnosis alone. The percentage of discharge summaries containing less than half and greater than half of inaccurate diagnoses was also assessed.

The relationship between the inaccuracy rate and the type of diseases (e.g. respiratory disease) was studied. The type of diseases was determined according to the ICD 10 chapter. We carried out a Kruskal Wallis test for all types of disease including at least 50 diagnoses, excluding diseases where the number of diagnoses was too small.

The correlations between inaccuracy rate per discharge summary, number of diagnoses per stay and the length of patient stay were studied for each patient. Data was analysed using the Pearson correlation coefficient and Pearson correlation test. All results are presented with 95% confidence intervals. A p value of <0.05 was taken as significant (Software R version 3.2.2).

RESULTS

We were able to complete the described 5-step process in 107 (97%) of 110 discharge summaries (Figure 1). Thirty-two patients had already been excluded from the initial cohort because the discharge summaries were not completed or the primary diagnosis was not respiratory. Two patients were excluded due to a delay in receiving the notes and one due to a breach in the study protocol (the consultant and the coder focused on the wrong care episode).

1) Characteristics of the case study cohort

Median (range) patient age was 67 years [22-95]. 51% were female. Patients had a median (range) number of 12 reference diagnoses [2-34], and a median (range) length of stay of 6 days [<1 -80].

2) Inaccuracy rate of diagnoses recorded in discharge summaries

For all diagnoses:

More than half of the discharge summary diagnoses were inaccurate in 67% [58 to 76%] of discharge summaries (Figure 3). The mean inaccuracy rate per discharge summary was 55% [52 to 58%] (Table 1). The majority of inaccurate diagnoses corresponded to missing diagnoses (41% [38 to 44%]). About 7% [5 to 9%] corresponded to diagnoses written in the discharge summary which were completely wrong. Other corresponded to partial or seriously inaccurate diagnoses.

For primary diagnosis:

The inaccuracy rate for primary diagnoses was 46% (Table 2). The majority of primary diagnoses were partially inaccurate (e.g. "asthma" instead of "near fatal asthma") or seriously inaccurate (e.g. "haemoptysis" instead of "community acquired pneumoniae") and/or due to incorrect position (i.e. the primary diagnosis was considered to be a secondary diagnosis in the reference list). 5% [1 to 9%] of discharge summary primary diagnoses were completely wrong (e.g. "chronic obstructive pulmonary disease" was recorded when it was absent clinically).

4) Correlation of the inaccuracy rate of recorded diagnoses per summary with the type of disease

The inaccuracy rate was significantly correlated with the type of diseases ($p=0.004$, Figure 4). Diagnoses associated with respiratory, neoplastic, psychiatric and rheumatological diseases contained fewer inaccuracies than infectious diseases (e.g. 80% [64 to 96%] for infection vs 50% [36 to 65%] for respiratory).

5) Correlation of the inaccuracy rate of recorded diagnoses per discharge summary with the number of diagnoses

The inaccuracy rate in each summary was not correlated with the number of diagnoses per summary ($p=0.83$).

6) Correlation of the inaccuracy rate of recorded diagnoses per discharge summary with the length of stay

The diagnostic inaccuracy rate for each summary was not correlated with the patient's length of stay ($p=0.27$).

DISCUSSION

We assessed the inaccuracy rate of primary and secondary diagnoses in discharge summaries. Our results demonstrate that over 50% of all the diagnoses in discharge summaries were inaccurate, and that half of the primary diagnoses are also inaccurate, missing, wrong or considered as a secondary diagnosis. This inaccuracy rate was not correlated with the number of diagnoses or length of stay, but was correlated with the type of diseases. Diagnoses related to infectious diseases had the highest inaccuracy rate in our cohort.

Our study took into account the key methodological elements described by Hogan et al.¹⁸ for studies designed to assess accuracy, using a prospective cohort design with consecutive cases to avoid sampling bias¹⁸. Only 2.7% of cases were unavailable for the final analysis. Further strengths of our study included determination of the patient's reference diagnoses blind to the discharge summary, and extraction of discharge summary diagnoses blind to the notes, to allow unbiased comparison by a third person not involved in the generation of the diagnosis list¹⁸. We also established a robust list of reference diagnoses by using the consultant responsible for the care episode, who was the person the most knowledgeable about the case¹⁸, and a specialist coder. A further strength of our method is the encoding of diagnoses using ICD 10, allowing us to measure objectively the accuracy of discharge summary diagnoses. Indeed the use of medical terminologies¹⁹ such as SNOMED CT or ICD 10¹⁷ standardises the medical vocabulary, so providing an objective approach to making a comparison and grading the level of accuracy on a 5-point scale; no other study has done this for the measurement of diagnostic accuracy.

In our study, bias could have been introduced by exposing physicians to the rather alien ICD 10 language. This disconnection between medical thinking and the terms used in terminologies²⁰ was reduced by supporting the physicians with the same specialist coder, to ensure reliable translation of medical diagnoses into encoding diagnoses. To avoid memory effects of the specialist coder, a period of wash out of at least 3 weeks was respected between the derivation of the both lists, and the discharge summary was anonymized. The study may have been limited by the fact we focused on accuracy of diagnoses and not impact of inaccuracy on clinical management. While a diagnosis of hypercholesterolemia would be very relevant to a patient with angina, it would have less acute impact on an individual with tonsillitis. Despite carrying out this study in a blind fashion, clinical teams may have become aware of the study through informal communication. This may have changed their behaviour towards writing discharge summaries²¹, although every effort was taken to limit this potential bias. However, the high level of inaccurate diagnoses in discharge summaries suggests that junior doctors remained unaware of the study. The study cohort originated in a single busy acute speciality in a large UK teaching hospital, so results may not be generalizable to all specialities, or all hospitals. However, most patients had multiple diagnoses so are likely to reflect the case mix in general medicine. In our hospital, discharge summaries are written in free text by junior doctors without software support by using a basic electronic template, and at the time of discharge, whereas in other hospitals, they may be written or dictated by consultants following discharge when all tests results are available. To test the generality of our results, further cohort studies in other specialties and other hospitals are clearly needed.

Our study focused on inaccuracy in diagnoses and not in other discharge summary components, such as procedures carried out, medication lists^{12,16,22}, investigation results^{12,16} or follow up^{12,16}. There are very few published studies assessing inaccuracy in primary and secondary diagnoses recorded in discharge summaries. Only limited information is available about the accuracy of final diagnoses showing inaccuracy occurring in 4.5% to 20% of discharge summaries^{12,16,23,24}. However, these

studies are severely limited by the methodology used: (i) they focused only on final diagnosis, (ii) some of them focused on one type of diagnosis (e.g. delirium) ^{24,25}; (iii) the determination of reference patient diagnosis may cause bias (e.g. lack of blinded assessment) ^{12,16,23}; (iv) the comparison between reference diagnosis and discharge summary diagnosis was done by a person involved in the generation of the diagnosis lists from the discharge summary ^{12,16}; (v) the metrics and methods used to measure accuracy are not clearly described ^{12,16,23}.

The high level of diagnostic inaccuracy in our study is likely to reflect lack of time and/or knowledge of junior doctors about the importance of specifying diagnoses to a high level of accuracy ²⁶. Even though the national college of physicians has proposed standards for discharge summaries ¹³, junior doctors still have no formal teaching on how to generate an accurate summary ²⁶. The process is often burdensome and is low priority for hectic front line clinicians. Delivering timely discharge summaries in the secondary care environment can be very time consuming, requiring approximately 10 +/- 3.5 hours per week for a junior doctor to complete the forms ²⁷. With high patient turnover and bed shortages, junior doctors are also under additional time pressure to deliver the summaries prior to patient discharge, as delays in delivering discharge summary to primary care may impact on community follow up and the quality of care ^{5,6,28,29}. This pressure on junior doctor time can be expensive ²⁷ and may impact on safe clinical care and reduce the quality of the discharge summaries. With the implementation of shift patterns, junior doctors may also discharge and write summaries on patients whom they have not previously seen. While advances in information technology have been significant, there is an absence of effective and intuitive software to support them in marshalling the relevant information and helping them complete the summary.

CONCLUSION

The results of our cohort study highlight the need to improve the accuracy of discharge summary diagnoses in respiratory wards, even in a teaching hospital. Those responsible for the quality and

safety of discharge summaries elsewhere might simply assume that such quality issues do not apply to all categories of wards, and to all institutions, but it would be more rational to apply our method to assess the quality of discharge summaries in their own organisation. Assuming that our findings are replicated in other centres, future research should focus on education and assessing how best to help junior doctors to deliver high quality, accurate discharge summaries within a hectic clinical environment. One potential solution would be to involve junior doctors in this quality improvement activity by alerting them to the problem and asking them to propose solutions^{30,31}. Another would be to electronically share diagnoses between specialties in an electronic patient record^{32,33}, avoiding duplication and allowing the creation of a more accurate in-depth record available at the time of discharge³⁴⁻³⁶. The introduction of templates^{5,37} with speciality-specific drop-down menus of diagnoses and a simple hierarchy for less frequent conditions might also support junior doctors generating more accurate discharge summaries. These templates should be displayed in a usable interface to increase user satisfaction, confidence and stimulate better uptake³⁸.

DECLARATIONS

ACKNOWLEDGMENTS

We would like to thank members of the coding department of St James's University Hospital for their help and support. Michael Routledge, the specialist respiratory coder, who coded all the notes and discharge summaries. Natasha Noble and Rebecca Marshall, the runners, who collected the notes. Ben Philliskirk and Catherine Hutchinson, the coders, who were involved in the official coding. Tracey Conroy and Victoria Anne Macwhirter, the coder leaders for their support and expert advice.

ETHICAL APPROVAL AND CONSENT TO PARTICIPATE

Not applicable. This study is part of service evaluation and audit. We have not collected extra data but assessed data collected as part of routine practice. This does not need ethics or patient consent.

CONSENT FOR PUBLICATION

Not applicable

AVAILABILITY OF DATA AND MATERIAL

The datasets generated and/or analysed during the current study are not publicly available due to use of these data for other research studies not yet published, but are available from the corresponding author on reasonable request.

COMPETING INTERESTS

The authors declare that they have no competing interests

FUNDING

Funding for RT salary from the Yorkshire & Humber NIHR CLAHRC

AUTHORS' CONTRIBUTIONS

Design of the study protocol: RT

Setting up the study in hospital: RT, DP

Derivation of the gold standard: IC, PB, PW, KR, MC, DG

Extraction and coding of diagnoses from discharge summaries: IC, PW, MC, DG

Analysis of Data and statistical analysis: RT

Writing the manuscript: RT

Revising the manuscript critically: DP, JW

Adding relevant suggestions to improve the manuscript: IC, PB, PW, KR, MC, DG

Agreement for all aspects of the work and approval of the final version to be published: RT, DP, JW,
IC, PB, PW, KR, MC, DG

REFERENCES

1. Long A, Atkins JB. Communications between general practitioners and consultants. *Br Med J*. 1974;4(5942):456-459.
2. Harding J. Study of discharge communications from hospital doctors to an inner London general practice. *J R Coll Gen Pract*. 1987;37(304):494-495.
3. Kripalani S, Jackson AT, Schnipper JL, Coleman EA. Promoting effective transitions of care at hospital discharge: a review of key issues for hospitalists. *J Hosp Med*. 2007;2(5):314-323. doi:10.1002/jhm.228.
4. Wimsett J, Harper A, Jones P. Review article: Components of a good quality discharge summary: a systematic review. *Emerg Med Australas EMA*. 2014;26(5):430-438. doi:10.1111/1742-6723.12285.
5. Kripalani S, LeFevre F, Phillips CO, Williams MV, Basaviah P, Baker DW. Deficits in communication and information transfer between hospital-based and primary care physicians: implications for patient safety and continuity of care. *JAMA*. 2007;297(8):831-841. doi:10.1001/jama.297.8.831.
6. van Walraven C, Seth R, Austin PC, Laupacis A. Effect of Discharge Summary Availability During Post-discharge Visits on Hospital Readmission. *J Gen Intern Med*. 2002;17(3):186-192. doi:10.1046/j.1525-1497.2002.10741.x.
7. Burns EM, Rigby E, Mamidanna R, et al. Systematic review of discharge coding accuracy. *J Public Health Oxf Engl*. 2012;34(1):138-148. doi:10.1093/pubmed/fdr054.
8. Campbell SE, Campbell MK, Grimshaw JM, Walker AE. A systematic review of discharge coding accuracy. *J Public Health Med*. 2001;23(3):205-211.
9. Murff HJ, Forster AJ, Peterson JF, Fiskio JM, Heiman HL, Bates DW. Electronically Screening Discharge Summaries for Adverse Medical Events. *J Am Med Inform Assoc JAMIA*. 2003;10(4):339-350. doi:10.1197/jamia.M1201.
10. Melton GB, Hripcsak G. Automated detection of adverse events using natural language processing of discharge summaries. *J Am Med Inform Assoc JAMIA*. 2005;12(4):448-457. doi:10.1197/jamia.M1794.
11. Newton J, Eccles M, Hutchinson A. Communication between general practitioners and consultants: what should their letters contain? *BMJ*. 1992;304(6830):821-824.
12. Legault K, Ostro J, Khalid Z, Wasi P, You JJ. Quality of discharge summaries prepared by first year internal medicine residents. *BMC Med Educ*. 2012;12:77. doi:10.1186/1472-6920-12-77.
13. Royal College of Physicians on behalf of the Health and Social Care Information Centre. *Technical Annex to the Consistent Structure and Content Standards for Admission, Handover, Discharge, Outpatient and Referral Records and Communications.*; 2013. https://www.rcplondon.ac.uk/sites/default/files/standards-for-admission-handover-discharge-outpatient-and-referral_-technical-annex-2013_0.pdf. Accessed September 21, 2015.

14. Tulloch AJ, Fowler GH, McMullan JJ, Spence JM. Hospital discharge reports: content and design. *Br Med J*. 1975;4(5994):443-446.
15. Callen JL, Alderton M, McIntosh J. Evaluation of electronic discharge summaries: a comparison of documentation in electronic and handwritten discharge summaries. *Int J Med Inf*. 2008;77(9):613-620. doi:10.1016/j.ijmedinf.2007.12.002.
16. Abdul-Rahman IS. Adequacy of Discharge Summaries Prepared by Junior Medical Residents in a University Hospital. *Clin Med Res*. 2012;1(1):1. doi:10.11648/j.cmr.20120101.11.
17. World Health Organization. ICD 10 - International Statistical Classification of Diseases and Related Health Problems 10th Revision -. <http://apps.who.int/classifications/icd10/browse/2016/en>. Accessed September 21, 2015.
18. Hogan WR, Wagner MM. Accuracy of Data in Computer-based Patient Records. *J Am Med Inform Assoc*. 1997;4(5):342-355.
19. Cimino JJ. Desiderata for Controlled Medical Vocabularies in the Twenty-First Century. *Methods Inf Med*. 1998;37(4-5):394-403.
20. Stausberg J, Lehmann N, Kaczmarek D, Stein M. Reliability of diagnoses coding with ICD-10. *Int J Med Inf*. 2008;77(1):50-57. doi:10.1016/j.ijmedinf.2006.11.005.
21. van der Tuuk K, Koopmans CM, Groen H, Mol BW, van Pampus MG, HYPITAT study group. Impact of the HYPITAT trial on doctors' behaviour and prevalence of eclampsia in the Netherlands. *BJOG Int J Obstet Gynaecol*. 2011;118(13):1658-1660. doi:10.1111/j.1471-0528.2011.03138.x.
22. McMillan TE, Allan W, Black PN. Accuracy of information on medicines in hospital discharge summaries. *Intern Med J*. 2006;36(4):221-225. doi:10.1111/j.1445-5994.2006.01028.x.
23. Jansen JO, Grant IC. Communication with general practitioners after accident and emergency attendance: computer generated letters are often deficient. *Emerg Med J EMJ*. 2003;20(3):256-257.
24. Adhiyaman V, Oke A, White AD, Shah IU. Diagnoses in discharge communications: how far are they reliable? *Int J Clin Pract*. 2000;54(7):457-458.
25. van Zyl LT, Davidson PR. Delirium in hospital: an underreported event at discharge. *Can J Psychiatry Rev Can Psychiatr*. 2003;48(8):555-560.
26. Penney TM. Delayed communication between hospitals and general practitioners: where does the problem lie? *BMJ*. 1988;297(6640):28-29.
27. Chan S, Maurice AP, Pollard CW, Ayre SJ, Walters DL, Ward HE. Improving the efficiency of discharge summary completion by linking to preexisting patient information databases. *BMJ Qual Improv Rep*. 2014;3(1):u200548.w2006. doi:10.1136/bmjquality.u200548.w2006.
28. Kendrick AR, Hindmarsh DJ. Which type of hospital discharge report reaches general practitioners most quickly? *BMJ*. 1989;298(6670):362-363. doi:10.1136/bmj.298.6670.362-a.
29. Williams EI, Fitton F. General practitioner response to elderly patients discharged from hospital. *BMJ*. 1990;300(6718):159-161. doi:10.1136/bmj.300.6718.159.

30. Talwalkar JS, Ouellette JR, Alston S, et al. A Structured Workshop to Improve the Quality of Resident Discharge Summaries. *J Grad Med Educ*. 2012;4(1):87-91. doi:10.4300/JGME-D-10-00249.1.
31. Bischoff K, Goel A, Hollander H, Ranji SR, Mourad M. The Housestaff Incentive Program: improving the timeliness and quality of discharge summaries by engaging residents in quality improvement. *BMJ Qual Saf*. 2013;22(9):768-774. doi:10.1136/bmjqs-2012-001671.
32. Graetz I, Reed M, Rundall T, Bellows J, Brand R, Hsu J. Care Coordination and Electronic Health Records: Connecting Clinicians. *AMIA Annu Symp Proc*. 2009;2009:208-212.
33. Branger PJ, van der Wouden JC, Schudel BR, et al. Electronic communication between providers of primary and secondary care. *BMJ*. 1992;305(6861):1068-1070.
34. Burke HB, Sessums LL, Hoang A, et al. Electronic health records improve clinical note quality. *J Am Med Inform Assoc*. October 2014:amiajnl-2014-002726. doi:10.1136/amiajnl-2014-002726.
35. Tang PC, LaRosa MP, Gorden SM. Use of Computer-based Records, Completeness of Documentation, and Appropriateness of Documented Clinical Decisions. *J Am Med Inform Assoc*. 1999;6(3):245-251. doi:10.1136/jamia.1999.0060245.
36. Rosenbloom ST, Denny JC, Xu H, Lorenzi N, Stead WW, Johnson KB. Data from clinical notes: a perspective on the tension between structure and flexible documentation. *J Am Med Inform Assoc*. 2011;18(2):181-186. doi:10.1136/jamia.2010.007237.
37. Llewelyn DE, Ewins DL, Horn J, Evans TG, McGregor AM. Computerised updating of clinical summaries: new opportunities for clinical practice and research? *BMJ*. 1988;297(6662):1504-1506.
38. Tsopra R, Jais J-P, Venot A, Duclos C. Comparison of two kinds of interface, based on guided navigation or usability principles, for improving the adoption of computerized decision support systems: application to the prescription of antibiotics. *J Am Med Inform Assoc JAMIA*. 2014;21(e1):e107-116. doi:10.1136/amiajnl-2013-002042.

LEGENDS OF FIGURES

Figure 1: 5-step process for evaluating the accuracy of discharge summary diagnoses – For each patient, the reference list of diagnoses was compared with the list of diagnoses recorded in discharge summaries by a third person not involved in the generation of the lists. To allow for an objective comparison, all diagnoses were encoded using a standardized terminology (ICD10).

Figure 2: Comparison of the reference list of diagnoses versus the list of diagnoses recorded in the discharge summaries for each patient. All reference diagnoses were compared to all discharge summary diagnoses using a 5 grade scale to determine the level of accuracy.

Figure 3: Inaccuracy rate per discharge summary. Each vertical line represents one discharge summary. There was a greater than 50% inaccuracy rate in 67% [58 to 76%] of discharge summaries. None were fully accurate.

Figure 4: Correlation of the inaccuracy rate in recorded diagnoses per summary with the type of disease ($p=0.004$). “n” is the number of total diagnoses for the related ICD10 chapter. The inaccuracy rate was highest for infectious diagnoses, symptom and endocrinology diagnoses, and lowest for psychiatry, rheumatology, neoplasms and respiratory diagnoses.