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Survey definitions of gout for epidemiological studies: comparison with crystal identification as the gold standard

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ABSTRACT

Aim: To identify the best performing survey definition of gout from items usually available in epidemiological studies.

Methods: Survey definitions of gout were identified from 34 epidemiological studies contributing to the Global Urate Genetics Consortium (GUGC) genome-wide association study. Data from the Study for Updated Gout Classification Criteria (SUGAR) were randomly divided into development and test datasets. A data-driven case-definition was formed using logistic regression in the development dataset. This definition, along with definitions used in GUGC studies and the 2015 ACR-EULAR Gout Classification criteria were applied to the test dataset, using monosodium urate crystal identification as the gold-standard.

Results: For all tested GUGC definitions, the simple definition of 'self-report of gout or urate-lowering therapy use' had the best test performance characteristics (sensitivity 82%, specificity 72%). The simple definition had similar performance to a SUGAR data-driven case-definition with five weighted items: self-report, self-report of doctor diagnosis, colchicine use, urate-lowering therapy use and hyperuricaemia (sensitivity 87%, specificity 70%). Both of these definitions performed better than the 1977 American Rheumatism Association survey criteria (sensitivity 82%, specificity 67%). Of all tested definitions, the 2015 ACR-EULAR criteria had the best performance (sensitivity 92%, specificity 89%).

Conclusions: A simple definition of 'self-report of gout or urate-lowering therapy use' has the best test performance characteristics of existing definitions that use routinely available data. A more complex combination of features is more sensitive, but still lacks good specificity. If more accurate case-definition is required for a particular study, the 2015 ACR-

EULAR Gout Classification criteria should be considered.

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SIGNIFICANCE AND INNOVATIONS

- Gout epidemiology studies are hampered by the lack of a consistently used survey definition of gout.
- This large international study, using MSU crystal identification as the gold standard, has identified a simple survey definition with good test performance characteristics.
- However, the 2015 ACR-EULAR Gout Classification criteria have substantially better performance characteristics than any tested survey definitions.

INTRODUCTION

Limited information is usually available for the case-definition of gout in epidemiological studies, particularly for multipurpose cohorts. A standard and accurate method of case-definition is important for epidemiological studies, for reasons of efficiency and validity.

However, many different combinations of data available from surveys or multipurpose cohorts have been used to identify gout cases in large population studies (1), and different case definitions of disease can lead to major variation in estimates of disease incidence and prevalence (2, 3). The aim of this study was to construct the best performing case-definition for gout from the limited items available in survey studies and multipurpose cohorts, testing these for accuracy against monosodium urate (MSU) crystal identification as the gold standard.

METHODS

Data from the Study for Updated Gout Classification Criteria (SUGAR) were analyzed. The methods of this study have been described in detail (4, 5); briefly, this was a large multinational cross-sectional study of 983 consecutive rheumatology clinic patients with at least one swollen joint or suspected subcutaneous tophus designed to identify clinical features that accurately distinguish gout from non-gout. At a standardized study visit, clinical features were recorded using case record forms, in addition to independent synovial fluid microscopy by a certified observer. Gout status was defined by synovial fluid or tophus aspirate microscopy result in all patients.

Items and combinations of these items used for definitions of gout in various surveys were identified from 32 studies contributing to the Global Urate Genetics Consortium (GUGC) genome wide association study of hyperuricaemia and gout (6), and were tested in the SUGAR dataset. The GUGC is a large genetic epidemiology study (>140,000 participants of European ancestry), 15 different definitions of gout were used, including the 1977 preliminary American Rheumatism Association survey definition (7) (Supplementary Table 1). Five items for survey definitions of gout were abstracted from the GUGC studies: patient self-report of gout, patient self-report of doctor diagnosis of gout, allopurinol or other urate-lowering therapy use, colchicine use, and self-report of elevated serum urate. These variables were all available in SUGAR with the exception of self-report of elevated serum urate, so actual serum urate level was used instead. Elevated serum urate (hyperuricaemia) was defined as serum urate greater than the upper limit of normal for the local laboratory.

Data from SUGAR were randomly divided into a development data subset (2/3) and test data subset (1/3). Items from the GUGC gout definitions were entered into a logistic regression analysis in the SUGAR development data subset to construct a data-driven case-definition, using MSU crystal defined gout/non-gout status as the dependent variable and backward selection. The score for the data-driven definition was derived from the beta coefficients in this model. The data-driven case-definition and definitions used in the GUGC studies (n=10 definitions with available data in SUGAR, including 7 composite definitions (Supplementary Table 1) were applied to the SUGAR test data subset and the sensitivity and specificity of each definition were calculated. The 2015 American College of Rheumatology (ACR)-European League Against Rheumatism (EULAR) Gout Classification Criteria were also applied to the test data subset (8, 9). Data were analyzed using SPSS v22 (SPSS Inc., Chicago, IL).

RESULTS

Development data subset

In the development data subset, all five items (patient self-report, patient self-report of doctor diagnosis, allopurinol or other urate-lowering therapy use, colchicine use, and elevated serum urate) independently contributed to the regression model (Table 1). Using these data, a score for case definition was derived from the five items: self-report of gout (3 points), self-report of doctor diagnosis of gout (2 points), colchicine use (1 point), urate-lowering therapy use (2 points), hyperuricaemia (3 points) (Table 1). The points were derived from rounding the beta coefficient from the multivariate model to the nearest 0.5 and multiplying by 2. A cut-point of >5 for the data-driven SUGAR survey definition provided maximal sensitivity and specificity according to the receiver operating characteristic curve (Figure 1).

Test data subset

The sensitivity and specificity for the data-driven SUGAR survey definition along with individual items, other definitions from GUGC studies, and the 2015 ACR-EULAR Gout Classification criteria were calculated in the SUGAR test data subset (Table 2). ‘Self-report of gout’ had the best overall performance as a single item (sensitivity 80%, specificity 72%). ‘Use of urate-lowering therapy’ as a single item had high specificity (91%), but very low sensitivity (36%). For all tested GUGC definitions, the simple definition of ‘self-report of gout or urate-lowering therapy use’ had the best test performance characteristics of existing definitions with sensitivity of 82% and specificity of 72%.

The data-driven SUGAR survey definition had a sensitivity of 87% and specificity of 70% in the test data subset. Overall, this performance was similar to the simple definition of 'self-report of gout or urate-lowering therapy use'. The simple definition of 'self-report of gout or urate-lowering therapy use' and the data-driven SUGAR survey definition both performed better than the 1977 American Rheumatism Association survey criteria (sensitivity 82%, specificity 67%). Of all tested definitions, the 2015 ACR-EULAR Gout Classification criteria had the best performance (sensitivity 92%, specificity 89%).

CONCLUSIONS

This analysis has identified that a simple definition of 'self-report of gout or urate-lowering therapy use' has the best, although not without limitations, test performance characteristics of existing survey definitions with sensitivity of 82% and specificity of 72%. Given the design features of SUGAR, the specificity is likely to be an under-estimate of test-performance for population studies and these values are therefore helpful in estimating worst-case misclassification rates from population studies. A more complex combination of features available from routinely collected data is more sensitive, but still lacks very high specificity. Importantly, none of these survey definitions perform as well as the 2015 ACR-EULAR Gout Classification criteria. However, the 2015 ACR-EULAR Gout Classification criteria require a patient interview for typical clinical characteristics of gout, physical examination and laboratory testing, with or without imaging assessment (8, 9). For large multipurpose epidemiological studies, particularly general cohorts not focused on gout or established before the 2015 Gout Classification criteria were published, such detailed information may not be feasible or available, and for this reason survey definitions may be required.

Limitations of this study include recruitment for SUGAR from rheumatology clinics. Patients presenting to secondary care may not be representative of people with gout in a community or general population setting due to disease severity or comorbid conditions. It is also likely that the predictive properties of all definitions will differ in a general population cohort in which the majority of participants do not have gout. In addition, although the specificity of all these case-definitions are likely to be even higher among general population non-gout controls, it is likely that the same order of specificity values we observed in SUGAR would hold true in a general population sample. SUGAR did not collect information about self-report of elevated serum urate, and this variable may have different properties to hyperuricaemia defined by a laboratory test. It is also possible that different serum urate cut-points may alter sensitivity and specificity of a survey definition. This study has a number of strengths. SUGAR is a large, multinational study designed specifically to identify features that classify gout. The case definition of gout using the pathological gold standard of crystal identification is a major strength. The findings of this study are likely to be widely applicable, noting that the items 'self-report of gout or urate-lowering therapy use' are available in many surveys and multipurpose cohorts.

In summary, a simple definition of 'self-report of gout or urate-lowering therapy use' has the best test performance of existing survey definitions for epidemiological gout studies. If more accurate case-definition is required for a particular study, the 2015 ACR-EULAR Gout Classification criteria should be considered.

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REFERENCES

1. Kuo CF, Grainge MJ, Zhang W, Doherty M. Global epidemiology of gout: prevalence, incidence and risk factors. *Nat Rev Rheumatol*. 2015;11:649-62.
2. Bardin T, Bouee S, Clerson P, Chales G, Flipo RM, Liote F, et al. Prevalence of gout in the adult population of France. *Arthritis Care Res (Hoboken)*. 2015; Epub ahead of print.
3. Richette P, Clerson P, Bouee S, Chales G, Doherty M, Flipo RM, et al. Identification of patients with gout: elaboration of a questionnaire for epidemiological studies. *Ann Rheum Dis*. 2015;74:1684-90.
4. Taylor WJ, Fransen J, Dalbeth N, Neogi T, Schumacher HR, Brown M, et al. Performance of classification criteria for gout in early and established disease. *Ann Rheum Dis*. 2014; Epub ahead of print.
5. Taylor WJ, Fransen J, Jansen TL, Dalbeth N, Schumacher HR, Brown M, et al. Study for Updated Gout Classification Criteria (SUGAR): identification of features to classify gout. *Arthritis Care Res (Hoboken)*. 2015; Epub ahead of print.
6. Kottgen A, Albrecht E, Teumer A, Vitart V, Krumsiek J, Hundertmark C, et al. Genome-wide association analyses identify 18 new loci associated with serum urate concentrations. *Nat Genet*. 2013;45:145-54.
7. Wallace SL, Robinson H, Masi AT, Decker JL, McCarty DJ, Yu TF. Preliminary criteria for the classification of the acute arthritis of primary gout. *Arthritis Rheum*. 1977;20:895-900.
8. Neogi T, Jansen TL, Dalbeth N, Fransen J, Schumacher HR, Berendsen D, et al. 2015 Gout classification criteria: an American College of Rheumatology/European League Against Rheumatism collaborative initiative. *Ann Rheum Dis*. 2015;74:1789-98.

9. Neogi T, Jansen TL, Dalbeth N, Fransen J, Schumacher HR, Berendsen D, et al. 2015 Gout Classification Criteria: An American College of Rheumatology/European League Against Rheumatism Collaborative Initiative. *Arthritis Rheumatol.* 2015;67:2557-68.

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Table 1. Regression model of individual survey items using the development data subset and data-driven SUGAR survey definition of gout. For the data-driven SUGAR survey definition of gout, a cut-point >5 points provided optimal sensitivity and specificity in the development data subset. Regression model chi-square 305, df 5, $p < 0.001$, R-squared 0.54.

Item	Odds ratio (95% CI)	B coefficient (SE)	Points
Self-report of gout	4.1 (2.4, 6.8)	1.40 (0.26)	3
Self-report of doctor diagnosis of gout	3.1 (1.8, 5.1)	1.12 (0.26)	2
Hyperuricaemia	5.3 (3.3, 8.4)	1.67 (0.24)	3
Colchicine use	1.6 (1.0, 2.6)	0.49 (0.24)	1
Urate-lowering therapy use	2.2 (1.2, 3.9)	0.77 (0.31)	2

Table 2. Performance of individual items and composite survey definitions in the SUGAR test data subset.

	Sensitivity	Specificity	Youden index†
Single items used in the Global Urate Genetics Consortium study			
Self-report of gout	80%	72%	0.52
Self-report of doctor diagnosis of gout	80%	69%	0.49
Hyperuricaemia	85%	60%	0.45
Colchicine use	48%	76%	0.24
Urate-lowering therapy (ULT) use	36%	91%	0.27
Composite definitions reported in the Global Urate Genetics Consortium study			
Self-report of gout <u>or</u> ULT use	82%	72%	0.54
Hyperuricaemia <u>and</u> ULT use	31%	94%	0.25
Gout specific medications (colchicine or ULT)	61%	72%	0.32
Self-report of gout <u>or</u> gout specific medications	87%	61%	0.48
Self-report of gout <u>and</u> gout specific medications	53%	83%	0.36
Self-report of gout <u>or</u> hyperuricaemia	96%	50%	0.46
1977 preliminary ARA survey criteria	82%	67%	0.49
New composite definitions			
Data-driven SUGAR survey definition	87%	70%	0.57
2015 ACR-EULAR gout classification criteria	92%	89%	0.81

†Youden index = sensitivity + specificity - 1 (perfect test is 1, test no better than chance is 0)

ARA: American Rheumatism Association, ACR-EULAR: American College of Rheumatology-European League Against Rheumatism

FIGURE LEGENDS

Figure 1. Receiver operator characteristic (ROC) curve for the data-driven SUGAR survey definition in the development data subset. AUC (95% CI) for curve 0.83 (0.78, 0.88).

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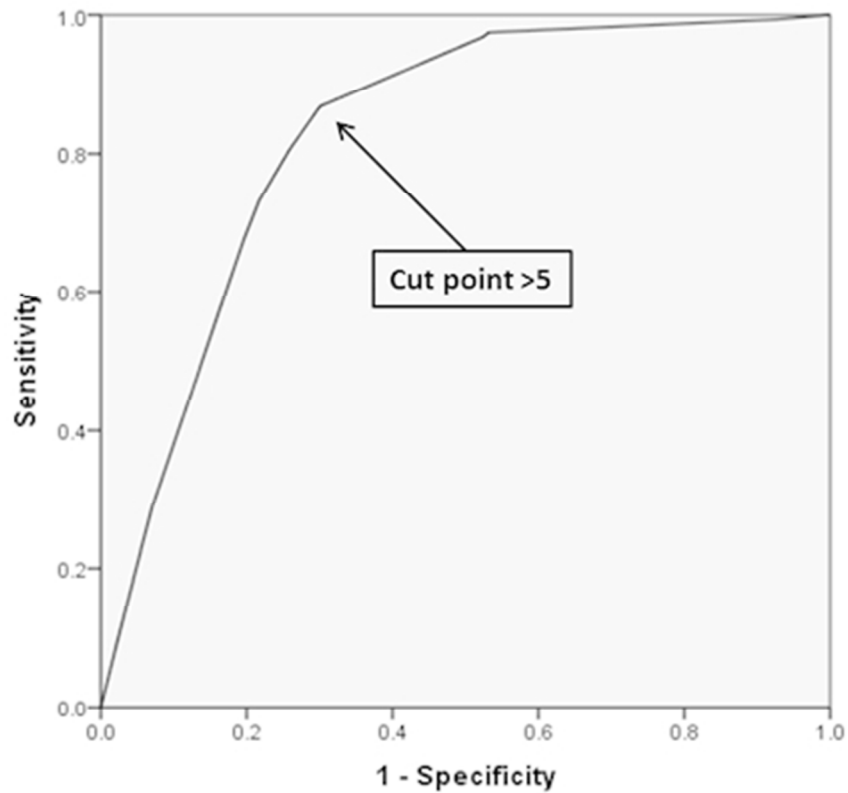


Figure 1. Receiver operator characteristic (ROC) curve for the data-driven SUGAR survey definition in the development data subset. AUC (95% CI) for curve 0.83 (0.78, 0.88).
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