

Prediction of myasthenia gravis crisis events by a machine learning algorithm

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INTRODUCTION

- Myasthenia gravis (MG) is a rare, chronic autoimmune disease of the neuromuscular junction characterized by fatigable muscle weakness.^{1,2}
- Approximately 15% to 20% of patients with MG experience myasthenic crisis, defined as respiratory failure necessitating either noninvasive positive pressure ventilation or mechanical ventilation.²⁻⁴
 - Treatment of MG crises often requires admission to an intensive care unit and usually involves additional acute care and supportive therapy, such as plasmapheresis, intravenous immunoglobulin, and/or corticosteroids.^{2,3}
- MG crisis is more likely to occur early in the disease course, usually within 3 years of diagnosis, and the mortality rate is ~ 5%, primarily resulting from comorbidities.^{2,4}
- Identifying characteristics of patients at high-risk for MG crisis can guide treatment decisions and help avoid critical MG disease progression to MG crisis.⁵

OBJECTIVE

- To identify key patient characteristics, disease symptoms, and comorbidities associated with an increased risk of MG crisis using a machine learning (ML) algorithm.

CONCLUSIONS

- In this novel analysis, the application of an ML algorithm identified several key patient-relevant characteristics, disease symptoms, and comorbidities as risk factors significantly associated with a higher probability of experiencing an MG crisis.
 - Variables related to the respiratory and bulbar domains of the Myasthenia Gravis Activities of Daily Living scale, including cough, shortness of breath, dyspnea, aphasia/speech disturbance, and dysphagia, appeared to be more important for MG crisis prediction than variables related to the ocular and limb domains.
- Collectively, these results suggest that more aggressive treatment strategies with novel medications may be necessary for patients deemed at higher risk of MG crises.

References

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METHODS

- The IQVIA PharMetrics® Plus and Optum Clinformatics® claims databases were retrospectively analyzed (**Figure 1**).

Figure 1. Study design

Retrospective claims database analysis

IQVIA PharMetrics® Plus

Optum Clinformatics® (for external model validation)

Eligibility criteria

- ✓ Aged ≥ 18 years
- ✓ ≥ 2 claims (≥ 30 days apart)* with MG diagnosis ICD-9 or ICD-10 codes filed by a nonophthalmologic specialist
- ✓ Continuous insurance enrollment from 6 months before (baseline period) to 12 months after the first MG diagnostic claim date (index date)

Study period

Baseline: 6 months Follow-up: 12 months

1/1/2015 12/31/2022

MG index date: date of the first MG diagnostic claim

MG crisis^b comparator populations

Patients with MG crisis in the year after MG diagnosis

Patients without MG crisis in the year after MG diagnosis

*Patients who had an MG crisis within 7 days of the first MG diagnostic claim date were excluded. ^bMG crisis was defined as an inpatient claim with a code for acute respiratory failure (ICD-10-CM J96.x; ICD-9-CM 168.81) or a Current Procedural Terminology code for mechanical intubation or ventilation (31500, 94660, 94662).
ICD, International Classification of Disease; MG, myasthenia gravis.

RESULTS AND INTERPRETATION

Patient characteristics

- Of the 205,142,792 patients in the IQVIA PharMetrics® Plus database, 7609 met the study eligibility criteria.
 - 7277 patients had not experienced an MG crisis within 1 year of MG diagnosis, and 332 patients had an MG crisis within that time.
- Certain characteristics (**Table 1**), comorbidities (**Table 2**), and pre-index medications (**Table 3**) were more common in patients with MG crisis than those without.

Variables predicting MG crisis

- Based on the claims data analysis, initial variables selected for the prediction model were
 - Age at MG index date and sex
 - Top 30 baseline symptoms/comorbidities by prevalence and respiratory failure
 - Top 20 pre-index medications by prevalence
 - Number of prior all-cause hospitalizations and the place of service (inpatient vs noninpatient) where the first MG diagnosis occurred
- Significant model-identified MG crisis risk factors were MG diagnosis at inpatient service, baseline presence of aphasia/speech disturbance or dysphagia, older age at diagnosis, baseline presence of shortness of breath or other dyspnea, and baseline presence of cough (**Figure 2**).

Risk factors predicting MG crisis

- Survival analyses confirmed significant associations between select risk factors and the likelihood of MG crises, with higher probabilities observed for patients with multiple risk factors (**Figure 3**).
 - Older patients, patients with aphasia/speech disturbance or dysphagia claims at baseline, and patients with multiple risk factors, such as cough and/or shortness of breath/other dyspnea, have a significantly higher probability of experiencing MG crisis at follow-up.

Figure 2. Odds ratio of variables predicting MG crisis

Variable	Variable importance	Odds ratio (95% CI)	P value
MG diagnosis at inpatient service	1.00	1.96 (1.38, 2.78)	< 0.001
Aphasia/speech disturbance or dysphagia	0.87	1.5 (1.13, 1.99)	< 0.01
Age at MG diagnosis	0.86	1.01 (1.0, 1.02)	0.03
Shortness of breath or other dyspnea	0.68	1.37 (1.01, 1.85)	0.04
Cough	0.66	1.45 (1.04, 2.02)	0.03
Albuterol sulfate	0.56	1.48 (0.99, 2.21)	0.06
Respiratory failure or prior hospitalization	0.53	1.27 (0.94, 1.71)	0.11
Malaise and fatigue	0.53	1.24 (0.94, 1.62)	0.12
Gabapentin	0.50	1.44 (0.95, 2.18)	0.09
Essential hypertension	0.46	1.16 (0.92, 1.47)	0.22

Variable importance was derived based on the magnitude of standardized coefficients. Bolded values indicate significance (P ≤ 0.05).
MG, myasthenia gravis.

Figure 3. Survival analysis stratified by MG crisis risk factors

Age

Strata: Aged 18-49, Aged 50-64, Aged 65+

Probability of survival without MG crisis vs Days since MG index date. P < 0.0001

Aphasia/speech disturbance and other dysphagia

Strata: Aphasia speech disturbance or dysphagia = No, Aphasia speech disturbance or dysphagia = Yes

Probability of survival without MG crisis vs Days since MG index date. P < 0.0001

Cough and shortness of breath/other dyspnea

Strata: Cough = No, Shortness of breath or other dyspnea = No; Cough = No, Shortness of breath or other dyspnea = Yes; Cough = Yes, Shortness of breath or other dyspnea = No; Cough = Yes, Shortness of breath or other dyspnea = Yes

Probability of survival without MG crisis vs Days since MG index date. P < 0.0001

Figure 4. Machine learning model performance

IQVIA PharMetrics® Plus database

ROC curves for models

True positive rate vs False positive rate

Optum Clinformatics® database

ROC curves for models

True positive rate vs False positive rate

For internal model (IQVIA PharMetrics® Plus database) performance assessment, 20 bootstrapping runs were performed. ROC, receiver operating characteristic.

Measure of ML model performance

- Internal validation of the model by ROC curve yielded an area under the curve (AUC) of 0.71 (95% CI: 0.69, 0.72), indicating fair overall predictive performance; external model validation yielded a similarly acceptable AUC (0.65) (**Figure 4**).
 - Factors such as demographic differences, data collection and coding practice differences, variations in healthcare resource utilization patterns, and external validation challenges influenced the AUC difference between the internal and external validation datasets.

Limitations

- The IQVIA PharMetrics® Plus database only contains claims data; electronic medical records data, such as lab measurements, are not available.
- The first MG diagnostic claims date in the IQVIA PharMetrics® Plus database may not be the actual initial MG diagnosis date.
- Mild multicollinearity issues exist because of the nature of the claims data, even after the combination of variables was used; this can influence the interpretation of odds ratios.