

Hina Garg, M.S., PT; Eduard Gappmaier, PT, PhD Department of Physical Therapy, University of Utah, Salt Lake City, UT.

CLINICAL NEUROMUSCULAR RESEARCH LAB

# Background

- Falls and its consequences form a major health concern in persons with Multiple Sclerosis (PwMS).
- Due to its multifactorial nature, it is vital to investigate the prime determinants that distinguish 'high-risk' fallers from those with a lesser risk.

- To identify the with respect to demographics, strength, balance, functional mobility, spasticity, self-report measures and cognition that predict falls in
- resulting significant predictors and identify cut-offs for their predictive accuracy of falls in PwMS.

# Objectives

- significant contributors
- To compare the

### • Sample of convenience (n=55). • Criteria for inclusion: Clinically definite MS. Criteria for exclusion: Recent or current acute exacerbation.

Table	1: Sample	e descriptive	[Mean	(Range

fall history.

		L \	/ ]		
Variable	Total sample (n=55)	Non-fallers (n=17)	Fallen at least once (n=34)	Recurrent fallers (n=25)	Others (0 or 1 fall, n=26)
Age (yrs)	56.2 (23-81)	57.5 (31-81)	56.7 (23-77)	56.6 (23-77)	57.3 (31-81)
Gender (F/M)	38/17	12/5	23/11	15/10	20/6
Height (cm)	168 (147-191)	169 (153-191)	167 (147-185)	168 (147-185)	167 (152-191)
Weight (kg)	78 (49-135)	74 (50-106)	77 (49-135)	77 (51-100)	76 (49-135)
EDSS	5.3 (3-8)	4.6 (3-8)	5.6 (3.5-7.5)	5.7 (3.5-7.5)	4.9 (3-8)
Time since diagnosis	14.3 (1-41)	11.6 (3-34)	16.9 (1-41)	15.5 (1-39)	14.8 (3-41)
ABC (%)	56 (0-97.8)	66.9 (6-94.7)	50.9 (4.7-97.8)	50.7 (4.7-97.8)	61.5 (5-94.7)
ASH-ave	0.55 (0-3)	0.35 (0-3)	0.65 (0-3)	0.76 (0-3)	0.35 (0-3)
SDMT	42.4 (11-63)	44.7 (32-55)	41.3 (11-63)	39.9 (11-60)	44.8 (21-63)

Pilot, cross-sectional design with retrospective

# Instruments and Measures

### The outcome (or the dependent) variable

- For multiple linear regression analysis, the number of falls recalled in the past year was used (falls/yr).
- For hierarchical binary logistic analysis, two separate dichotomous variables comparing all falls versus no fall events and recurrent falls (2 or more falls) versus others (0 or 1 fall) were created and utilized.

#### The predictor (or the independent) variables: Seven covariates based on existing literature were identified.

- Demographics included age, gender and EDSS scores.
- Strength determined by the average of maximal isometric knee extensor strength for both extremities.
- · Clinical balance measures included berg balance test and functional reach Functional mobility measures included timed-up
- and-go, twenty-five feet walk test, stairs test and six-minute walk distance
- Spasticity assessed by the average of modified ashworth scale for both ankle plantarflexors (ASH-ave). Self-report measures included the multiple sclerosis
- walking scale, the Activities-specific Balance Confidence scale (ABC), and the modified fatigue impact scale.
- Cognitive status measured by paced auditory serial addition test and Symbol Digit Modalities Test (SDMT).

The uncertainty regarding the relationship of these variables to falling<sup>1,2</sup> led to no 'a priori' hypothesis.

#### Identification of significant Predictors: A stepwise multiple regression analysis was performed to select a model that predicts the maximum number of

falls in PwMS from the predictor variables. Predictive accuracy: Then, a hierarchical binary logistic regression model was used to examine the effectiveness of the identified predictors to

predict people who fell in the past year. These models either compared all falls versus no reported falls or those with 2 or more (recurrent) falls versus

none or 1 fall. · Cut-off determination: The predictors were, thereafter, analyzed using hierarchical logistic regression for improvements in prediction of falls in different cut-off combinations for ABC. The cut-offs which maximized the sensitivity, specificity and the overall model's predictive value were

SPSS version 20.0 was used for analysis. For regression, missing values

Of fifty-five PwMS, 34 (61.8%) reported 156 falls. Of the fallers, 25 (73.5%) reported recurrent falls (>1 fall). In combination, all the predictor variables accounted for 72.3% of the variance in the number of falls.

### Identification of predictors

ABC, ASH-ave and SDMT, in the respective order, were found to be the prime predictors, R<sup>2</sup> =0.612, F (3, 34) = 17.86, p<0.0001, indicating that PwMS who have higher self-reported balance confidence, low levels of spasticity and higher cognitive function tend to have an overall lower number of falls. The assumptions for normality, independence and linearity (except homoscedasticity) were met.

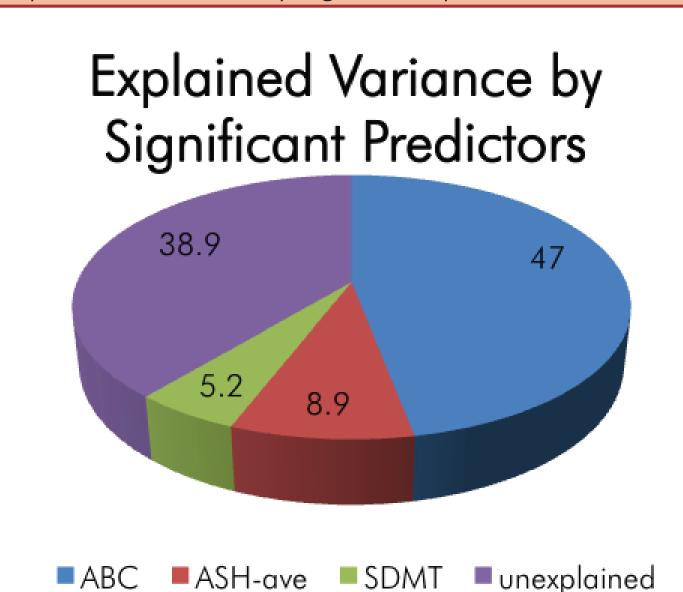
 The regression equation: Predicted score (number of falls) = 12.531 - 0.096 (ABC)score) + 2.523 (ASH-ave) - 0.091 (SDMT score).v

### Table 2: Multiple Linear Regression Model

Model	R	R square	R square change	F change	df1	df2	sig. F change	Durbin Watson	
1	0.686	0.470	0.47	31.96	1	36	0.000	1.65	
2	0.748	0.560	0.089	7.11	1	35	0.012		
3	0.782	0.612	0.052	4.56	1	34	0.040		

Model 1 Predictors: ABC Model 2 Predictors: ABC, ASH-ave Model 3 Predictors: ABC, ASH-ave, SDMT

Figure 1: Explained variance by significant predictors

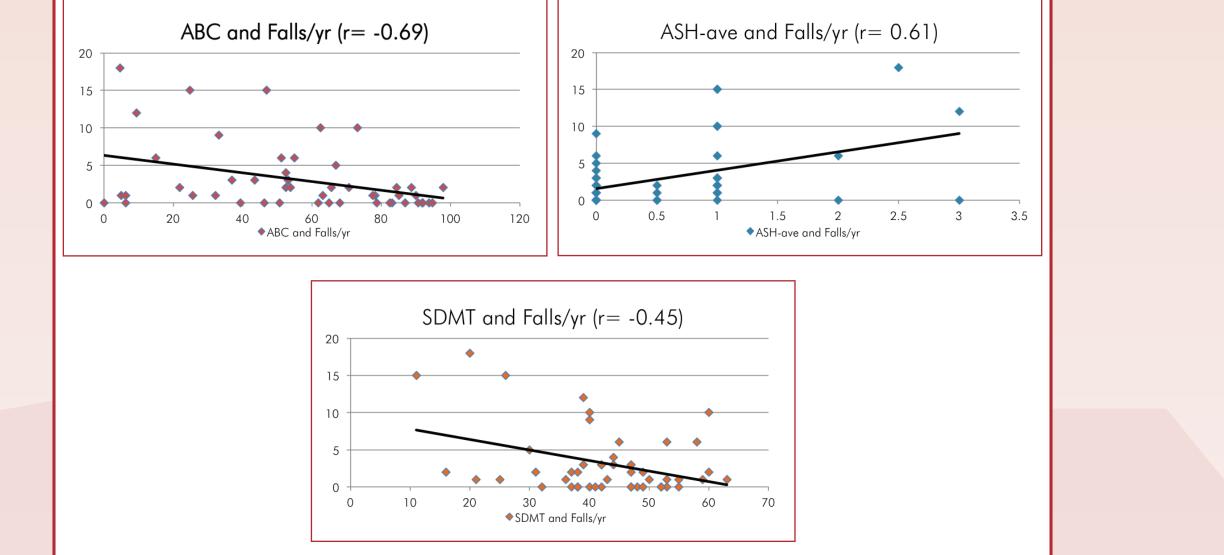


### Probability (Predictive accuracy) of model containing ABC, ASH-ave and SDMT

Hierarchical Logistic Regression Models:

Fallers vs. Non-fallers The combined model of all variables was significant and demonstrated that the probability of determining a faller from a non-faller was higher in people with low self-reported balance (ABC), higher spasticity (ASH-ave) and low cognition (SDMT), with ABC found to be the most significant predictor. Interestingly, similar predictive accuracy (78.7% accurate lassification of fallers) and explained variance (Nagelkerke R2=0.229) were found in the combined model of SDMT, ASH-ave and ABC, as opposed to a model with ABC as the lone predictor (78.3% accurate and R2=0.264). A cut-off measure of 60 for ABC in the combined model resulted in improvements in specificity & model accuracy. This highlights the ability of a self-report balance measure such as the ABC to accurately distinguish PwMS with 'high-risk'

#### Figure 2: Correlations of variables identified



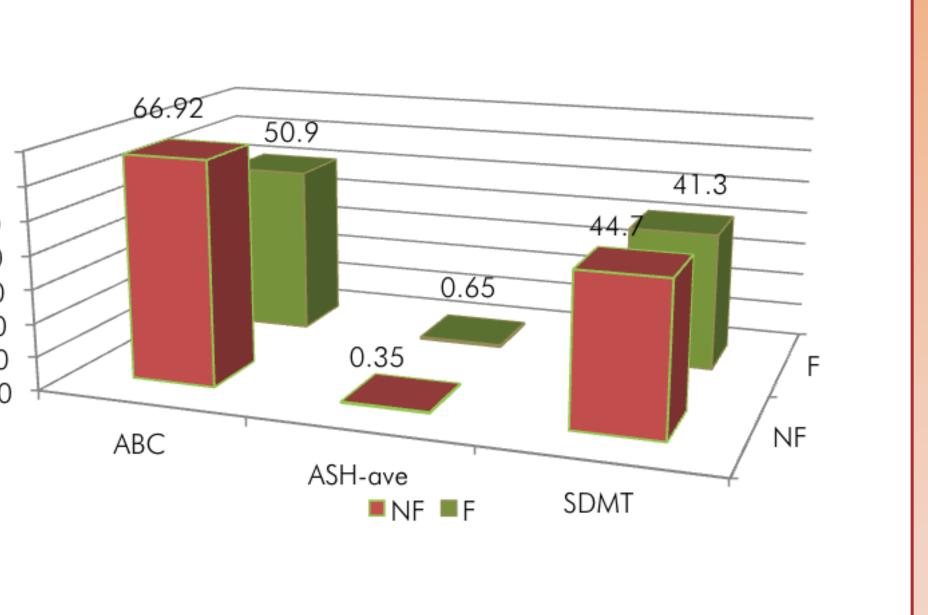
#### Hierarchical Logistic Regression Models: Recurrent Fallers vs. Others (0 or 1 fall)

• The combined model approached significance (p=0.06) and demonstrated that the probability of determining a recurrent faller from others was higher in people with low cognition (SDMT), higher spasticity (ASH-ave) and low self-reported balance (ABC). Similar predictive accuracy (70.2%) and explained variance (Nagelkerke R2=0.191) were again noted in the combined model as compared to a model with ABC as the lone predictor (71.7% and R2=0.123) indicating the importance of the self-report balance measure. Importantly, a cut-off measure of 60 for ABC in the combined model resulted in a significant model and improvements in sensitivity & model accuracy.

### **Table 3:** HLR models – Fallers vs. Non-fallers (n=47)

Model	Predictors	Coetticients (SE)	Wald's □2 (p)	OR (95%CI)	Overall model =2 (p)	Sensitivity (Sn), Specificity (Sp) (%)	Predictive accuracy (%)	False positive/ False negative rates (%)
1	SDMT	-0.016 (0.03)	0.24 (0.63)	0.98 (0.92-1.05)	8.38 (0.039)	90.6, 53.3	78.7	19.4, 27.3
	ASH-ave	0.243 (0.67)	0.13 (0.71)	1.27 (0.35-4.71)				
	ABC	-0.035 (0.02)	3.95 (0.04)	0.97 (0.93-1.00)				
2	SDMT	-0.011 (0.03)	0.12 (0.73)	0.99 (0.93-1.05)	8.33 (0.04)	87.5, 66.7	80.9	15.1, 28.5
	ASH-ave	0.382 (0.63)	0.37 (0.54)	1.46 (0.43-5.04)				
	ABC (60 cut-off)	-1.64 (0.8)	4.18 (0.04)	0.19 (0.04-0.93)				
3	ABC	-0.045 (0.02)	6.63 (0.01)	0.96 (0.92-0.99)	9.51 (0.002)	87.5, 57.1	78.3	17.6, 33.3

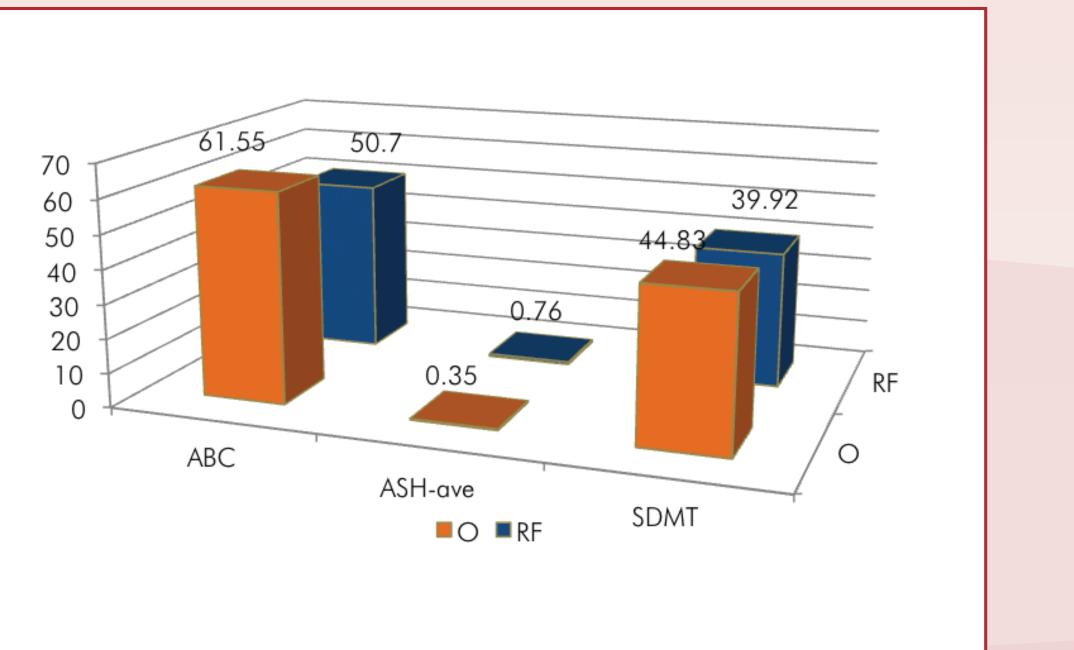
#### Figure 3: Mean Differences: Fallers vs. Non-Fallers

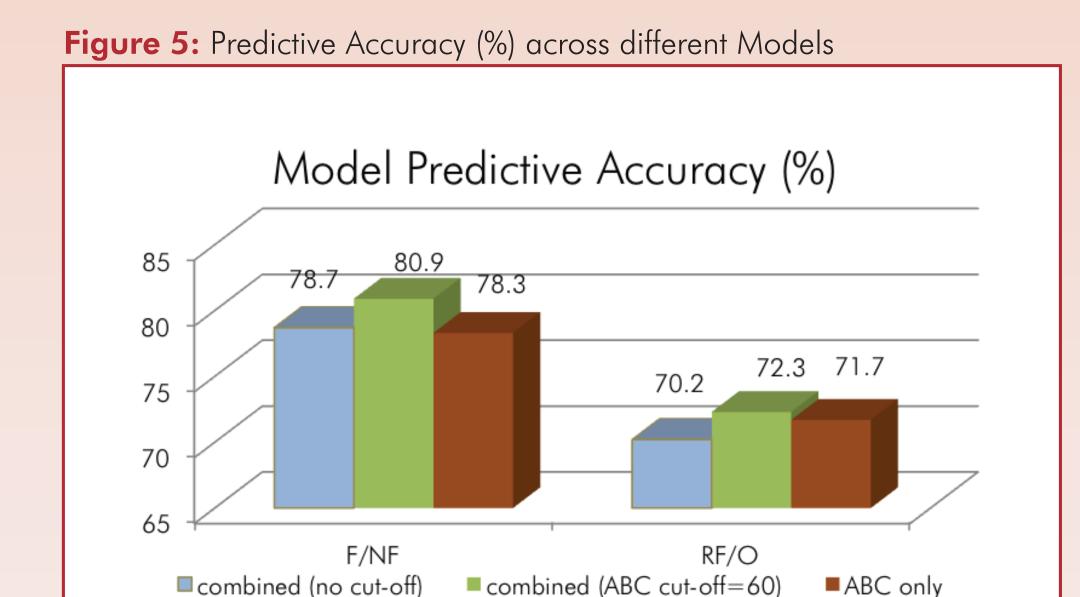


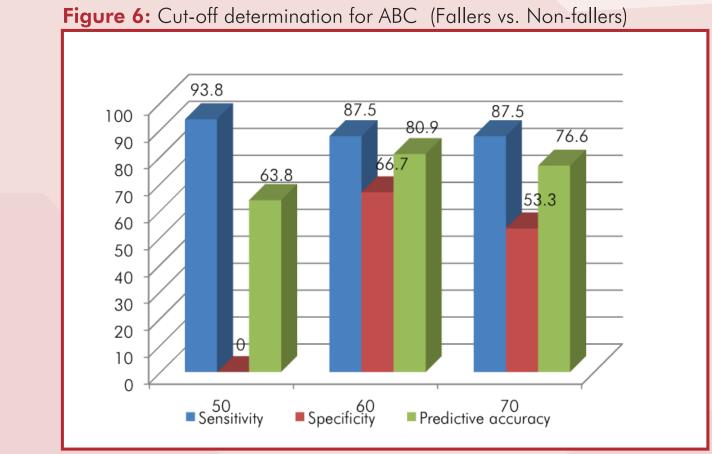
#### **Table 4:** HLR models – Recurrent Fallers vs. Others (0 or 1 fall, n=47)

Model	Predictors	Coefficients (SE)	Wald's □2 (p)	OR (95%CI)	Overall model □2 (p)	Sensitivity (Sn), Specificity (Sp) (%)	Predictive accuracy (%)	False positive/ False negative rates (%)
1	SDMT	-0.029 (0.03)	1.02 (0.311)	0.97 (0.92-1.03)	7.25 (0.064)	60.9, 79.2	70.2	26.3, 32.1
	ASH-ave	0.724 (0.55)	1.72 (0.190)	2.06 (0.70-6.08)				
	ABC	-0.012 (0.01)	0.77 (0.38)	0.99 (0.96-1.01)				
2	SDMT	-0.025 (0.03)	0.672 (0.41)	0.98 (0.92-1.03)	10.15 (0.02)	69.6, 75	72.3	27.2, 28
	ASH-ave	0.586 (0.53)	1.21 (0.27)	1.8 (0.63-5.1)				
	ABC (60 cut-off)	-1.307 (0.69)	3.56 (0.06)	0.27 (0.07-1.05)				
3	ABC	-0.024 (0.01)	4.01 (0.045)	0.98 (0.95-0.99)	4.47 (0.035)	69.6, 73.9	71.7	27.3, 29.2

#### Figure 4: Mean Differences: Recurrent Fallers vs. Others (0 or 1 fall)







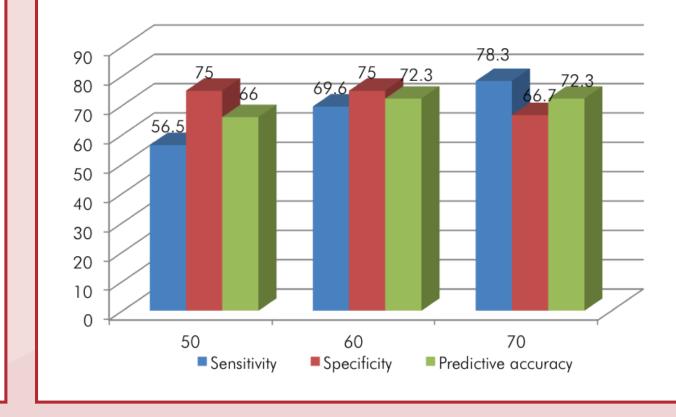


Figure 7: Cut-off determination for ABC (Recurrent fallers vs. Others)

# Limitations

- Cross-sectional design with retrospective fall history.
- Small sample size.
- Selection of a convenience sample.

## Conclusions

- Concurrent with previous literature, our sample depicted a high incidence (61.8%) of falls in PwMS, with a reported recurrence of 73.5%.
- Poor self-reported balance confidence, higher levels of spasticity and poor cognition resulted in a significantly higher number of falls in
- Probability of determining a faller from a non-faller was significantly higher in people with low self-reported balance, higher spasticity and low cognition. Although not significant, a similar trend was noted in the recurrent fallers.
- The cut-off score of 60 on ABC achieved model significance and improvements in sensitivity, specificity and predictive accuracy.
- Since the self-reported balance confidence emerged as a substantial predictor of falls in PwMS, a cut-off score of 60 on ABC has been suggested to enhance model prediction.
- Thus these factors, respectively, should be considered vital for fall-risk screening and fall-intervention programs.

# References

- . Finlayson, M.L., Peterson, E.W., Cho, C.C. (2006). Risk factors for falling among people aged 45 to 90 years with multiple sclerosis. Arch Phys Med Rehabil, 87, 1274-9.
- 2. Nilsagård, Y., Lundholm, C., Denison, E., Gunnarsson, L-G. (2009). Predicting accidental falls in people with multiple sclerosis - a longitudinal study. Clin Rehabil, 23,