**Additional File 4: Summary of Limitations of PHMRC Data**

A Critique of the PHMRC data used for assessments of automated VA COD classification methods

* Quality-related issues with the data:
	+ Data inconsistencies/contradictories:
		- About 66% of deaths had respiratory symptoms identified, yet only 18% of the deaths were respiratory-related (Byass, 2014)
		- 147 stillborn deaths described as neonatal deaths in VA interviews (Byass, 2014)
		- 20% of maternal deaths not described as being pregnant or recently delivered (Byass, 2014)
		- All hierarchical algorithms assessed in Kalter et al. (2016) performed poorly when classifying ‘Measles’ for neonatal and child deaths due to lack of responses for rash, which was only reported in 3 of the 23 reference standard measles cases
	+ Serina et al. (2016) found question reliability to be low and high response incompleteness between the initial and repeat interviews; though concluding that it does not affect Tariff 2.0 (moderate 0.61 kappa), this could potentially affect the performance of other data-driven COD classification algorithms that require training in order to classify records
	+ Cause list categories and definitions:
		- Categories are arbitrary (grouping of many causes in ‘Other cause’) (Garenne, 2014)
		- Cause definitions differ from WHO ICD-10 definitions (Byass, 2014)
		- Not all algorithms produce an ill-defined or indeterminate category which captures the deceased records that may not be complete or accurate enough, or too complex, or else, have a pathology that could not be identified via interviews for COD classification
	+ Different epidemiologic profiles captured:
		- Potentially excludes certain causes (causes that are unlikely to result in death due to receiving treatment) that are common in community settings (Garenne, 2014)
		- Different symptomatology and pathogen distributions captured, which may not be generalizable for rural or medically unattended deaths (Miasnikof et al., 2015; Byass, 2014; Garenne, 2014)
* Use has led to some implausible results:
	+ Studies that report on the comparison of performance between algorithms and physician COD classification (PCVA) are unclear on the number of physicians involved for COD classification
		- Often assumed that single physician COD assignment was used as opposed to the standard practice of using dual physician review
		- Examples of papers lacking clarity or mentioning single physician COD assignment include: Murray et al., 2014; Flaxman et al., 2011; James et al., 2011; Lozano et al., 2011
	+ Garenne (2014) had the following criticisms with regards to the Murray et al. (2014) findings:
		- The values for certain physician diagnoses were abnormally low
			* Sensitivity was only 41% for congenital defects, 39% for measles, 45% for accidental falls and 62% for maternal deaths
		- The sensitivity and positive predictive values (PPVs) for the algorithms had abnormally low values for specific diseases that are ‘easy to diagnose’
			* For example, algorithms and physician sensitivity values for the experiments without healthcare experience variables were less than 60% for prematurity, less than 40% for child diarrhoea and less than 50% for adult epilepsy or asthma
* Problems of external validity:
	+ Limited in time and geographic scope
		- Deaths occurred in hospital: population covered are those who access health services, and so, responses will be different from respondents who did not access services/where deaths were unattended (Kalter et al., 2016; McCormick et al., 2016; Murray et al., 2014)
		- Not suitable for training and testing on more current community VAIs (post-2012) because prevalence rates of diseases change considerably over time and area (McCormick et al., 2016)
			* PHMRC data was collected from 2007 to 2010 within certain sites: Andhra Pradesh and Uttar Pradesh in India, Distrito Federal in Mexico, Bohol in the Philippines, and Dar es Salaam and Pemba (North and South) in Tanzania

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