

Residence, Living Situation, and Living Conditions Information Documentation in Clinical Practice

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Abstract

Social determinants of health (SDOH) have an important role in diagnosis, prevention, health outcomes, and quality of life. Currently, SDOH information in electronic health record (EHR) systems is often contained in unstructured text. The objective of this study is to examine an important subset of SDOH documentation for Residence, Living Situation and Living Conditions in an enterprise EHR informed by previous model representations. In addition to two publically available clinical note sources, notes created by Social Work, Physical Therapy, and Occupational Therapy, along with free text Social Documentation entries were reviewed. Sentences were classified, annotated, and evaluated once mapped to element entities and attributes. Overall, 2,491 total notes yielded 616, 813, and 30 sentences related to Residence, Living Situation, and Living Conditions. This study demonstrated the need for additional elements in the model representation, more representative values and content culminating in a more comprehensive model representation for these key SDOH.

Introduction and Background

Social and individual behavioral factors play an important role in diagnosis, prevention, health outcomes, and quality of life.^{1,2} As defined by the World Health Organization, “social determinants of health are the conditions in which people are born, grow, live, work and age”.³ Social determinants of health (SDOH) can cause illness, exacerbate or contribute to chronic illness, and conversely can also improve health.

Previous studies have demonstrated the deleterious effect of behaviors such as alcohol and tobacco use on health outcomes.⁴⁻⁷ Housing has also been relatively well studied, especially the impact of homelessness on various conditions.⁸⁻¹⁴ Other housing conditions have been correlated with health outcomes. Residential status, specifically housing instability, has been studied in relation to outcomes in certain disease or treatment groups, and found to be a risk for poor outcomes.⁸⁻¹⁰ It overall appears that there is a complex interconnectedness between poor housing and poor health.¹⁵ Costa-Font found that owning a home, or housing equity overrides the effect of income as a determinant of health and (absences) of disability in old age.¹² In contrast, permanent supportive housing can address homelessness and health disparities.¹⁶ This is also some evidence that improving housing can contribute to improved health.¹⁷ For example, Jacobs et al. found evidence that specific housing interventions can improve certain health outcomes.¹⁸

However, other aspects of social determinants related to *Residence*, *Living Situation*, and *Living Conditions* have not been investigated as thoroughly. For example, the impact of housing type (*Residence*) such as single family home, assisted living, group living situations, has not been investigated. Studies have also shown that certain housing models can be beneficial to specific patient groups.^{14,19} In addition, significant exposures risks have been associated with indoor environments.^{20,21} For example, with multi-unit dwellings there is risk of exposure to hazards such as second-hand smoke as smoke can seep into neighboring units resulting in involuntary exposure.²² Knowledge of a patient’s physical living space, type of dwelling, stairs, safety mechanisms, etc. could be of benefit to the clinician or therapist in providing care to that patient and obtaining the proper support in further promoting their wellness.

With whom a patient lives (*Living Situation*) as well as the conditions (*Living Conditions*) under which they live also have health implications. While living with others creates a support network for the patient, housing density increases exposure to communicable diseases, causes stress in adults and poor long-term health in both children and adults.²³

The increase in the use of EHRs provides unprecedented opportunity to collect and analyze SDOH information in conjunction with clinical data in secondary use for research and process improvement. SDOH information can be

used in a myriad of ways from health outcomes evaluations to predictive modeling for prevention. The National Academy of Medicine has completed a consensus study on social determinants in the EHR.^{24, 25} The Committee on Recommended Social and Behavioral Domains and Measures for Electronic Health Records has identified domains and measures that capture the social determinants of health to inform the development of recommendations for Stage 3 meaningful use of electronic health records (EHRs). The NAM final report recommended documentation of race and ethnicity, education, financial resource strain, stress, depression, physical activity, tobacco use and exposure, alcohol use, social connections and social isolation, exposure to violence (intimate partner violence), and neighborhood and community compositional characteristics. The final recommendations unfortunately did not include *Residence, Living Situation, or Living Conditions*.

Currently, comprehensive documentation standards for many SDOH do not exist.²⁶ As a result, EHR systems and their associated user interfaces are not optimized for the consistent collection of discrete social history information leaving this important information often buried in free text notes or as unstructured text fields in the social history sections of the EHR. Natural language processing techniques are being developed that allow us to extract social history information from the notes and into discrete datasets but primarily only around substance use information which has more developed and robust information models providing a “target” for discrete representation.^{27, 28}

The overall goal of this study is to evaluate documentation of three SDOH topic areas (*Residence, Living Situation, and Living Conditions*) using publically available note sources and Enterprise EHR free text documentation. This work also builds from work of Chen *et al.* and Melton *et al.* to model social history from progress notes and public health surveys, including living situation, residence, social support, and occupation.²⁹ Other previous work to harmonize interface terminologies, standards, specifications, coding terminologies, vocabularies, documentation guidelines, measures, and surveys provides a model representation of our three topic areas.²⁶ This study also further refines the model representation of *Residence, Living Situation, and Living Conditions* informed by additional interdisciplinary EHR system content.

Methods

This study is focused on the three topic areas *Residence, Living Situation, and Living Conditions*. Clinical notes from three sources were evaluated. Individual sentences related to the topic areas were identified and were classified, into one or more of the target topic areas. Lastly, statement-level annotation was performed, with a 10% secondary review for Kappa, and elements were mapped to generate a model representation (Figure 1).

Topic areas: Using definitions developed from previous work, *Residence* describes dwelling types, physical residence, and geographic location and includes safety considerations such as railings or number of floors and steps.²⁶ *Living Situation* describes with whom the patient lives such as roommates, family members, multi-resident dwelling as well as how many others they live with. Lastly, *Living Conditions* describes environmental cleanliness and precautions against infection and disease and includes such things as animals, and presence of mold or an unclean living space

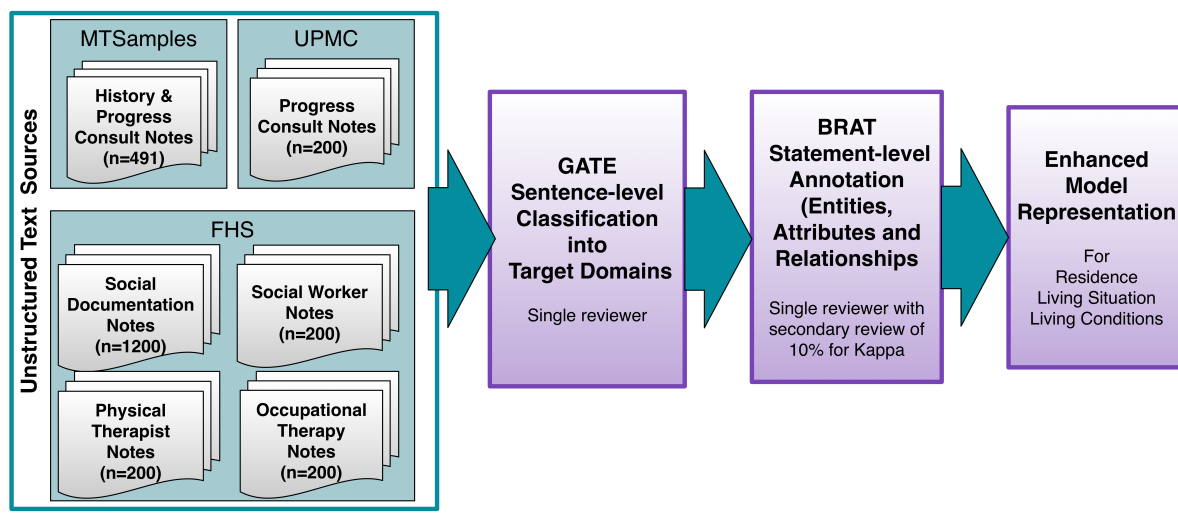


Figure 1. Overview of Methods.

Data Sources: The data sources utilized for this work were: 1) MTSamples.com (MTS), a publicly accessible clinical note data source; 2) University of Pittsburgh Medical Center (UPMC) NLP Repository with de-identified clinical notes (following execution of a data use agreement for research); and 3) multiple sources of data from the University of Minnesota-affiliated Fairview Health Services (FHS) electronic health record system.³⁰ Overall, there were 491 history and physical and consult notes from MTS and 200 history and physical and consult notes from the UPMC included.³⁰ The majority of clinical data for the study otherwise originated from the FHS EHR system through the University of Minnesota Academic Health Center Information Exchange (AHC-IE) data repository. We included only patients who had consented for their medical records to be used in research with inpatient encounters in 2013. For the purposes of this study to also obtain a broader understanding of documentation of this information by non-provider clinicians including several inter-disciplinary fields, we randomly selected 200 random progress notes authored by social workers, 200 progress notes authored by physical therapists, and 200 progress notes authored by occupational therapists, as well as 1,200 social documentation notes. Social Documentation is a portion of the Social History section in the EHR composed of a single free text field that can be documented on by any EHR clinical user.

Annotation Guidelines Development: Guidelines for sentence-level annotation were developed through literature review and included 28 classifications covering most social determinants of health. Related to this work, of those 28, 7 classifications were further analyzed for this study which included: Residence, Residence Exposure, Residence Other, Living Condition, Living Situation, Living Situation Exposure, and Living Situation Other.

Guidelines for statement-level annotations were developed through previous work reviewing existing standards and terminologies.²⁶ Separate schemas were developed for each of the three topic areas (Tables 1, 2 and 3).

Table 1. Residence Annotation Guidelines Entities, Attributes and Relationships.

Entities (27)	Attributes (16)
<ul style="list-style-type: none"> • Status • Subject <ul style="list-style-type: none"> ○ Family member ○ Side of family ○ Other • Negation • Certainty • Temporal <ul style="list-style-type: none"> ○ Start Date ○ End Date ○ Start age ○ End age ○ Duration ○ Duration Since Time Point ○ Time point ○ Time frame ○ Residence Age ○ Residence Build Time Point • Residence Type <ul style="list-style-type: none"> ○ Residence Subtype (Type Details) • Residence Name • Geographic Location <ul style="list-style-type: none"> ○ Location Detail • Residence Detail <ul style="list-style-type: none"> ○ Detail Subtype (Type details) • Quantity • Other 	<ul style="list-style-type: none"> • Specificity <ul style="list-style-type: none"> ○ Exact ○ InexactQuantitative ○ InexactQualitative ○ Other • Location <ul style="list-style-type: none"> ○ City ○ State ○ County ○ Country ○ Other • Certainty <ul style="list-style-type: none"> ○ Unknown ○ Uncertain ○ Certain ○ Other • Side of family <ul style="list-style-type: none"> ○ Paternal ○ Maternal ○ Both paternal and maternal ○ Unknown ○ Other
	Relationships
	<ul style="list-style-type: none"> • Type Relationships • Status Relationships • Amount Relationships • Negation, Certainty, and Context Relationships

Table 2. Living Situation Annotation Guidelines Entities, Attributes and Relationships.

Entities (21)	Attributes
<ul style="list-style-type: none"> • Status • Subject <ul style="list-style-type: none"> ○ Family member ○ Side of family ○ Other • Negation • Certainty • Temporal <ul style="list-style-type: none"> ○ Start Date ○ End Date ○ Start age ○ End age ○ Duration ○ Duration Since Time Point ○ Time point ○ Time frame • Quantity • Current age • Living Situation Detail <ul style="list-style-type: none"> ○ Detail Subtype (Type details) • Other 	<ul style="list-style-type: none"> • Specificity <ul style="list-style-type: none"> ○ Exact ○ InexactQuantitative ○ InexactQualitative ○ Other • Certainty <ul style="list-style-type: none"> ○ Unknown ○ Uncertain ○ Certain ○ Other • Side of family <ul style="list-style-type: none"> ○ Paternal ○ Maternal ○ Both paternal and maternal ○ Unknown ○ Other
	Relationships
	<ul style="list-style-type: none"> • Type Relationships • Status Relationships • Amount Relationships • Negation, Certainty, and Context Relationships to all entities

Table 3. Living Condition Annotation Guidelines Entities, Attributes and Relationships.

Entities (21)	Attributes (13)
<ul style="list-style-type: none"> • Status • Subject <ul style="list-style-type: none"> ○ Family Member ○ Side of Family ○ Other • Negation • Certainty • Temporal <ul style="list-style-type: none"> ○ Start Date ○ End Date ○ Start Age ○ End Age ○ Duration ○ Duration Since Time Point ○ Time Point • Quantity • Living Conditions Type <ul style="list-style-type: none"> ○ Type Subtype • Living Conditions Detail <ul style="list-style-type: none"> ○ Detail Subtype • Other 	<ul style="list-style-type: none"> • Specificity <ul style="list-style-type: none"> ○ Exact ○ InexactQuantitative ○ InexactQualitative ○ Other • Certainty <ul style="list-style-type: none"> ○ Unknown ○ Uncertain ○ Certain ○ Other • Side of Family <ul style="list-style-type: none"> ○ Paternal ○ Maternal ○ Both Paternal and Maternal ○ Unknown ○ Other
	Relationships (6)
	<ul style="list-style-type: none"> • Type Relationships • Status Relationships • Amount Relationships • Negation, Certainty, and Context Relationships to all entities

Sentence-level Annotation

All notes were initially reviewed and sentence-level annotation was performed by a single reviewer using General Architecture for Text Engineering (GATE) to identify and classify sentences related to the three topic areas of *Residence*, *Living Situation*, and *Living Conditions*.³¹ Sentences containing information related to more than one of the three topic areas were classified into each appropriate topic for statement-level annotation. For example, the sentence “He lives in <city name> with his mother and step-father” was classified as both Residence (“lives in <city name>”) and Living Situation (“lives...with his mother and step-father”). Table 4 shows example statements for each topic area.

Table 4. Example sentences classified into each topic area.

Topic Area	Example Sentences
Residence	<ul style="list-style-type: none"> • Lives in 4 level house • Lives in <city, state name> • The patient has been residing at <name of facility>

	<ul style="list-style-type: none"> • Home age 10-25 years
Living Situation	<ul style="list-style-type: none"> • Lives at home with her husband and two daughters • Lives at home with husband and 5 children. • Lives with his parents and 11 year old sister
Living Conditions	<ul style="list-style-type: none"> • They have city water • Childs home has well water • ...had some mold behind the stove and refrigerator

Statement-level Annotations

Statement-level annotation was performed on the classified sentences using the brat rapid annotation tool (brat) to identify elements, attributes and relationships.³² The schema was modified iteratively to accommodate newly found elements and attributes for three sustentative iterations to include subject and temporal entities not previously encountered to ensure a stabilized schema. The original Residence annotation schema was amended to include “Subject”, “Time Point”, and “Time Frame”. The Living Situation schema was amended to include one new element “Current Age” that refers to the age of the persons with whom the patient lives. And the Living Conditions schema was amended to include the element “Living Conditions Detail Subtype” which refers to the subcategory of type.

The statement-level brat annotations was performed by a single reviewer and a subset of 10% of sentences were annotated by a second reviewer to ensure internal consistency and to assess inter-rater reliability. Values sets were compiled from the annotations for each entity found in the data sources as were schema amendments. The model representations from previous work²⁶ were then amended with additional elements from this analysis to create and enhanced model representation of *Residence*, *Living Situation*, and *Living Conditions* (Tables 6A, B, C).

Results

In total, 2,491 notes were reviewed by two reviewers, resulting in 1,459 sentences classified into the three topic areas of *Residence*, *Living Situation*, and *Living Conditions*. The initial classification analysis resulted in 616 sentences categorized as *Residence*, 813 sentences categorized as *Living Situation*, and 30 sentences categorized as *Living Conditions* (Table 5). MTSamples, FHS Physical Therapy, and FHS Occupational Therapy notes did not have any sentences that could be classified under the *Living Conditions* topic.

Table 5. Number of notes reviewed and number of sentences classified. (*Total number of sentences are not mutually exclusive)

Data Source	Total Notes	Sentences Classified				
		Residence	Living Situation	Living Conditions	Total Sentences*	
MT Samples	491	36	88	0	124	
UPMC	200	42	54	5	101	
FHS	Social History Documentation Notes	1200	296	453	24	773
	Social Worker Notes	200	88	64	1	153
	Physical Therapist Notes	200	98	86	0	184
	Occupational Therapist Notes	200	56	68	0	124
Total sentences reviewed		2491				
Total number of sentences classified*		616	813	30	1459	

The statement-level annotation yielded an inter-rater reliability of $K = 0.84\%$ and proportion agreement of 0.98%. Overall the FHS set of notes were the most comprehensive and this had the highest contribution to this work. In totality, the 616 sentences or statements for *Residence* yielded significant contributions to the overall model. Of these sentences, *Status* was documented in 60.1% (416) of sentences, *Residence Type* 51.8% (359), and Geographic Location Detail (i.e., specific city, state country locations) in 38.4% (302) (Table 6A). Temporal elements were present but to a much lower degree.

For *Living Situation*, a total of 813 sentences were analyzed and, as with *Residence*, *Status* was highly prevalent being present 823 times and in total there were 1303 references to Subject other than patient or family member (Table 6B).

For *Living Conditions* in the 30 sentences were annotated, Living Condition Type was the most prevalent entity found with 39 instances with many sentences referencing more than one Living Condition Type per sentence, followed by subject being found 16 times (Table 6C).

Tables 6A,B,C. Elements, counts, and example values/patterns Residence, Living Situation, and Living Conditions. For each source n= the total number of unique sentences that we eventually annotated for distinct elements. Percent of unique sentences that contained the element (Total number of instances of that element). Example values and patterns in bold are newly added to the existing model as a result of this work. Bolded Example Values represent items added to the existing model through this study.

Table 6A. Residence					
Elements	MTS (n=36)	UPMC (n=42)	FHS (n=504)	Total (n=581)	Example Values and Patterns
Status	65.7% (30)	76.2% (35)	58.3% (351)	60.1% (416)	lives in, resides in, homeless <ownership status> lives, live, living, moved, resides, residing, lived, staying, built, buying, came
Subject	2.9% (1)	-	0.8% (5)	0.9% (6)	mother's, in-laws, friends, daughter and son-in-law, <family member>
Negation	-	-	0.2% (1)	0.2% (1)	no <residence detail>, don't
Certainty	2.9% (1)	-	0.4% (2)	0.5% (3)	yes/present, no/absent, unknown, didn't know, apparently
Quantity	-	-	0.2% (1)	0.2% (1)	<#> steps, <#> floors/levels <#> <i>residence detail,</i> several, <#>
Temporal	14.3% (6)	7.1% (3)	6.9% (40)	7.4% (49)	currently, prior to hosp, recently, now
Duration	8.6% (3)	-	1.6% (9)	1.9% (12)	<#> years, <#> months, <#> days, few weeks/years
Duration Since Time Point	2.9% (1)	-	1.2% (6)	1.2% (7)	Since <year>, end of <month>, after <medical incident>
End Date	-	-	0.2% (1)	0.2% (1)	<date>
Residence Age	-	-	1.2% (7)	1.0% (7)	New, newer, 10-25 years, built before 1950
Residence Build Time Point	-	-	0.6% (3)	0.5% (3)	<date>
Start Age	-	-	0.2% (1)	0.2% (1)	<age>
Start Date	-	-	1.2% (6)	1.0% (6)	Summer, <year>, <MM/YYYY>, <DD/MM/YYYY>
Residence Type	45.7% (17)	76.2% (36)	50.2% (306)	51.8% (359)	house, apartment, nursing home, mobile home, <dwelling type>, home, assisted living, house. townhome, group home, condominium, senior housing
Residence Subtype	5.7% (2)	9.5% (4)	4.2% (21)	4.6% (27)	Multi-level, level, story, bedroom, floor, split-level
Residence Detail	5.7% (2)	2.4% (1)	2.2% (12)	2.4% (15)	own/rent safety devices, stairs, appliances, carpeted, independent, living, rear entry, basement

Residence Name	11.4% (5)	7.1% (4)	4.0% (23)	4.6% (32)	<facility name>
Geographic Location	8.6% (3)	-	2.6% (14)	2.8% (17)	<general geographic location>, campus, locally, nearby, community, up here, there, here
Location Detail	42.9% (18)	14.3% (7)	40.1% (277)	38.4% (302)	Specific geographic <country> <state> <neighborhood> <zip> <street address>
Other	-	-	-	-	-

Table 6B. Living Situation					
Elements	MTS (n=96)	UPMC (n=53)	FHS (n=665)	TOTAL (n=814)	Example Values and Patterns
Status	90.6% (100)	96.2% (53)	86.9% (670)	88.0% (823)	lives, lives with, live, living, resides, residing, visiting, lived, moving in, moved, stay, staying
Subject	26.0% (27)	22.6% (12)	37.3% (293)	35.0% (332)	spouse, parents, mother, father, child, roommate, family, <Family member> alone, child, <name>, boyfriend, significant other, roommate
Negation	1.0% (1)	-	0.5% (3)	0.5% (4)	no <subject> <living situation detail>, do not, don't, not, no
Certainty	2.1% (2)	-	-	0.2% (2)	yes/present, no/absent, unknown, apparently, either
Quantity	6.3% (7)	3.8% (3)	3.9% (32)	4.2% (42)	<#> subjects <#> in household
Temporal	-	-	0.2% (1)	0.1% (1)	Every other week
Duration	1.0% (1)	-	0.3% (2)	0.4% (3)	Few weeks, <##> years
Duration Since Time Point	-	-	0.3% (2)	0.2% (2)	Approximately, past several years
End Date	-	-	0.2% (1)	0.1% (1)	<MM/YY>
Timeframe	7.3% (8)	5.7% (3)	3.5% (28)	4.1% (39)	Currently, previously, prior to hospitalization, recently
Time Point	2.1% (2)	-	1.1% (8)	1.1% (10)	<MM/DD/YY>, <YYYY>, this week, now,
Living Situation Detail	4.2% (4)	5.7% (4)	2.6% (20)	2.9% (28)	inadequate, crowded, alone, privacy, together, independently, foster care
Family Member	65.6% (104)	75.5% (50)	68.7% (817)	68.8% (971)	Wife, brother, Child(ren), dad, daughter, father, husband, mom, mother
Side of Family	-	-	0.2% (1)	0.1% (1)	maternal
Current Age	8.3% (13)	-	3.0% (26)	3.4% (39)	<##> year old, age <##> years,
Other	-	-	-	-	-

Table 6C. Living Conditions					
Element	MTS (n=0)	UPMC (n=5)	FHS (n=33)	Total (n=38)	Example Values and Patterns
Status	-	40.0% (2)	12.1% (4)	15.8% (6)	housing contains, lives, no, live, living
Subject	-	-	24.2% (16)	21.1% (16)	<i>patient, <family member,></i>
Negation	-	40.0% (2)	24.2% (10)	26.3% (12)	no <i><living condition detail></i> , without
Certainty	-	20.0% (1)	3.0% (2)	5.3% (3)	yes/present, no/absent, unknown, apparently
Quantity	-	20.0% (1)	15.2% (8)	15.8% (9)	Excessive animals <i><#></i> <i>Living conditions type, good deal</i>
Temporal	-	-	-	-	-
Living Conditions Detail	-	20.0% (1)	12.1% (6)	13.2% (7)	water damage, home smelled of urine
Living Conditions Type	-	40.0% (9)	48.5% (30)	47.4% (39)	mold, insects, rodents, animals, water, heat, well water, filtered water, city water, condemned, electricity, excrement, light, water fluorinated
Living Conditions Type Subtype	-	20.0% (1)	3.0% (1)	5.3% (2)	Modifier of Living Conditions Type, unhygienic , presence of laundry facilities
Other	-	-	-	-	-

Discussion

While social determinants of health (SDOH) play an important role in the provision of patient care, they also play an important role in secondary use of data for research and quality improvement. Unfortunately, SDOH information is not well documented as discrete data in the EHR. By leveraging a diverse collection of notes including general social history documentation notes and notes authored by different types of clinical authors our evaluation helps information about the three topic areas of *Living Situation*, *Residence*, and *Living Conditions*. By far, *Living Situation* was the most common topic found of the three topic areas followed by *Residence* with much less information around *Living Conditions* in our dataset. The prevalence of documentation related to *Living Situation* appears to be an effort by clinicians to indicate the relative amount of potential social support for patients.

Many sentences included elements that crossed topic areas. For example, “She lives in <CITY NAME> with her parents and 2 older sisters.”, which includes *Residence* and *Living Situation*. Another example: “Lives at home with mom and two pets.”, crosses all three topic areas. Not surprisingly, physical therapist and social worker authored notes had a higher proportion of sentences related to *Residence*. The Social Documentation notes, which could have been authored by any clinician type, had the highest number of sentences related to *Living Situation* and *Living Conditions*.

Further analysis and mapping of the *Residence* sentences to the element axes showed persistent use of status, residence type and geographic location. Temporal entities and attributes were less prevalent with most documentation describing current state with some references to past situations. For *Living Situation* there was again significant presence of status as we as subject specifically family member. *Living conditions* was much less represented in these data sources and most references were related to type of water available.

The enhanced model representations (Tables 6A, B, C), built upon the foundation of previous standards evaluation work, now represent the analysis of 27 data sources including existing standards, terminologies, guidelines, and measures and Surveys as well as the analysis of 2491 notes. The analysis of the notes added more elements. For

Residence and Living Situation the temporality elements “Duration Since Time Point”, “Time Point”, and “Time Frame” were added as was the element “Residence Type Subtype” and “Living Situation Detail Subtype”. For Living Conditions the temporal elements “Duration Since Time Point” and “Time Point” were added. Overall, the EHR unstructured text significantly contributed to enhance and strengthen the model representation. Value lists are much more complete for each of the elements.

Although we obtained an inter-rater reliability of $K = 0.84$ and proportion agreement of 0.98, we observed some inconsistency between reviewers regarding the difference between Geographic Location as opposed to Location Detail. This inconsistency was resolved and the annotation schema was amended accordingly. Another challenge area was around statements regarding safety concepts. In some cases safety items such as railings and stairs were annotated as Residence Detail and in other they were annotated as Living Conditions Detail. More work will be needed to sort out where these concepts logically fit the best. Lastly, while it is expected that not every sentence would have every element, in examination of these results it was noted that in the Residence annotations the number of sentences that had a “status” documented was lower than expected. After further manual review the issue was traced back to several of the question answer type “Living Arrangements” could be considered a section header and, in past work, section headers have been identified separately from text. For this work these sentences were left as is and will be considered in future work and iterations of the annotation schema.

In summary, this work has demonstrated that the SDOH topic areas of *Residence*, *Living Situation*, and *Living Conditions* are being documented in the EHR within unstructured text, specifically general progress notes, Social Documentation notes, and notes authored by Social Workers, Physical Therapists, and Occupational Therapists. This analysis contributes to overall representation models for these three topic areas. Next steps will include an evaluation of flow sheet documentation related to these three topic areas and further enhancement of the model representation that can be used to extract information from EHR text, to design discrete data collection tools for the EHR, and to contribute to the development of ontology for the social history topics of *Residence*, *Living Situation*, and *Living Conditions*.

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