CMPE 287 Software Quality Testing HW1

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Question #1: Basic Concepts

a. What are the challenges in testing AI software?

Ans. Al software quality testing field is comparatively new, hence not been exploit much. Due to which, it faces many challenges. Challenges are as follows:

- All applications are built with machine learning algorithms and big data. So to finalize the requirements and coverage criteria becomes difficult in this case
- Al software are learning based system. So to develop quality test models using systematic methods become difficult
- Development of automatic solution tools to support AI based software is challenging
- b. Why software testing is so hard?

Ans. There are various reasons why software testing is so hard. Some of them are as follows-

- While testing an application, tester not check just for correctness. Performance is also an important parameter, that should be checked while performing testing
- Testing should be performed using negative test cases as well. In negative test cases, application should give proper error message instead of abrupt failure
- Testing has no definite end limits. But at least the application should be tested for all paths possible (path coverage) in the application. And it should be tested using all possible inputs randomly
- c. What are three major challenges in testing software components?

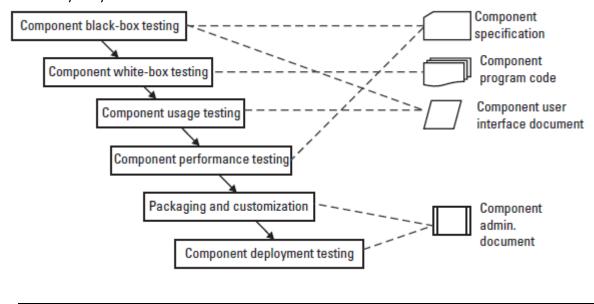
 Ans. Testing of the software components is very important. But while executing it various challenges come across. Those are as follows-
- **Reusability of component test:** For efficient testing of software components, reuse of component test is very vital. The important factor in reusing component test is to develop systematic methods and tools to prepare test suites to store various test resources such as test cases, test data, test scripts.
- Construction of testable components: Software component should be testable with the use
 of standardized test resources. Constructing the testable component is another challenge.
 Testable component have below features-
 - 1. It should be traceable
 - 2. It must have a set of built-in interfaces to interact with set if testing resources.
 - 3. It must have well defined test architecture model and built-in test interfaces to support their interactions with component test suits and test beds.
 - 4. Testable components with built-in tests must use the standard technique to enclose built-in tests.
- Construction of component test drivers/stubs in a symmetrical way: In real world practice, engineers use ad-hoc approach. They create product specific test drivers and stubs according to requirements and specification given for that particular component. The disadvantage of this approach is that generated test drivers and stubs are used only for that specific product.

d. Identify the three major differences between white-box testing and black-box testing. Ans.

Black-box testing	White-box testing
In Black-box testing, internal structure or the	In White-box testing, the tester has
program or the code is hidden and nothing is	knowledge about the internal structure of
known about it to the tester.	the code or the program of the software.
It is a functional/behavioural testing, mostly	It is a structural/logic testing, mostly done by
done by software testers.	software developers.
No knowledge of implementation and	Knowledge of implementation and
programming is needed.	programming is required.

- e. List vendor-oriented test processes for reusable components.
- Ans. Following are the steps-
 - **Step 1:** In this step, any traditional black-box testing can be used to check correctness and completeness of an application. It basically checks the functions and behaviours based on the given specifications
 - **Step 2:** In this step, developer uses any of the white-box testing techniques to check the internal errors in logic, data objects, data structures
 - **Step 3:** In this step, Tester exercises various component usage patterns through component interfaces to final check the correct functions and behaviours
 - Step 4: In this step, testers and QA engineers all together evaluate the performance
 - **Step 5:** In this step, testing is performed on built-in customization features and packaging approaches
 - **Step 6:** In this last step, deployment testing is performed to check the design and implementation according to the component model

Following image is taken from a book "Testing and Quality Assurance for Component-Based Software" by Jerry Gao et al.



Question #2: Equivalence Partitioning Testing Questions

(a). List of identified equivalence classes-

Length of the			10 <= length <= 16													
Letter(a-z)						(a-z)										
Digits(0-6)						(0)-6)									
Uppercase																
letter(A-Z)	length < 10	th < 10	(A-Z)				length > 16									
Special	no letter(a-z)		no digit/0 6\	no digit(0-6)	nonnoresso					ieugui > 10						
Char(o,&,.,*,%,\$,			no digit(o-o)		no digit(o-o)	no digit(0-0)	no digit(0-0)	no digit(o-o)	no digit(0-0)	- ' '	No special char at all		No special char at a		othe	r than
#)					letter(A-Z)	(o		(0,&,.,*,%,\$,#	(o,&,.,*,%,\$,#)	(0,&,.,	*,%,\$,#)					
Incorrect/correct					Incorrect	correct		Incorrect	correct							
password					password	password		password	Password							
Partitions	P0	P1	P2	P3	P4	P5	P6	P7	P8	P9						

(b). Test-cases:

Test case	Equivalent class	Input	Expected Output
id	partitioning		
00	number P0	Length of PIN is less than 10	"Invalidated
		e.g-"america"	Password"
01	P1	1. Length of PIN is from 10 to 16	"Invalidated
		2. No letter(a-z) e.g- "0123456789"	Password"
02	P2	1. Length of PIN is from 10 to 16	"Invalidated
		2. Atleast one letter(a-z)	Password"
		3. No digit	
		e.g- "helloamerica"	
03	P3	1. Length of PIN is from 10 to 16	"Invalidated
		2. Atleast one letter(a-z)	Password"
		3. Atleast one digit(0-6)	
		4. No uppercase letter	
		e.g- "helloamerica1"	
04	P4	1. Length of PIN is from 10 to 16	"Incorrect Password"
		2. Atleast one letter(a-z)	
		3. Atleast one digit (0-6)	
		4. Atleast one uppercase letter(A-Z)	
		5. No special character	
		6. Incorrect PIN number	
		e.g- "Helloamerica1"	
05	P5	1. Length of PIN is from 10 to 16	"Successfully login"
		2. Atleast one letter(a-z)	
		3. Atleast one digit (0-6)	
		4. Atleast one uppercase letter(A-Z)	
		5. No special character	
		6. Correct PIN number	
		e.g- "Helloamerica1"	

06	P6	1. Length of PIN is from 10 to 16	"Invalidated
		2. Atleast one letter(a-z)	Password"
		3. Atleast one digit(0-6)	
		4. Atleast one special char from (o,&,.,*,%,\$,#)	
		e.g- "Helloamerica1*"	
07	P7	1. Length of PIN is from 10 to 16	
		2. Atleast one letter(a-z)	"Incorrect Password"
		3. Atleast one digit(0-6)	
		4. Atleast one special char other than	
		(o,&,.,*,%,\$,#)	
		5. Incorrect PIN number	
		e.g- "Helloamerica@1"	
80	P8	1. Length of PIN is from 10 to 16	"Successfully login"
		2. Atleast one letter(a-z)	
		3. Atleast one digit(0-6)	
		4. Atleast one special char other than	
		(o,&,.,*,%,\$,#)	
		5. Correct PIN number	
		e.g- "Helloamerica@1"	
09	P9	1. Length of PIN is less than 10	"Invalidated
		e.g- "helloamericahelloamerica"	Password"

Question #3: Boundary Value Testing

(a). Define all boundaries for Z(X, Y) and (Answer shown in the below table)

(b) Define the boundary values for each boundary

Boundary Name Boundary Range		Boundary values for X	Boundary values for Y
B1	X[1,5] & Y[3,9]	(0,1,2) & (4,5,6)	(2,3,4) & (8,9,10)
B2	X[6] & Y[0]	(5,6,7)	(-1,0,1)
B3	X[7,10] & Y[20,34]	(6,7,8) & (9,10,11)	(19,20,21) & (33,34,35)

(c). Define the test cases for each boundary.

Test cases for B1:

Input value of X	Input value of Y	Expected Output (Z)
(if X has any of these values)	(if Y has any of these values)	(Put the respective X,Y value in
		the equation)
(0)	(2,3,4,8,9,10)	-1
(1,2,4,5)	(2,10)	-1
(1,2,4,5)	(3,4,8,9)	2X – 4Y + 5
(6)	(2,3,4,8,9,10)	-1

Test cases for B2:

Input value of X	Input value of Y	Expected Output (Z)
------------------	------------------	---------------------

(if X has any of these values)	(if Y has any of these values)	(Put the respective X,Y value in the equation)
(5)	(-1,0,1)	-1
(6)	(-1,1)	-1
(6)	(0)	Undefined
(7)	(-1,0,1)	-1

Test cases for B3:

Input value of X (if X has any of these values)	Input value of Y (if Y has any of these values)	Expected Output (Z) (Put the respective X,Y value in the equation)		
(6)	(19,20,21,33,34,35)	-1		
(7,8,9,10)	(19,35)	-1		
(7,8,9,10)	(20,21,33,34)	X + Y + 1		
(11)	(19,20,21,33,34,35)	-1		

Question #4: Decision table testing question

(a) A decision table (Table A) for sending a reply message without attachments.

		5	cen	ario	S
		1	2	3	4
	"To" (adding recipient)	Т	Т	Т	F
Conditions	Valid "To/CC/BCC"	Т	Т	F	X
	Filled Subject Line/ Email body	Т	F	X	Х
	Error				
Actions	Warning				
	Success				

A decision table (Table B) for sending a reply message with attachments.

		Scenarios					
		1	2	3	4	5	6
	"To" (adding recipient)	Т	Т	Т	Т	Т	F
Conditions	Valid "To/CC/BCC"	Т	Т	Т	Т	F	Χ
	Filled Subject Line/ Email body	Т	Т	Т	F	X	Χ
	Suggestive text in email body	Т	Т	F	X	X	Χ
	Attachment	Т	F	Χ	X	X	Χ
	Error						
Actions	Warning						
	Success						
					1		

(b). Test cases for Table A:

Test case	"To" (recipient)	Validation of	Subject line/	Expected output
id		(To/CC/BCC)	Body text	

00	(Empty)	-	Proper or	Error message: "Please
			improper	add recipient"
01	Entered like	Invalid	Proper or	Error message: "Please
	"akshatadeo"	recipient	improper	enter valid email id"
02	Entered like	Valid	(Empty)	Warning message: "Do
	"Akshata.deo@sjsu.edu"	recipients		you want to send email
				without subject?"
03	Entered like	Valid	Proper subject	Success message: "Email
	"Akshata.deo@sjsu.edu"	recipients	and body text	has been sent"
04	"To" is empty but CC or	Valid	Proper subject	Success message: "Email
	BCC is filled	recipients	and body text	has been sent"
05	"To" is empty but CC or	Invalid	(Empty)	Error message: "Please
	BCC is filled	recipient		enter valid email id"

Question #5: AI testing and AI-based software testing questions

(a)

- [1] Tosun, Ayse, Ayse Bener, and Resat Kale. "Ai-based software defect predictors: Applications and benefits in a case study." *In Twenty-Second IAAI Conference*. 2010.
- [2] D'silva, Vijay, Daniel Kroening, and Georg Weissenbacher. "A survey of automated techniques for formal software verification." *IEEE Transactions on Computer-Aided Design of Integrated Circuits and Systems* 27.7 (2008): 1165-1178.
- [3] Atif M. Memon, "Using tasks to automate regression testing of GUIs." *In IASTED International Conference on Artificial Intelligence and Applications-AIA*, p. 477-82. 2004.
- [4] Nguyen, B.N., Robbins, B., Banerjee, I. and Memon, A., 2014. GUITAR: an innovative tool for automated testing of GUI-driven software. *Automated software engineering*, 21(1), p.65-105.
- [5] C. V. Ramamoorthy and S. F. Ho, "Testing large software with automated software evaluation systems," *in IEEE Transactions on Software Engineering*, vol. SE-1, no. 1, pp. 46-58, March 1975.

Paper ID	Issues and challenges (if there is any)	Testing Methods (if there is any)	Quality parameters (if there is any)	Quality process (if there is any)	Supported Testing types and applications
	Model Calibration				
	requirement	Team Software		Team	
	Limitation to track	Process (TSP).	simplicity	Software	General
	predictive	Naïve Bayes	accuracy	Process	software
[1]	performance	classifier	Consistency	(TSP).	application
[2]	completeness is only obtainable on very 'shallow' programs	static analysis with abstract domains Model Checking Bounded Model Checking	symbolic analysis Abstraction Concurrency	-	C,C++,Ada
[3]	Regression test selection problem coverage identification problem	Regression	cost of testing		GUI

	Requires manually		Modularity		
	specifying the		Event		
	ignored and		Selection		
	terminal widgets in	GUITAR	oracle		
[4]	a UI.	Framework	specification	-	GUI
	tools not well-				
	structured	selective testing			large
	machine dependent	exhaustive testing	correctness		software
[5]	tools	formal proofs	Performance	-	systems

(b)

- [1] Leofante F, Narodytska N, Pulina L, Tacchella A. Automated verification of neural networks: Advances, challenges and perspectives. arXiv preprint arXiv:1805.09938. 25 May 2018.
- [2] Van Wesel, Perry, and Alwyn E. Goodloe. "Challenges in the verification of reinforcement learning algorithms." (2017).
- [3] Fei Liu and Ming Yang, "Verification and validation of AI simulation systems," *Proceedings of 2004 International Conference on Machine Learning and Cybernetics (IEEE Cat. No.04EX826), Shanghai, China*, p. 3100-3105 vol.5, 2004.
- [4] Seshia SA, Sadigh D, Sastry SS. Towards verified artificial intelligence. arXiv preprint arXiv:1606.08514. 27 June 2016.
- [5] Xie, X., Ho, J.W., Murphy, C., Kaiser, G., Xu, B. and Chen, T.Y., 2011. Testing and validating machine learning classifiers by metamorphic testing. *Journal of Systems and Software*, 84(4), p.544-558.
- [6] Murphy, Christian, Gail E. Kaiser, and Marta Arias. "An approach to software testing of machine learning applications." (2007).

Paper ID	Issues and challenges (if there is any)	Testing Methods (if there is any)	Quality parameters (if there is any)	Quality process (if there is any)	Supported Testing types and applications
	Coordinate the				
	efforts of two	CDCL-style SAT			
	"separated at	solving			
	birth" Al	algorithm			
	communities:	Satisfiability			
	Machine	Modulo			
	Learning and	Theories (SMT)	Local-Invariance,		
	Automated	Mixed Integer	Global-		BNN,
	Reasoning.	Linear	Invariance,	State-of-	DNN(ReLU),
	Mate precision	Programming	Invertibility,	the-art	DNN(ReLU+Pooling),
[1]	with scalability	(MIP)	Equivalence	design	DNN, NN
	Assumptions of		State to action		
	operating		mappings,		
	environments		Action		
	Assumptions of	Formal	sequences,		
	Platform	Verification	Algorithm		
	(Failure and	Runtime	properties		
	Fidelity)	Verification (RV)	independent of		
	Assumption of	Thrun's Validity	data,		reinforcement
[2]	Data	Interval Analysis	Validating	-	learning algorithms

	(Distribution,		assumptions on		
	samples etc.)		training data,		
	Algorithm assumption				
	assumption	logic formalisms			
		graph theory,			
		empirical,			
		heuristic,			
		formal			
		methods, and			
	System	optimization Probability			
	representation	methods, gray			
	assumption	theory, turing			
	The efficacy of	test and			Rule-based systems
	the inductively	expert			Knowledge-based
	learnt model	judgement	_		systems
	can never be	Runtime	Completeness,		Nondeterministic
	better than the quality of the	verification tools (Java Path	conflicts, redundancies,		systems Complex software
[3]	input examples.	Finder)	Nondeterminism	_	Adaptive systems
[-]	The state of the s	Boolean			Thoract o systems
	Environment	satisfiability			
	Modeling	(SAT) solvers			
	Formal	Binary Decision			
	Specification	Diagrams			
	Modeling Systems that	(BDDs) satisfiability			
	Learn	modulo			
	Computational	theories (SMT)			
	Engines for	solvers			
	Training,	Formal			
	Testing, and	Verification	Effectiveness,		
[4]	Verification	Methods	Scalability	-	General AI methods
		Metamorphic testing: - It uses			
		properties of			
		function			
		properties to			
		predict and	Average		
		analyse the	violation		
		changes resulting in the	percentage of all MR		
		output by	(Metamorphic		supervised
		implementing	Relation)		classification
		the changes in	Performance		algorithms: -
		the input test	Effectiveness		k-Nearest Neighbors
		cases.	Mutation		(kNN)
[6]	The oracle	Cross-validation	analysis (for Bio-		Naïve Bayes
[5]	problem	Analysis	informatics)	-	Classifier (NBC)

	Negative			
	handles could			
	not be			
	addressed.			
	Repeated or			
	missing			
	attribute values			
	are not handled			
	properly.			
	It does not			
	provide how			
	many non-	SVM-Light		
	failures should	MartiRank		
	be made each	ranking	Effectiveness,	
[6]	partition.	Algorithm	Scalability	SVM