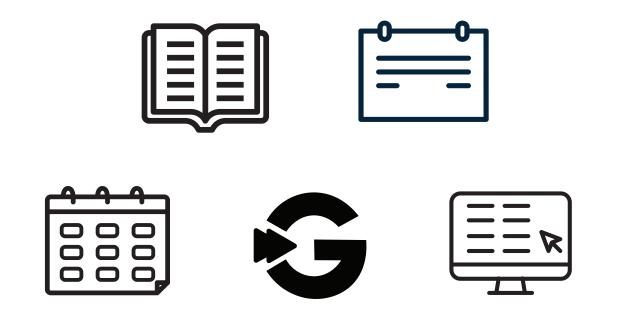


2nd Edition

Sam A. Broverman, PhD, ASA



An SOA Exam



Study Manual for Exam FAM-S

2nd Edition

Sam A. Broverman, PhD, ASA



Actuarial & Financial Risk Resource Materials Since 1972

Copyright © 2024, ACTEX Learning, a division of ArchiMedia Advantage Inc.

No portion of this ACTEX Study Manual may be reproduced or transmitted in any part or by any means without the permission of the publisher.

Welcome to Actuarial University

Actuarial University is a reimagined platform built around a more simplified way to study. It combines all the products you use to study into one interactive learning center.

You can find integrated topics using this network icon.

When this icon appears, it will be next to an important topic in the manual. Click the **link** in your digital manual, or search the <u>underlined topic</u> in your print manual.

Pareto Distribution

The (Type II) **Pareto distribution** with parameters $\alpha, \beta > 0$ has pdf

1. Login to: www.actuarialuniversity.com

2. Locate the **Topic Search** on your exam dashboard and enter the word or phrase into the search field, selecting the best match.

3. A topic "**Hub**" will display a list of integrated products that offer more ways to study the material.

4. Here is an example of the topic **Pareto Distribution**:

$$f(x) = \frac{\alpha \beta^{\alpha}}{(x + \beta)^{\alpha + 1}}, \quad x > 0$$
and cdf
$$F_{P}(x) = 1 - \left(\frac{\beta}{x + \beta}\right)^{\alpha}, \quad x > 0.$$
If *X* is Type II Pareto with parameters α, β , then
$$E[X] = \frac{\beta}{\alpha - 1} \text{ if } \alpha > 1,$$
and
$$Far[X] = \frac{\alpha \beta^{2}}{\alpha - 2} - \left(\frac{\alpha \beta}{\alpha - 1}\right)^{2} \text{ if } \alpha > 2.$$
ACTEX Manual for P

ACTEX Manual for P

Y

Probability for Risk Management, 3rd Edition

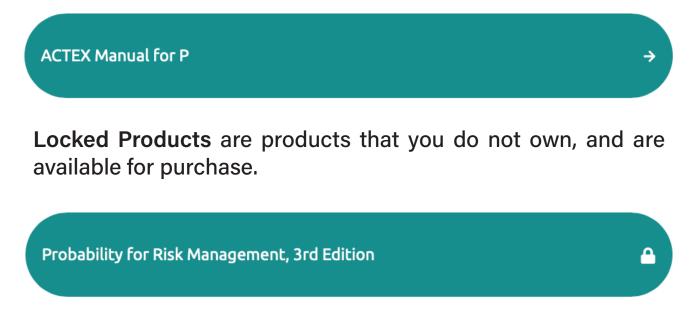
COAL for SRM

ASM Manual for IFM

Lxam FAM-S Video Library

Within the Hub there will be unlocked and locked products.

Unlocked Products are the products that you own.



Many of Actuarial University's features are already unlocked with your study program, including:

Instructional Videos*	Planner
Topic Search	Formula & Review Sheet

Make your study session more efficient with our Planner!

🖶 Planner				
Template A	ACTEX FM Study Manual - New 2022 sylla	ibus		\$
Begin Study	07/01/2023	End Study 11/14/2023		
7	///2023 - 7/16/2023	Interest Rates and the Time Value of Money	\$	→
 ✓ 7 	/16/2023 - 8/12/2023	Annuities	\$	→
✓ 8	8/12/2023 - 8/27/2023	Loan Repayment	*	→
✓ 8	8/27/2023 - 9/15/2023	Bonds	*	→
۶ و)/15/2023 - 9/22/2023	Yield Rate of an Investment	\$	→
 ✓ 9)/22/2023 - 10/11/2023	The Term Structure of Interest Rates	\$	÷
1	0/11/2023 - 10/30/2023	Asset-Liability Management	\$	

*Available standalone, or included with the Study Manual Program Video Bundle



Practice. Quiz. Test. Pass!

- 16,000+ Exam-Style Problems
- Detailed Solutions
- Adaptive Quizzes
 - 3 Learning Modes

.

3 Difficulty Modes

Free with your ACTEX or ASM Interactive Study Manual

Available for P, FM, FAM, FAM-L, FAM-S, ALTAM, ASTAM, MAS-I, MAS-II, CAS 5, CAS 6U & CAS 6C

Prepare for your exam confidently with GOAL custom Practice Sessions, Quizzes, & Simulated Exams

Actuarial	University	/			-	<u> </u>							7	
QUESTION 19	DF 704	Questio	on #		Go!	2) (F			♦ Prev	Next 🕨	X			Quickly access the
Question									D	ifficulty: A	dvanced 😧	Ē,		Hub for additional
An airport purchas every full ten inche							ts of snowfa	dl. The insu	rer pays tł	ne airport	300 for		l	learning.
The following table							winter seaso	on) snowfal	l. in inche	s, at the ai	irport.		(
Inches	[0,20)	[20,30]	[30,40)	[40,50)	[50,60)	[60,70)	[70,80)	[80,90)	[90,inf					Flag problems for review, record
Probability	0.06	0.18	0.26	0.22	0.14	0.06	0.04	0.04	0.00	<i>,</i>				notes, and email
Calculate the stand						0.00	0.04	0.04	0.00					your professor.
Possible Answers			I	1										
A	134	~	235	E	× 271		D 31	3	Е	352			(View difficulty level.
Help Me Start											^			
Find the probabiliti	es for the fo	our possible	payment an	10unts: 0, 30	00, 600, and	700.								
Solution							Helpful strategies to get you started.							
With the amount of	snowfall as	X and the a	amount paid	-									l	
$\begin{array}{c c c c c c c c c c c c c c c c c c c $														
Common Questions	& Errors										*	Ē	(
Students shouldn't overthink the problem with fractional payments of 300. Also, account for probabilities in which payment cap of 700 is reached.														
In these problems, we must distinguish between the REALT RV (how much snow falls) and the PAYMENT RV (when does the insurer pay)?. The problem states "The insurer pays the airport 300 for every full ten inches of snow in excess of 40 inches, up to a policy maximum of 700 ." So the insurer will not start paying UNTIL AFTER 10 full inches in excess of 40 inches of snow is reached (say at 50+ or 51). In other words, the insurer will pay nothing if X<50.					-									
Rate this problem	් Exce	llent 🕞	Needs Impro	ovement	Inadequ	ate								give feedback.

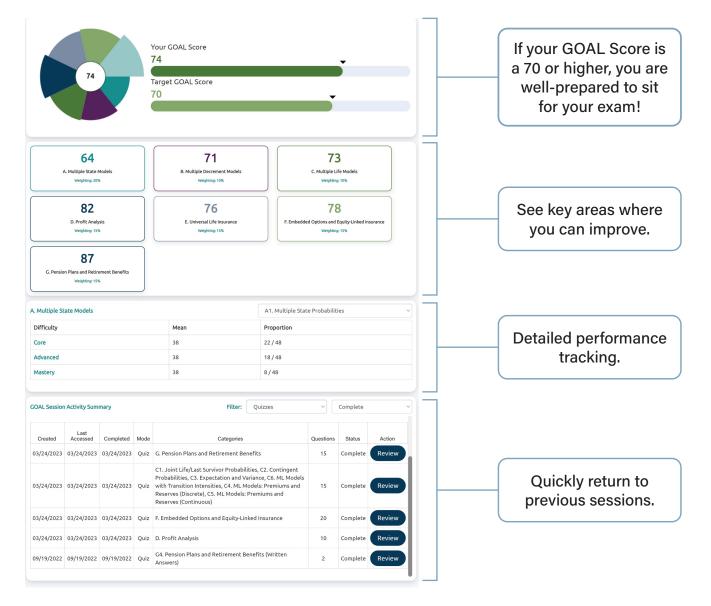


Track your exam readiness with GOAL Score!

Available for P, FM, FAM, FAM-L, FAM-S, ALTAM, ASTAM, MAS-I, MAS-II, & CAS 5

GOAL Score tracks your performance through GOAL Practice Sessions, Quizzes, and Exams, resulting in an aggregate weighted score that gauges your exam preparedness.

By measuring both your performance, and the consistency of your performance, **GOAL Score** produces a reliable number that will give you confidence in your preparation before you sit for your exam.



Contents

INTRODUCTORY COMMENTS

Section 1	Review of Probability	1
1.1	Basic Probability Concepts	1
1.2	Conditional Probability and Independence of Events	4
Section 1 Pr	coblem Set	8
Section 2	Review of Random Variables I	19
2.1	Calculus Review	19
2.2	Discrete Random Variable	20
2.3	Continuous Random Variable	21
2.4	Mixed Distribution	23
2.5	Cumulative Distribution, Survival and Hazard Functions	23
2.6	Examples of Distribution Functions	24
2.7	The Empirical Distribution	27
2.8	Gamma Function and Related Functions	28
Section 2 Pr	coblem Set	30
Section 3	Review of Random Variables II	35
3.1	Expected Value and Other Moments of a Random Variable	35
3.2	Percentiles and Quantiles of a Distribution	38
3.3	The Mode of a Distribution	38
3.4	Random Samples and The Sampling Distribution	40
3.5	The Normal Distribution	41
3.6	Approximating a Distribution Using a Normal Distribution	43
3.7	Distribution of a Transformation of Random Variable X	45
Section 3 Pr	oblem Set	46
Section 4	Review of Random Variables III	53
4.1	Joint Distribution of Random Variables X and Y	53
4.2	Marginal distribution of X found from a joint distribution of X and Y \ldots	56
4.3	Independence of Random Variables X and Y	58
4.4	Conditional Distribution of Y Given $X = x$	58
4.5	Covariance Between Random Variables X and Y	61
4.6	Coefficient of correlation between random variables X and Y	62

 $\mathbf{x}\mathbf{v}$

Section 4 Pr	roblem Set	63
Section 5	Parametric Distributions	69
5.1	Parametric Distributions	69
5.2	Families of Distributions	71
5.2.	.1 The Linear Exponential Family	71
5.3	The Single Parameter Pareto and Two Parameter Pareto Distributions $\ . \ .$	72
5.4	A Note on the Poisson Distribution	73
Section 5 Pr	roblem Set	74
Section 6	Distribution Tail Behavior	81
6.1	Measuring Tail Weight Using Existence of Moments	81
6.2	Comparing the Tail Weights of Two Distributions	82
Section 6 Pr	roblem Set	83
Section 7	Mixture Of Two Distributions	85
7.1	Mixture of Two Distributions	85
7.2	Formulating a Mixture Distribution as a Combination of	
	Conditional Distributions	88
7.3	The Variance of a Mixed Distribution IS (usually) NOT	
	the Weighted Average of the Variances	89
Section 7 Pr	roblem Set	91
Section 8	Mixture Of <i>n</i> Distributions	01
8.1	Mixture of n Distributions $\ldots \ldots 1$	01
8.2	Two Important Rules of Probability	.03
Section 8 Pr	roblem Set	.06
Section 9	Frequency Models The Number Of Claims 1	13
9.1	Poisson Distribution	.14
9.2	Binomial Distribution	.15
9.3	Negative Binomial Distribution	.16
9.4	The $(a,b,0)$ Class of Discrete Distributions $\ldots \ldots \ldots$.17
9.5	The $(a,b,1)$ Class of Discrete Distributions $\ldots \ldots \ldots$	
Section 9 Pr	roblem Set	.20
Section 10	Policy Limits And The Limited Loss 1	39
10.1	Policy Limit u and the Limited Loss Random Variable $\ldots \ldots \ldots \ldots \ldots \ldots \ldots$	
10.2	Limited Expected Value	
Section 10 P	Problem Set	.44
Section 11	Policy Deductible I - The Cost Per Loss 1	47

11.3 Franchise Deductible 152 Section 11 Problem Set 154 Section 12 Policy Deductible II - The Cost Per Payment 167 12.1 Ordinary Policy Deductible d and Cost Per Payment 167 12.2 Expected Cost Per Payment With Deductible d 172 12.3 Variance of Cost Per Payment With Deductible d 172 12.4 Franchise Deductible d 174 Section 12 Problem Set 175 Section 13 Deductibles Applied To The Uniform, Exponential And Pareto Distributions 189 Section 13 Deductibles Applied To The Uniform, Exponential And Pareto Distributions 189 Section 14 Combined Limit And Deductible 203 14.1 Maximum Covered Loss u Combined With Policy Deductible $d < u$ 203 14.2 PDF and CDF of Y_L and Y_P 207 14.3 A Few Comments 208 14.4 Graphical Representation of $E[Y_L]$ 209 Section 15 Additional Policy Adjustments 227 15.2 Inflation 229 15.3 Policy Deductible d in Combination With Maximum Covered Loss u and Coinsurance Factor α and Inflation Factor	11.1	Ordinary Policy Deductible d and Cost Per Loss $\ldots \ldots \ldots \ldots \ldots \ldots 147$
Section 11 Problem Set 154 Section 12 Policy Deductible II - The Cost Per Payment 167 12.1 Ordinary Policy Deductible d and Cost Per Payment 167 12.2 Expected Cost Per Payment With Deductible d 172 12.3 Variance of Cost Per Payment With Deductible d 172 12.4 Franchise Deductible d 174 Section 12 Problem Set 175 Section 13 Deductibles Applied To The Uniform, Exponential And Pareto Distributions 189 Section 13 Problem Set 197 Section 14 Combined Limit And Deductible 203 14.1 Maximum Covered Loss u Combined With Policy Deductible $d < u$ 203 14.2 PDF and CDF of Y_L and Y_P 207 14.3 A Few Comments 209 Section 14 Problem Set 210 Section 15 Additional Policy Adjustments 227 15.1 Coinsurance and Copay 227 15.3 Policy Deductible d in Combination With Maximum Covered Loss u and Coinsurance Factor α and Inflation Factor r 231 16.1 The Compound Distribution 243 16.2 The Compound Distribution Model for Aggregate Claims in One Period of Time 243 16.3 The Normal Approximation to a Compound Distribution 245 16.3 The Normal Ap	11.2	Modeling Bonus Payments
Section 12 Policy Deductible II - The Cost Per Payment 167 12.1 Ordinary Policy Deductible d and Cost Per Payment 169 12.2 Expected Cost Per Payment With Deductible d 172 12.3 Variance of Cost Per Payment With Deductible d 172 12.4 Franchise Deductible d 174 Section 12 Problem Set 175 Section 13 Deductibles Applied To The Uniform, Exponential And Pareto Distributions 189 Section 13 Polem Set 197 Section 14 Combined Limit And Deductible 203 14.1 Maximum Covered Loss u Combined With Policy Deductible $d < u$ 203 14.2 PDF and CDF of Y_L and Y_P 207 14.3 A Few Comments 208 14.4 Graphical Representation of $E[Y_L]$ 209 Section 15 Additional Policy Adjustments 227 15.1 Coinsurance and Copay 227 15.2 Inflation 231 and Coinsurance Factor α and Inflation Factor r 231 Section 15 Aggregate Models - Compound Distributions 243 16.1 The Compound Distributi	11.3	Franchise Deductible
12.1 Ordinary Policy Deductible d and Cost Per Payment 167 12.2 Expected Cost Per Payment 169 12.3 Variance of Cost Per Payment With Deductible d 172 12.4 Franchise Deductible d 174 Section 12 Problem Set 175 Section 13 Deductibles Applied To The Uniform, Exponential And Pareto Distributions 189 Section 13 Problem Set 197 Section 14 Combined Limit And Deductible 203 14.1 Maximum Covered Loss u Combined With Policy Deductible $d < u$ 203 14.2 PDF and CDF of Y_L and Y_P . 207 14.3 A Few Comments 208 14.4 Graphical Representation of $E[Y_L]$ 209 Section 15 Additional Policy Adjustments 227 15.1 Coinsurance and Copay 227 15.2 Inflation 229 15.3 Policy Deductible d in Combination With Maximum Covered Loss u and Coinsurance Factor α and Inflation Factor r 231 16.1 The Compound Distribution Model for Aggregate Claims in One Period of Time 243 16.2 The Compound Distribution Model for Aggregate Claims in One Period of Time 243	Section 11 P	roblem Set $\ldots \ldots 154$
12.2 Expected Cost Per Payment 169 12.3 Variance of Cost Per Payment With Deductible d 172 12.4 Franchise Deductible d 174 Section 12 Problem Set 175 Section 13 Deductibles Applied To The Uniform, Exponential And Pareto Distributions 189 Section 13 Problem Set 197 Section 14 Combined Limit And Deductible 203 14.1 Maximum Covered Loss u Combined With Policy Deductible $d < u$ 203 14.2 PDF and CDF of Y_L and Y_P . 207 14.3 A Few Comments 208 14.4 Graphical Representation of $E[Y_L]$ 209 Section 14 Problem Set 210 Section 15 Additional Policy Adjustments 227 15.1 Coinsurance and Copay 227 15.2 Inflation 229 15.3 Policy Deductible d in Combination With Maximum Covered Loss u 234 Section 15 Problem Set 234 Section 16 Aggregate Models - Compound Distributions 243 16.1 The Compound Distribution Model for Aggregate Claims in One P	Section 12	Policy Deductible II - The Cost Per Payment 167
12.3 Variance of Cost Per Payment With Deductible d 172 12.4 Franchise Deductible d 174 Section 12 Problem Set 175 Section 13 Deductibles Applied To The Uniform, Exponential And Pareto Distributions 189 Section 13 Deductibles Applied To The Uniform, Exponential And Pareto Distributions 189 Section 14 Combined Limit And Deductible 203 14.1 Maximum Covered Loss u Combined With Policy Deductible $d < u$ 203 14.2 PDF and CDF of Y_L and Y_P . 207 14.3 A Few Comments 208 14.4 Graphical Representation of $E[Y_L]$ 209 Section 14 Problem Set 210 Section 15 Additional Policy Adjustments 227 15.1 Coinsurance and Copay 227 15.2 Inflation 229 15.3 Policy Deductible d in Combination With Maximum Covered Loss u and Coinsurance Factor α and Inflation Factor r 231 Section 15 Problem Set 234 Section 16 Aggregate Models - Compound Distribution 243 I6.1 The Compound Distribution Model for Aggregate Claims in One Period	12.1	Ordinary Policy Deductible d and Cost Per Payment $\ldots \ldots \ldots$
12.4 Franchise Deductible d 174 Section 12 Problem Set 175 Section 13 Deductibles Applied To The Uniform, Exponential And Pareto Distributions 189 Section 13 Problem Set 197 Section 14 Combined Limit And Deductible 203 14.1 Maximum Covered Loss u Combined With Policy Deductible $d < u$ 203 14.2 PDF and CDF of Y_L and Y_P 207 14.3 A Few Comments 208 14.4 Graphical Representation of $E[Y_L]$ 209 Section 14 Problem Set 210 Section 15 Additional Policy Adjustments 227 15.1 Coinsurance and Copay 227 15.2 Inflation 229 15.3 Policy Deductible d in Combination With Maximum Covered Loss u 234 Section 15 Problem Set 234 Section 16 Aggregate Models - Compound Distributions 243 16.1 The Compound Distribution Model for Aggregate Claims in One Period of Time 16.2 The Compound Poisson Distribution 245 16.3 The Normal Approximation to a Compound Distribution 246 16.4 The Convolution Method for Finding the Dis	12.2	Expected Cost Per Payment
Section 12 Problem Set 175 Section 13 Deductibles Applied To The Uniform, Exponential And Pareto Distributions 189 Section 13 Problem Set 197 Section 14 Combined Limit And Deductible 203 14.1 Maximum Covered Loss u Combined With Policy Deductible $d < u$ 203 14.2 PDF and CDF of Y_L and Y_P 207 14.3 A Few Comments 208 14.4 Graphical Representation of $E[Y_L]$ 209 Section 14 Problem Set 210 Section 15 Additional Policy Adjustments 227 15.1 Coinsurance and Copay 222 15.2 Inflation 229 15.3 Policy Deductible d in Combination With Maximum Covered Loss u and Coinsurance Factor α and Inflation Factor r 231 Section 15 Problem Set 234 Section 16 Aggregate Models - Compound Distributions 243 16.1 The Compound Distribution Model for Aggregate Claims in One Period of Time 243 16.2 The Compound Poisson Distribution 245 16.3 The Normal Approximation to a Compound Distribution 246 16.4 The Convolution Method for Finding the Distribution 24	12.3	Variance of Cost Per Payment With Deductible d
Section 13 Deductibles Applied To The Uniform, Exponential And Pareto Distributions 189 Section 13 Problem Set 197 Section 14 Combined Limit And Deductible 203 14.1 Maximum Covered Loss u Combined With Policy Deductible $d < u$ 203 14.2 PDF and CDF of Y_L and Y_P . 207 14.3 A Few Comments 208 14.4 Graphical Representation of $E[Y_L]$ 209 Section 15 Additional Policy Adjustments 227 15.1 Coinsurance and Copay 221 15.2 Inflation 221 15.3 Policy Deductible d in Combination With Maximum Covered Loss u 234 Section 15 Aggregate Models - Compound Distributions 243 16.1 The Compound Distribution Model for Aggregate Claims in One Period of Time 243 16.2 The Compound Distribution Model for Aggregate Claims in One Period of Time 243 16.2 The Compound Distribution Model for Aggregate Claims in One Period of Time 243 16.2 The Compound Distribution Model for Aggregate Claims in One Period of Time 243 16.2 The Compound Distribution Ot a Compound Distribution Of a Sum of Random Vari	12.4	Franchise Deductible d
Distributions189Section 13 Problem Set	Section 12 P	roblem Set $\ldots \ldots \ldots$
Section 13 Problem Set 197 Section 14 Combined Limit And Deductible 203 14.1 Maximum Covered Loss u Combined With Policy Deductible $d < u$ 203 14.2 PDF and CDF of Y_L and Y_P 207 14.3 A Few Comments 208 14.4 Graphical Representation of $E[Y_L]$ 209 Section 14 Problem Set 210 Section 15 Additional Policy Adjustments 227 15.1 Coinsurance and Copay 229 15.2 Inflation 229 15.3 Policy Deductible d in Combination With Maximum Covered Loss u 231 36ction 15 Poblem Set 231 Section 16 Aggregate Models - Compound Distributions 243 Section 16 Aggregate Models - Compound Distributions 243 16.1 The Compound Distribution Model for Aggregate Claims in One Period of Time 243 16.2 The Compound Poisson Distribution 245 16.3 The Normal Approximation to a Compound Distribution 246 16.4 The Convolution Method for Finding the Distribution 247 Section 16 Problem Set	Section 13	
Section 14 Combined Limit And Deductible 203 14.1 Maximum Covered Loss u Combined With Policy Deductible $d < u$ 203 14.2 PDF and CDF of Y_L and Y_P 207 14.3 A Few Comments 208 14.4 Graphical Representation of $E[Y_L]$ 209 Section 14 Problem Set 210 Section 15 Additional Policy Adjustments 227 15.1 Coinsurance and Copay 229 15.2 Inflation 229 15.3 Policy Deductible d in Combination With Maximum Covered Loss u 231 15.3 Policy Deductible d in Combination With Maximum Covered Loss u 234 Section 15 Problem Set 234 Section 16 Aggregate Models - Compound Distributions 243 16.1 The Compound Distribution Model for Aggregate Claims in One Period of Time 243 16.2 The Compound Poisson Distribution 245 16.3 The Normal Approximation to a Compound Distribution 246 16.4 The Convolution Method for Finding the Distribution 247 16.3 The Normal Approximation to a Compound Distribution 246	~	
14.1 Maximum Covered Loss u Combined With Policy Deductible $d < u$ 203 14.2 PDF and CDF of Y_L and Y_P . 207 14.3 A Few Comments 208 14.4 Graphical Representation of $E[Y_L]$ 209 Section 14 Problem Set 210 Section 15 Additional Policy Adjustments 227 15.1 Coinsurance and Copay 229 15.2 Inflation 229 15.3 Policy Deductible d in Combination With Maximum Covered Loss u 231 Section 15 Problem Set 234 Section 15 Problem Set 234 Section 15 Problem Set 231 Section 15 Problem Set 231 Section 16 Aggregate Models - Compound Distributions 243 16.1 The Compound Distribution Model for Aggregate Claims in One Period of Time 243 16.2 The Compound Poisson Distribution 246 16.3 The Normal Approximation to a Compound Distribution 246 16.4 The Convolution Method for Finding the Distribution 247 Section 16 Problem Set 250	Section 13 P	roblem Set
14.2 PDF and CDF of Y_L and Y_P . 207 14.3 A Few Comments 208 14.4 Graphical Representation of $E[Y_L]$ 209 Section 14 Problem Set 210 Section 15 Additional Policy Adjustments 227 15.1 Coinsurance and Copay 227 15.2 Inflation 229 15.3 Policy Deductible d in Combination With Maximum Covered Loss u 231 3 Section 15 Problem Set 233 Section 16 Aggregate Models - Compound Distributions 243 16.1 The Compound Distribution Model for Aggregate Claims in One Period of Time 243 16.2 The Compound Poisson Distribution 246 16.3 The Normal Approximation to a Compound Distribution 246 16.4 The Convolution Method for Finding the Distribution 247 Section 16 Properties Of The Aggregate Loss 279 17.1 The Individual Risk Model 279	Section 14	Combined Limit And Deductible 203
14.3 A Few Comments 208 14.4 Graphical Representation of $E[Y_L]$ 209 Section 14 Problem Set 210 Section 15 Additional Policy Adjustments 227 15.1 Coinsurance and Copay 227 15.2 Inflation 229 15.3 Policy Deductible d in Combination With Maximum Covered Loss u 231 Section 15 Problem Set 231 Section 16 Aggregate Models - Compound Distributions 243 16.1 The Compound Distribution Model for Aggregate Claims in One Period of Time 243 16.2 The Compound Distribution to a Compound Distribution 246 16.3 The Normal Approximation to a Compound Distribution 246 16.4 The Convolution Method for Finding the Distribution 247 Section 16 Problem Set 250 Section 17 More Properties Of The Aggregate Loss 279 17.1 The Individual Risk Model 279	14.1	Maximum Covered Loss u Combined With Policy Deductible $d < u$ 203
14.4 Graphical Representation of $E[Y_L]$ 209 Section 14 Problem Set 210 Section 15 Additional Policy Adjustments 227 15.1 Coinsurance and Copay 227 15.2 Inflation 229 15.3 Policy Deductible d in Combination With Maximum Covered Loss u 231 15.3 Policy Deductible d in Combination Factor r 231 Section 15 Problem Set 234 Section 16 Aggregate Models - Compound Distributions 243 16.1 The Compound Distribution Model for Aggregate Claims in One Period of Time 243 16.2 The Compound Poisson Distribution 245 16.3 The Normal Approximation to a Compound Distribution 246 16.4 The Convolution Method for Finding the Distribution 247 Section 16 Problem Set 247 Section 16 Problem Set 247 16.3 The Normal Approximation to a Compound Distribution 246 16.4 The Convolution Method for Finding the Distribution 247 Section 16 Problem Set 250 Section 17 More Propertie	14.2	PDF and CDF of Y_L and Y_P
Section 14 Problem Set 210 Section 15 Additional Policy Adjustments 227 15.1 Coinsurance and Copay 227 15.2 Inflation 229 15.3 Policy Deductible d in Combination With Maximum Covered Loss u 231 and Coinsurance Factor α and Inflation Factor r 231 Section 15 Problem Set 234 Section 16 Aggregate Models - Compound Distributions 243 16.1 The Compound Distribution Model for Aggregate Claims in One Period of Time 243 16.2 The Compound Poisson Distribution 245 16.3 The Normal Approximation to a Compound Distribution 246 16.4 The Convolution Method for Finding the Distribution 247 Section 16 Problem Set 247 17.1 The Individual Risk Model 279	14.3	A Few Comments
Section 15 Additional Policy Adjustments 227 15.1 Coinsurance and Copay 229 15.2 Inflation 229 15.3 Policy Deductible d in Combination With Maximum Covered Loss u and Coinsurance Factor α and Inflation Factor r 231 Section 15 Problem Set 234 Section 16 Aggregate Models - Compound Distributions 243 16.1 The Compound Distribution Model for Aggregate Claims in One Period of Time 243 16.2 The Compound Poisson Distribution 245 16.3 The Normal Approximation to a Compound Distribution 246 16.4 The Convolution Method for Finding the Distribution 247 Section 16 Problem Set 247 16.1 The Convolution Method for Finding the Distribution 246 16.3 The Normal Approximation to a Compound Distribution 246 16.4 The Convolution Method for Finding the Distribution 247 Section 16 Problem Set 250 Section 17 More Properties Of The Aggregate Loss 279 17.1 The Individual Risk Model 279	14.4	Graphical Representation of $E[Y_L]$
15.1 Coinsurance and Copay 227 15.2 Inflation 229 15.3 Policy Deductible d in Combination With Maximum Covered Loss u and Coinsurance Factor α and Inflation Factor r 231 Section 15 Problem Set 234 Section 16 Aggregate Models - Compound Distributions 243 16.1 The Compound Distribution Model for Aggregate Claims in One Period of Time 243 16.2 The Compound Poisson Distribution 245 16.3 The Normal Approximation to a Compound Distribution 246 16.4 The Convolution Method for Finding the Distribution 247 Section 16 Problem Set 247 Section 16 Problem Set 270 17.1 The Individual Risk Model 279	Section 14 P	roblem Set
15.2 Inflation 229 15.3 Policy Deductible d in Combination With Maximum Covered Loss u and Coinsurance Factor α and Inflation Factor r 231 Section 15 Problem Set 234 Section 16 Aggregate Models - Compound Distributions 243 16.1 The Compound Distribution Model for Aggregate Claims in One Period of Time 243 16.2 The Compound Poisson Distribution 245 16.3 The Normal Approximation to a Compound Distribution 246 16.4 The Convolution Method for Finding the Distribution 247 Section 16 Problem Set 250 Section 17 More Properties Of The Aggregate Loss 279 17.1 The Individual Risk Model 279	Section 15	Additional Policy Adjustments 227
15.3Policy Deductible d in Combination With Maximum Covered Loss u and Coinsurance Factor α and Inflation Factor r 231Section 15 Problem Set234Section 16Aggregate Models - Compound Distributions24316.1The Compound Distribution Model for Aggregate Claims in One Period of Time24316.2The Compound Poisson Distribution24316.3The Normal Approximation to a Compound Distribution24516.4The Convolution Method for Finding the Distribution of a Sum of Random Variables247Section 16 Problem Set250Section 17More Properties Of The Aggregate Loss27917.1The Individual Risk Model279	15.1	Coinsurance and Copay
and Coinsurance Factor α and Inflation Factor r 231Section 15 Problem Set234Section 16 Aggregate Models - Compound Distributions24316.1The Compound Distribution Model for Aggregate Claims in One Period of Time24316.2The Compound Poisson Distribution24316.3The Normal Approximation to a Compound Distribution24516.4The Convolution Method for Finding the Distribution24616.4The Convolution Method for Finding the Distribution247Section 16 Problem Set250Section 17 More Properties Of The Aggregate Loss27917.1The Individual Risk Model279	15.2	Inflation
Section 15 Problem Set234Section 16Aggregate Models - Compound Distributions24316.1The Compound Distribution Model for Aggregate Claims in One Period of Time	15.3	Policy Deductible d in Combination With Maximum Covered Loss u
Section 16Aggregate Models - Compound Distributions24316.1The Compound Distribution Model for Aggregate Claims in One Period of Time24316.2The Compound Poisson Distribution24316.3The Normal Approximation to a Compound Distribution24516.4The Convolution Method for Finding the Distribution247Section 16 Problem Set247Section 17More Properties Of The Aggregate Loss27917.1The Individual Risk Model279		and Coinsurance Factor α and Inflation Factor r
16.1 The Compound Distribution Model for Aggregate Claims in One Period of Time	Section 15 P	roblem Set $\ldots \ldots 234$
Time 243 16.2 The Compound Poisson Distribution 245 16.3 The Normal Approximation to a Compound Distribution 246 16.4 The Convolution Method for Finding the Distribution 247 Section 16 Problem Set 250 Section 17 More Properties Of The Aggregate Loss 279 17.1 The Individual Risk Model 279	Section 16	Aggregate Models - Compound Distributions 243
16.2 The Compound Poisson Distribution 245 16.3 The Normal Approximation to a Compound Distribution 246 16.4 The Convolution Method for Finding the Distribution 247 Section 16 Problem Set 247 Section 17 More Properties Of The Aggregate Loss 279 17.1 The Individual Risk Model 279	16.1	The Compound Distribution Model for Aggregate Claims in One Period of
16.3 The Normal Approximation to a Compound Distribution		Time
16.4 The Convolution Method for Finding the Distribution of a Sum of Random Variables	16.2	The Compound Poisson Distribution
of a Sum of Random Variables247Section 16 Problem Set250Section 17 More Properties Of The Aggregate Loss27917.1The Individual Risk Model279	16.3	The Normal Approximation to a Compound Distribution
Section 16 Problem Set250Section 17 More Properties Of The Aggregate Loss27917.1The Individual Risk Model279	16.4	The Convolution Method for Finding the Distribution
Section 17More Properties Of The Aggregate Loss27917.1The Individual Risk Model		of a Sum of Random Variables
17.1 The Individual Risk Model	Section 16 P	roblem Set $\ldots \ldots 250$
	Section 17	More Properties Of The Aggregate Loss 279
Section 17 Problem Set	17.1	The Individual Risk Model
	Section 17 P	roblem Set $\ldots \ldots 281$

Section 18	Stop Loss Insurance 295
Section 18 P	roblem Set
Section 19	Risk Measures 315
19.1	Value at Risk
19.2	Tail-Value-at-Risk
Section 19 P	roblem Set
Section 20	Parametric Models Data And Estimation Review 323
20.1	Review of Parametric Distributions
20.2	Data and Estimation
Section 21	Maximum Likelihood Estimation Based On Complete Data 327
Section 21 P	roblem Set
Section 22	Maximum Likelihood Estimation Based On Incomplete Data 341
22.1	MLE Based on Interval Grouped Data
22.2	Maximum Likelihood Estimation Based On Censored and/or Truncated Data343
Section 22 P	roblem Set $\ldots \ldots 350$
Section 23	MLE For The Exponential Distribution 357
23.1	MLE of the Exponential Distribution Based on Complete Data $\ldots \ldots 357$
23.2	MLE of the Exponential Distribution
	Based on Policy Limit (Right-Censored) Data
23.3	MLE of the Exponential Distribution Based on Policy Deductible Data 358
23.4	MLE of the Exponential Distribution
	Based on General Policy Limit and Deductible Data
23.5	Interval Grouped Data and MLE Estimation of the Exponential Distribution 361
Section 23 P	roblem Set
Section 24	MLE For Pareto And Weibull Distributions 375
24.1	Maximum Likelihood Estimation and Transformations
24.2	Pareto Distribution α , θ , where θ is given $\ldots \ldots \ldots \ldots \ldots \ldots 377$
24.3	Single Parameter Pareto Distribution, α, θ , where θ is given $\ldots \ldots 381$
24.4	Weibull Distribution, τ , θ , where τ is given
Section 24 P	roblem Set
Section 25	MLE Applied To FAM Table Distributions 399
25.1	Inverse Exponential Distribution, θ
25.2	Inverse Pareto Distribution, τ , θ , where θ is given
25.3	Inverse Weibull Distribution, τ , θ , where τ is given $\ldots \ldots \ldots \ldots \ldots 402$
25.4	Normal Distribution, μ , σ^2

25.5	Lognormal Distribution, μ , σ^2
25.6	Gamma Distribution, α , θ , where α is given $\ldots \ldots \ldots$
25.7	Inverse Gamma Distribution, α , θ , where α is given
25.8	poisson Distribution λ
25.9	Binomial Distribution m, q
25.10	Negative Binomial Distribution
Section 25 P	roblem Set
Section 26	Limited Fluctuation Credibility 421
26.1	Introductory Comments on Credibility Theory
26.2	The Standard for Full Credibility
26.3	Full credibility applied to a frequency distribution
26.4	The Standard for Full Credibility Applied to Compound Distributions \ldots 428
26.5	Standard for Full Credibility Applied to Poisson Random Variable N 430
26.6	Standard For Full Credibility Applied to a Compound Poisson Distribution 430
26.7	Partial Credibility
Section 26 P	roblem Set
Section 27	Pricing and Reserving Short-Term Insurance 459
Section 27 P	roblem Set
Section 28	Short-Term Insurance Loss Reserving 473
28.1	Basic Elements of Loss Reserving
28.2	Case Reserves Plus and Expected Loss Ratio Methods
28.3	Chain-Ladder (Loss Development Triangle) Method
28.4	Bornhuetter Ferguson Method
28.5	Discounted Reserves
28.6	Additional Notes and Comments On Chapter 3 of the Text
28.6 Section 28 P	-
	-
Section 28 P	roblem Set
Section 28 Pa Section 29	roblem Set
Section 28 P: Section 29 29.1	roblem Set
Section 28 P Section 29 29.1 29.2	roblem Set
Section 28 P Section 29 29.1 29.2 29.3	roblem Set 482 Short-Term Insurance Ratemaking 501 Objectives of Ratemaking 501 Data for Ratemaking 501 Loss-Development and Trend Factors 502
Section 28 P Section 29 29.1 29.2 29.3 29.4	roblem Set 482 Short-Term Insurance Ratemaking 501 Objectives of Ratemaking 501 Data for Ratemaking 501 Loss-Development and Trend Factors 502 Expenses, Profit and Contingency Loading 504
Section 28 P Section 29 29.1 29.2 29.3 29.4 29.5	roblem Set 482 Short-Term Insurance Ratemaking 501 Objectives of Ratemaking 501 Data for Ratemaking 501 Loss-Development and Trend Factors 502 Expenses, Profit and Contingency Loading 503 Credibility Factors 505

Section 30 Reinsurance

Section 30 P	roblem Set $\ldots \ldots \ldots$
Section 31	Option Pricing 522
31.1	Arbitrage
31.2	Introduction to Options
31.3	The Binomial Option Pricing Model For One Period
31.4	The Binomial Option Pricing Model For Two Periods
31.5	The Black-Scholes-Merton Option Pricing Model
Section 31 P	roblem Set $\ldots \ldots \ldots$
Practice Ex	cam 1 54'
Solutions to	Practice Exam 1
Practice Ex	
Solutions to	Practice Exam 2
Practice Ex	cam 3 573
Solutions to	Practice Exam 3
Practice Ex	tam 4 58'
Solutions to	Practice Exam 4 $\dots \dots $
Practice Ex	cam 5 603
Solutions to	Practice Exam 5
Index	617

INTRODUCTORY COMMENTS

The FAM exam is divided almost equally into FAM-S and FAM-L topics. This study manual is designed to help in the preparation for the FAM-S part of the Society of Actuaries FAM Exam.

The first part of this manual consists of a summary of notes, illustrative examples and problem sets with detailed solutions. The second part consists of 5 practice exams. The SOA exam syllabus for the FAM exam indicates that the exam is 3.5 hours in length with 34 multiple choice questions. The practice exams in this manual each have 17 questions, reflecting the fact that FAM-S is 50% of the full FAM exam. The appropriate time for the 17 question FAM-S practice exams in this manual is one hour and forty-five minutes.

The level of difficulty of the practice exam questions has been designed to be similar to those on past exams covering the same topics. The practice exam questions are not from old SOA exams.

I have attempted to be thorough in the coverage of the topics upon which the exam is based, and consistent with the notation and content of the official references. I have been, perhaps, more thorough than necessary on reviewing maximum likelihood estimation.

Because of the time constraint on the exam, a crucial aspect of exam taking is the ability to work quickly. I believe that working through many problems and examples is a good way to build up the speed at which you work. It can also be worthwhile to work through problems that you have been done before, as this helps to reinforce familiarity, understanding and confidence. Working many problems will also help in being able to more quickly identify topic and question types. I have attempted, wherever possible, to emphasize shortcuts and efficient and systematic ways of setting up solutions. There are also occasional comments on interpretation of the language used in some exam questions. While the focus of the study manual is on exam preparation, from time to time there will be comments on underlying theory in places that I feel those comments may provide useful insight into a topic.

The notes and examples are divided into 31 sections of varying lengths, with some suggested time frames for covering the material. There are almost 180 examples in the notes and over 440 exercises in the problem sets, all with detailed solutions. The 5 practice exams have 17 questions each, also with detailed solutions. Some of the examples and exercises are taken from previous SOA exams. Some of the problem set exercises are more in depth than actual exam questions, but the practice exam questions have been created in an attempt to replicate the level of depth and difficulty of actual exam questions. In total there are almost 700 examples/problems/sample exam questions with detailed solutions. ACTEX gratefully acknowledges the SOA for allowing the use of their exam problems in this study manual.

I suggest that you work through the study manual by studying a section of notes and then attempting the exercises in the problem set that follows that section. The order of the sections of notes is the order that I recommend in covering the material, although the material on short-term insurance pricing and reserving in Sections 27 to 30 and option pricing in Section 31 is independent of the other material on the exam. The order of topics in this manual is not the same as the order presented on the exam syllabus.

It has been my intention to make this study manual self-contained and comprehensive, however it is important to be familiar with original reference material on all topics.

While the ability to derive formulas used on the exam is usually not the focus of an exam question, it is useful in enhancing the understanding of the material and may be helpful in memorizing formulas. There may be an occasional reference in the review notes to a derivation, but you are encouraged to review the official reference material for more detail on formula derivations.

In order for the review notes in this study manual to be most effective, you should have some background at the junior or senior college level in probability and statistics. It will be assumed that you are reasonably familiar with differential and integral calculus. The prerequisite concepts to modeling and model estimation are reviewed in this study manual. The study manual begins with a detailed review of probability distribution concepts such as distribution function, hazard rate, expectation and variance. Of the various calculators that are allowed for use on the exam, I am most familiar with the BA II PLUS. It has several easily accessible memories. The TI-30X IIS has the advantage of a multi-line display. Both have the functionality needed for the exam.

There is a set of tables that has been provided with the exam in past sittings. These tables consist of some detailed description of a number of probability distributions along with tables for the standard normal and chi-squared distributions. The tables can be downloaded from the SOA website www.soa.org.

If you have any questions, comments, criticisms or compliments regarding this study manual, please contact the publisher ACTEX, or you may contact me directly at the address below. I apologize in advance for any errors, typographical or otherwise, that you might find, and it would be greatly appreciated if you would bring them to my attention. ACTEX will be maintaining a website for errata that can be accessed from https://actexlearning.com/errata. It is my sincere hope that you find this study manual helpful and useful in your preparation for the exam. I wish you the best of luck on the exam.

Samuel A. Broverman			September 2024
Department of Statistical	Sciences	www	v.sambroverman.com
University of Toronto	E-mail: sam.broverman@utoronto.ca	or	2brove@rogers.com

Section 7

Mixture Of Two Distributions

The material in this section relates to Section 4.2.3 of "Loss Models". The suggested time frame for this section is 2 hours. The topic of distribution mixtures is not mentioned in the learning objectives for the FAM exam but all of Chapter 4 of "Loss Models" is listed in the reference reading.

7.1 Mixture of Two Distributions

We begin with a formal algebraic definition of how a mixture of two distributions is constructed, and later we will look at how a mixture distribution is described by general reasoning.

Given random variables X_1 and X_2 , with pdf's or pf's $f_{X_1}(x)$ and $f_{X_2}(x)$, and given 0 < a < 1, we construct the random variable Y with pdf

$$f_Y(y) = a \times f_{X_1}(y) + (1-a) \times f_{X_2}(y)$$
(7.1)

Y is called a mixture distribution or a two-point mixture of the distributions of X_1 and X_2 .

The two-point mixture random variable Y can also be defined in terms of the cdf,

$$F_{Y}(y) = a \times F_{X_{1}}(y) + (1 - a) \times F_{X_{2}}(y)$$
(7.2)

 X_1 and X_2 are the **component distributions** of the mixture, and the factors a and 1-a are referred to as **mixing weights**. It is important to understand that we are **not adding** aX_1 and $(1-a)X_2$, Y **is not** a sum of random variables. Y is defined in terms of a pdf (or cdf) that is a weighted average of the pdf's (or cdf's) of X_1 and X_2 . We are adding af_{X_1} and $(1-a)f_{X_2}$ to get f_Y .

Example 7.1. As a simple illustration of a mixture distribution, consider two bowls. Bowl A has 5 balls with the number 1 on them and 5 balls with the number 2 on them, and bowl B has 3 balls with the number 1 and 7 balls with the number 2. Let X_1 denote the number on a ball randomly chosen from bowl A, and let X_2 denote the number on a ball randomly chosen from bowl B. The probability functions of X_1 and X_2 are $f_{X_1}(1) = f_{X_1}(2) = 0.5$ and $f_{X_2}(1) = 0.3$, $f_{X_2}(2) = 0.7$.

Suppose we create the mixture distribution with mixing weights a = 0.5 and 1 - a = 0.5. The mixture distribution Y has probability function $f_Y(1) = 0.5 \times 0.5 + 0.5 \times 0.3 = 0.4$, $f_Y(2) = 0.5 \times 0.5 + 0.5 \times 0.7 = 0.6$.

Note that the outcomes (ball numbers) of the mixture distribution Y come from the possible outcomes of the component distributions X_1 and X_2 .

An alternative way of looking at this mixture distribution is by means of conditioning on a "parameter". This will be important when we look at continuous mixing. The parameter approach to describe the mixture distribution in Example 7.1 is as follows.

Suppose that a fair coin is tossed. If the toss is a head, a ball is chosen at random from bowl A and if the toss is a tail, a ball is chosen at random from bowl B. We define the random variable Z to be the number on the ball. We will see that Z has the same distribution as the mixture distribution labeled Y above.

The random variable Z can be interpreted as follows. Consider the 2-point random variable Θ , for which $\Theta =$ Bowl A if the coin toss is a head, and $\Theta =$ Bowl B if the toss is a tail. Then $P(\Theta = A) = P(\Theta = B) = .5$ (since the coin is fair). Θ is used to indicate which bowl the ball will be chosen from depending on the outcome of the coin toss.

If the toss is a head, the bowl is A, and then Z has the X_1 distribution for the number on the ball, so $f_{X_1}(1) = P(Z = 1|\Theta = A) = 0.5$ and $f_{X_1}(2) = P(Z = 2|\Theta = A) = 0.5$. In a similar way, if the toss is a tail, the bowl is B, and then Z has the X_2 distribution for the number on the ball, so $f_{X_2}(1) = P(Z = 1|\Theta = B) = 0.3$ and $f_{X_2}(2) = P(Z = 2|\Theta = B) = 0.7$.

Z is described as a combination of two conditional distributions based on the parameter $\Theta.$

To find the overall, or unconditional distribution of Z, we use some basic rules of probability. Since Θ must be A or B, we can think of bowl B as the "complement" of bowl A, and then

$$P(Z = 1) = P[(Z = 1) \cap (\Theta = A)] + P[(Z = 1) \cap (\Theta = B)]$$

= $P(Z = 1|\Theta = A) \times P(\Theta = A) + P(Z = 1|\Theta = B) \times P(\Theta = B)$
= $0.5 \times 0.5 + 0.5 \times 0.3 = 0.4$

We have used the rule $P(C) = P[C \cap D] + P[C \cap D'] = P(C|D) \times P(D) + P(C|D') \times P(D').$

This shows that the distribution of Z is the same as the mixture distribution Y in Example 7.1. The mixing weights for the two bowls are the probabilities of the coin indicating bowl A or bowl B.

Language used on exam questions that identifies a mixture distribution

There is some typical language that is used in exam questions that indicates that a mixture of distributions is being considered. This language will be illustrated using the distributions involved in the bowl example.

Suppose that we are told that there are two types of individuals. Type A individuals have a mortality probability of .5 (and survival probability of .5) in the coming year, and Type B individuals have a mortality probability of .3 in the coming year. In a large group of these individuals, 50% are Type A and 50% are Type B. An individual is chosen at random from the group. We want to find this randomly chosen individual's mortality probability.

We can use the usual rules of conditional probability to formulate this probability, just as above:

$$\begin{aligned} P(\text{dying this year}) &= P(\text{dying} \cap \text{Type A}) + P(\text{dying} \cap \text{Type B}) \\ &= P(\text{dying}|\text{Type A}) \times P(\text{Type A}) + P(\text{dying} \cap \text{Type B}) \times P(\text{Type B}) \\ &= 0.5 \times 0.5 + 0.5 \times 0.3 = 0.4. \end{aligned}$$

This is exactly $f_Y(1) = f_{X_1}(1) \times a + f_{X_2}(1) \times (1-a)$ where "1" corresponds to the event of dying within the year. Note that the phrase "50% are Type A" is interpreted as meaning that if an individual is chosen from the large group, the probability of being Type A is 0.5. This is language that is often used in exam questions to indicate a mixture.

Mixture of a discrete and a continuous distribution

In Section 2 of this study guide we looked at "mixed distributions". In that section, a mixed distribution referred to a random variable that was continuous on part of its probability space and also had one or more discrete points. The concept of mixed distribution just introduced in this section can be used to describe the Section 2 type of mixed distribution. The following example uses the example from Section 2 to illustrate this.

Example 7.2. Suppose that X is defined in the following piecewise way: X has a discrete point of probability of .5 at X = 0, and on the interval (0,1) X is a continuous random variable with density function f(x) = x for 0 < x < 1, and X has no density or probability elsewhere. Show that this random variable can be described as a mixture of two distributions with appropriate definitions for component distributions X_1 , X_2 and mixing weights a and 1 - a.

Solution.

Let $X_1 = 0$ be a constant (not actually a random variable, but is sometimes called a degenerate random variable), so that $f_{X_1}(0) = P(X_1 = 0) = 1$ and $f_{X_1}(x) = 0$ for $x \neq 0$.

Let X_2 be continuous on (0,1) with pdf $f_{X_2}(x) = 2x$.

With mixing weights of a = 0.5 and 1 - a = 0.5, using the definition of the mixture of two distributions, we have the mixed random variable Y satisfying

 $f_Y(0) = a \times f_{X_1}(0) + (1-a) \times f_{X_2}(0) = (0.5 \times 1) + 0 = 0.5, \text{ and}$ $f_Y(x) = a \times f_{X_1}(x) + (1-a) \times f_{X_2}(x) = 0.5 \times 0 + 0.5 \times 2x = x \text{ for } 0 < x < 1.$ Y has the same distribution as the original X.

We should be a little careful about the situation in Example 7.2. When we are mixing discrete probability points with a continuous density, for any particular discrete point we always assign a probability of 0 at that point for any continuous component distribution. For instance, in Example 7.2, if X_1 was at the single point 0.4, then $f_{X_1}(0.4) = 1$ and $f_{X_1}(x) = 0$ elsewhere, and for the mixture distribution, then $f_Y(0.4) = 0.5 \times 1 = 0.5$ (we do not add $0.5 \times f_{X_2}(0.4)$).

Some important relationships for mixture distributions

We have already seen the defining relationships

$$f_Y(y) = a f_{X_1}(y) + (1-a) f_{X_2}(y)$$

and

$$F_Y(y) = aF_{X_1}(y) + (1-a)F_{X_2}(y).$$

We can interpret these relationships by saying that the pdf and cdf of Y are weighted averages of the component pdf's and cdf's. This weighted average interpretation can be applied to a number of other distribution related quantities.

- if C is any event related to Y, then $P(C) = a \times P_{X_1}(C) + (1-a) \times P_{X_2}(C);$ (7.3) $P_{X_1}(C)$ is the event probability based on random variable X_1 , and the same for X_2
- If g is any function (that doesn't involve the parameters of Y), then $E[q(Y)] = a \times E[q(X_1)] + (1-a) \times E[q(X_2)]$ (7.4)

The justification for this relationship follows from the form of the mixed density;

$$E[g(Y)] = \int g(y) f_Y(y) dy = \int g(y) \times [a \times f_{X_1}(y) + (1-a) \times f_{X_2}(y)] dy$$

= $a \times \int g(y) \times f_{X_1}(y) dy + (1-a) \times \int g(y) \times f_{X_2}(y) dy$
= $a \times E[g(X_1)] + (1-a) \times E[g(X_2)]$

Some of the important examples of these weighted average relationships are:

Interval Probability: the event C is $c < Y \leq d$

$$P(c < Y \le d) = a \times P(c < X_1 \le d) + (1 - a) \times P(c < X_2 \le d)$$
(7.5)

kth moment of Y: the function g is $g(Y) = Y^k$

$$E[Y^{k}] = a \times E[X_{1}^{k}] + (1 - a) \times E[X_{2}^{k}]$$
(7.6)

7.2 Formulating a Mixture Distribution as a Combination of Conditional Distributions

The relationships above can also be formulated as applications of probability rules involving conditional probability and conditional expectation.

(i) For any random variable Y and event D,

$$f_Y(y) = f(y|D) \times P(D) + f(y|D') \times P(D')$$
(7.7)

(ii) For any events C and D,

$$P[C] = P[C \cap D] + P[C \cap D'] = P[C|D] \times P[D] + P[C|D'] \times P[D'].$$
(7.8)

(iii) For any random variable Y and any event D,

$$E[Y] = E[Y|D] \times P[D] + E[Y|D'] \times P[D']$$
(7.9)

We can define the random variable Θ to be 1 or 2, indicating whether or not event D has occurred, so $\Theta = 1 \equiv D$ has occurred, and $\Theta = 2 \equiv D'$ has not occurred.

We can think of X_1 as the conditional distribution of Y given $\Theta = 1$ (event D) and we can think of X_2 as the conditional distribution of Y given $\Theta = 2$ (event D'), so that

$$f(y|D) = f(y|\Theta = 1) = f_{X_1}(y)$$
 and $f(y|D') = f(y|\Theta = 2) = f_{X_2}(y)$.

The cdf, pdf and expectation of Y are adaptations of the conditional probability rules above. For instance, the mean of the mixed distribution is the "mixture of the means" of the component distributions.

$$E[Y] = E[Y|\Theta = 1] \times P[\Theta = 1] + E[Y|\Theta = 2] \times P[\Theta = 2] = E[X_1] \times a + E[X_2] \times (1-a).$$

~

Example 7.3. A collection of insurance policies consists of two types. 25% of policies are Type 1 and 75% of policies are Type 2. For a policy of Type 1, the loss amount per year follows an exponential distribution with mean 200, and for a policy of Type 2, the loss amount per year follows a Pareto distribution with parameters $\alpha = 3$ and $\theta = 200$. For a policy chosen at random from the entire collection of both types of policies, find the probability that the annual loss will be less than 100, and find the average loss.

Solution.

The two types of losses are the random variables X_1 and X_2 .

 X_1 has an exponential distribution with mean 200, so it has cdf $F_{X_1}(x) = 1 - e^{-x/200}$.

$$X_2$$
 has cdf $F_{X_2}(x) = 1 - \left(\frac{\theta}{x+\theta}\right)^{\alpha} = 1 - \left(\frac{200}{x+200}\right)^3$.

These are the component distributions. The mixing weights are the proportions of policies of each type, so that a = 0.25 (the proportion of Type 1 policies) and 1 - a = 0.75. The loss random variable Y of the randomly chosen policy has a distribution that is the mixture of X_1 and X_2 . In this context, "randomly chosen" is taken to mean that the probabilities of choosing a Type 1 or Type 2 policy are the proportions of those policy types in the full collection of policies.

The cdf of Y is

$$F_Y(y) = a \times F_{X_1}(y) + (1-a) \times F_{X_2}(y)$$

= 0.25 × (1 - e^{-y/200}) + 0.75 × $\left(1 - \left(\frac{200}{y+200}\right)^3\right)$

Then, $F_Y(100) = 0.626$.

The mean of Y is $0.25 \times E[X_1] + 0.75 \times E[X_2] = 0.25 \times 200 + 0.75 \times \frac{200}{3-1} = 125$.

7.3 The Variance of a Mixed Distribution IS (usually) NOT the Weighted Average of the Variances

For a mixture of two random variables, the weighted average pattern applies to most quantities for the mixed distribution.

One important exception to this rule is the formulation of the variance of the mixture. If $Y \cdot P$ is any random variable, then we can formulate the variance of Y as $Var[Y] = E[Y^2] - (E[Y])^2$.

If Y is the mixture of X_1 and X_2 with mixing weights a and 1-a, then we know that $E[Y] = a \times E[X_1] + (1-a) \times E[X_2]$ and $E[Y^2] = a \times E[X_1^2] + (1-a) \times E[X_2^2]$.

It is tempting to guess that Var[Y] is the weighted average of $Var[X_1]$ and $Var[X_2]$.

This is <u>not true</u> in general (we will see a special case later for which it is true).

Example 7.4. Find the variance of the loss on the policy chosen at random in Example 7.3, and compare it to the weighted average of the variances of the two component loss distributions.

Solution.

The mean of Y is $0.25 \times E[X_1] + 0.75 \times E[X_2] = 0.25 \times 200 + 0.75 \times \frac{200}{3-1} = 125$, and the second moment of Y is $0.25 \times E[X_1^2] + 0.75 \times E[X_2^2] = 0.25 \times 2 \times 200^2 + 0.75 \times \frac{200^2 \times 2}{(3-1)(3-2)} = 50,000$, so $Var[Y] = E[Y^2] - (E[Y])^2 = 34,375$. $Var[X_1] = 40,000$ (variance of an exponential distribution is the square of the mean). $Var[X_2] = \frac{200^2 \times 2}{(3-1)(3-2)} - 100^2 = 30,000$. The weighted average of the two variances is $0.25 \times 40,000 + 0.75 \times 30,000 = 32,500$, which is not the variance of Y.

Weighted average does not apply to percentiles in a mixture distribution In Example 7.3, the 50th percentile of X_1 is m_1 , the solution of the equation $F_{X_1}(m_1) = 1 - e^{-m_1/200} = 0.5$, so that $m_1 = 138.63$.

The 50th percentile of X_2 is m_2 , the solution of $F_{X_2}(m_2) = 1 - \left(\frac{200}{m_2 + 200}\right)^3 = 0.5$, so that $m_2 = 51.98$.

The weighted average of the two 50th percentiles is $0.25 \times 138.63 + 0.75 \times 51.98 = 73.64$. The cdf of the mixed distribution Y is $F_Y(y) = 0.25 \times F_{X_1}(y) + 0.75 \times F_{X_2}(y)$, and $F_Y(73.64) = 0.53$, so 73.64 is not the 50th percentile of Y.

To find the 50th percentile, say m, of the mixed distribution in Example 7.3, we must solve the equation

$$50 = F_Y(m) = 0.25 \times F_{X_1}(m) + 0.75 \times F_{X_2}(m)$$
$$= 0.25 \times (1 - e^{-m/200}) + 0.75 \times \left(1 - \left(\frac{200}{m + 200}\right)^3\right).$$

There is no algebraic solution, and m would have to be found by a numerical approximation method. The 50th percentile turns out to be about 65.7.

Section 7 Problem Set

Mixture Of Two Distributions

- 1. A portfolio of insurance policies is divided into low and high risk policies. 80% of the policies are low risk and the rest are high risk. The annual number of claims on a low risk policy has a Poisson distribution with a mean of .25 and the annual number of claims on a high risk policy has a Poisson distribution with a mean of 2. A policy is randomly chosen from the portfolio.
 - (a) Find the mean and variance of the number of annual claims for the policy.
 - (b) Find the probability that the policy has at most 1 claim in the coming year.
- 2. \checkmark The distribution of a loss, X, is a two-point mixture:
 - (i) With probability 0.8, X has a two-parameter Pareto distribution with $\alpha=2$ and $\theta=100.$
 - (ii) With probability 0.2, X has a two-parameter Pareto distribution with $\alpha=4$ and $\theta=3000.$

Calculate $Pr(X \le 200)$.

$$(A) 0.76 (B) 0.79 (C) 0.82 (D) 0.85 (E) 0.88 (E) 0.88$$

3. \checkmark The random variable N has a mixed distribution:

- (i) With probability p, N has a binomial distribution with q = 0.5 and m = 2.
- (ii) With probability 1 p, N has a binomial distribution with q = 0.5 and m = 4.

Which of the following is a correct expression for $\operatorname{Prob}(N=2)$?

- (A) $0.125p^2$ (B) 0.375 + 0.125p
- (C) $0.375 + 0.125p^2$ (D) $0.375 0.125p^2$
- (E) 0.375-0.125p

4. • Y is a mixture of two exponential distributions, $f_Y(y) = \frac{1}{2}e^{-y} + \frac{1}{6}e^{-y/3}$. The random variable Z defined by the equation Z = 2Y. Z is a mixture of two exponentials. The means of those two exponential distributions are

- (A) 1 and 3 (B) 1 and 6
- (C) 2 and 3 (D) 2 and 6
- (E) 3 and 6
- 5. \checkmark X_1 has a uniform distribution on the interval (0,1000) and X_2 has a uniform distribution on the interval (0,2000). Y is defined as a mixture of X_1 and X_2 with mixing weights of .5 for each mixture component. Find the pdf, cdf and median (50th percentile) of Y.

- 6. Solution of people aged 50 consists of twice as many non-smokers as smokers. Non-smokers at age 50 have a mortality probability of .1 and smokers at age 50 have a mortality probability of .2. Two 50-year old individuals are chosen at random from the population.
 - (a) Find the probability that at least one of them dies before age 51.
 - (b) Suppose that the mortality probabilities for smokers and non-smokers remain the same at age 51. Find the mortality probability of a randomly chosen survivor at age 51 in this population.
- 7. \checkmark Y is the mixture of an exponential random variable with mean 1 and mixing weight $\frac{2}{3}$, and an exponential distribution with mean 2 and mixing weight $\frac{1}{3}$.

Find the pdf, cdf, mean, variance and 90th percentile of Y.

- 8. V Which of the following statements are true?
 - I. A mixture of two different exponentials with the mixture having a mean 2 has a heavier right tail than a single exponential distribution with mean 2.
 - II. If $f_Y(y) = a_1 \times f_{X_1}(y) + a_2 \times f_{X_2}(y)$, where $0 < a_1 < 1$ and $0 < a_2 < 1$, then $e_Y(d) = a_1 \times e_{X_1}(d) + a_2 \times e_{X_2}(d)$.
- 9. \checkmark You are given two independent estimates of an unknown quantity, μ :
 - (i) Estimate A: $E(\mu_A) = 1000$ and $\sigma(\mu_A) = 400$
 - (ii) Estimate B: $E(\mu_B) = 1200$ and $\sigma(\mu_B) = 200$

Estimate C is a weighted average of the two estimates A and B, such that:

$$\mu_C = w \times \mu_A + (1 - w) \times \mu_B$$

Determine the value of w that minimizes $\sigma(\mu_C)$.

- (A) 0 (B) 1/5 (C) 1/4 (D) 1/3 (E) 1/2
- 10. You are given the claim count data for which the sample mean is roughly equal to the sample variance. Thus you would like to use a claim count model that has its mean equal to its variance. An obvious choice is the Poisson distribution. Determine which of the following models may also be appropriate.
 - (A) A mixture of two binomial distributions with different means.
 - (B) A mixture of two Poisson distributions with different means.
 - (C) A mixture of two negative binomial distributions with different means.
 - (D) None of A, B, or C
 - (E) All of A, B, and C

- 11. Cosses come from an equally weighted mixture of an exponential distribution with mean m_1 , and an exponential distribution with mean m_2 . Determine the least upper bound for the coefficient of variation of this distribution.
 - (A) 1 (B) $\sqrt{2}$ (C) $\sqrt{3}$ (D) 2 (E) $\sqrt{5}$

12. X has the following distribution : Pr[X = 0] = 0.4, Pr[X = 1] = 0.6. The distribution of Y is conditional on the value of X: if X = 0, then the distribution of Y is Pr[Y = 0] = 0.6, Pr[Y = 1] = 0.2, Pr[Y = 2] = 0.2, and if X = 1, then the distribution of Y is Pr[Y = 0] = 0.2, Pr[Y = 1] = 0.3, Pr[Y = 2] = 0.5. Z is the sum of Y independent normal random variables, each with mean and variance 2.

What is Var[Z]?

- (A) 5.0 (B) 6.0 (C) 7.0 (D) 8.0 (E) 9.0
- 13. Subway trains arrive at your station at a Poisson rate of 20 per hour. 25% of the trains are express and 75% are local. The types and number of trains arriving are independent. An express gets you to work in 16 minutes and a local gets you there in 28 minutes. You always take the first train to arrive. Your co-worker always takes the first express. You are both waiting at the same station.

Calculate the conditional probability that you arrive at work before your co-worker, given that a local arrives first.

(A)
$$37\%$$
 (B) 40% (C) 43% (D) 46% (E) 49%

Section 7 Problem Set Solutions

1. (a) Y is the mixture of two Poisson distributions. $E[Y] = 0.8 \times E[X_L] + 0.2 \times E[X_H] = 0.8 \times 0.250 + 0.2 \times 2 = 0.6$. To find the second moment of Y, we recall that for a Poisson distribution with mean λ , the variance is also λ . Since $Var[X] = E[X^2] - (E[X])^2$, it follows that $E[X^2] = Var[X] + (E[X])^2 = \lambda + \lambda^2$ for a Poisson distribution. Then, $E[Y^2] = 0.8 \times E[X_L^2] + 0.2 \times E[X_H^2] = 0.8 \times (0.25 + 0.25^2) + 0.2 \times (2 + 2^2) = 1.45$. Then $Var[Y] = 1.45 - (0.6)^2 = 1.09$.

(b)
$$P(Y \le 1) = 0.8 \times P(X_1 \le 1) + 0.2 \times P(X_2 \le 1)$$

= $0.8 \times [e^{-0.25} + 0.25e^{-0.25}] + 0.2 \times [e^{-2} + 2e^{-2}] = 0.860.$

2. The probability is a mixture of the probabilities for the two components. $P(X \le 200) = 0.8 \times P(X_1 \le 200) + 0.2 \times P(X_2 \le 200)$, where X_1 and X_2 are the two Pareto distributions.

$$P(X_1 \le 200) = 1 - \left(\frac{100}{200 + 100}\right)^2 = .8889, \text{ and}$$

$$P(X_2 \le 200) = 1 - \left(\frac{3000}{200 + 3000}\right)^4 = .2275.$$

$$P(X \le 200) = (0.8 \times 0.8889) + (0.2 \times 0.2275) = 0.757.$$

Answer A

3.
$$P(N = 2) = p \times P(N_1 = 2) + (1 - p) \times P(N_2 = 2)$$

= $p \times (0.5)^2 + (1 - p) \times 6 \times (0.5)^4 = 0.375 - 0.125p.$

We have used the binomial probabilities of the form $\binom{m}{k}q^k(1-q)^{m-k}$.

Answer E

4.
$$f_Y(y) = \frac{1}{2} \times e^{-y} + \frac{1}{2} \times \frac{1}{3} e^{-y/3}$$
. $F_Y(y) = \frac{1}{2} \times (1 - e^{-y}) + \frac{1}{2} \times (1 - e^{-y/3})$.
 $F_Z(z) = F_Y(\frac{z}{2}) = \frac{1}{2} \times (1 - e^{-z/2}) + \frac{1}{2} \times (1 - e^{-z/6})$.
Z is a mixture of two exponentials with means 2 and 6.

Answer D

5.
$$f_Y(y) = 0.5 \times f_{X_1}(y) + 0.5 \times f_{X_2}(y)$$

$$= \begin{cases} 0.5 \times 0.001 + 0.5 \times 0.0005 = 0.00075 & \text{if } 0 < y < 1000 \\ 0.5 \times 0.0005 = 0.00025 & \text{if } 1000 \le y < 2000 \end{cases}$$
 $F_Y(y) = 0.5 \times F_{X_1}(y) + 0.5 \times F_{X_2}(y)$

$$= \begin{cases} 0.5 \times 0.001x + 0.5 \times 0.0005x = 0.00075x & \text{if } 0 < y < 1000 \\ 0.5 \times 1 + 0.5 \times 0.0005x = 0.5 + 0.00025x & \text{if } 1000 \le y < 2000 \end{cases}$$

The median of Y, m, must satisfy the equation $F_Y(m) = 0.5$. We see that at y = 1000, $F_Y(1000) = 0.75$. Therefore, m < 1000, so that $F_Y(m) = 0.00075m = 0.5$. Therefore, m = 666.67. Note that the mean of Y is $0.5 \times 500 + 0.5 \times 1000 = 750$. 6. For a randomly chosen individual at age 50, the mortality probability is the mixture of the mortality probabilities for non-smokers and smokers. This is $q = \frac{2}{3} \times 0.1 + \frac{1}{3} \times 0.2 = \frac{2}{15}$.

The probability that both of two independent 50-year old individuals survive the year is $(1 - \frac{2}{15})^2 = 0.7511$. The probability at least one of them dies by age 51 is 1 - 0.7511 = 0.25.

Suppose that there are 1000 non-smokers and 500 smokers at age 50. The expected number of surviving non-smokers at age 51 is $1000 \times 0.9 = 900$, and the number of surviving smokers is $500 \times 0.8 = 400$.

A randomly chosen survivor at age 51 has a $\frac{9}{13}$ chance of being a non-smoker and a $\frac{4}{13}$ chance of being a smoker.

The mortality probability at age 51 for the randomly chosen survivor is $\frac{9}{13} \times 0.1 + \frac{4}{13} \times 0.2 = .131$.

7. $f_Y(y) = \frac{2}{3} \times e^{-y} + \frac{1}{3} \times \frac{1}{2} e^{-y/2}$. $F_Y(y) = \frac{2}{3} \times (1 - e^{-y}) + \frac{1}{3} \times (1 - e^{-y/2})$. $E[Y] = \frac{2}{3} \times 1 + \frac{1}{3} \times 2 = \frac{4}{3}$. $E[Y^2] = \frac{2}{3} \times (2 \times 1^2) + \frac{1}{3} \times (2 \times 2^2) = 4 \rightarrow Var[Y] = 4 - (\frac{4}{3})^2 = \frac{20}{9}$.

The 90th percentile of Y is c, which must satisfy the equation $F_Y(c) = \frac{2}{3} \times (1 - e^{-c}) + \frac{1}{3} \times (1 - e^{-c/2}) = 0.9.$

If we let $r = e^{-c/2}$, then this becomes a quadratic equation in r, $\frac{2}{3} \times (1 - r^2) + \frac{1}{3} \times (1 - r) = 0.9$, or equivalently, $2r^2 + r - .3 = 0$.

Solving for r results in r = .21 or r = -0.71.

We ignore the negative root, since $r = e^{-c/2}$ must be positive.

Then $c = -2 \ln r = -2 \ln .21 = 3.12$ is the 90th percentile of Y.

8. I. If $f_Y(y) = a_1 \times f_{X_1}(y) + a_2 \times f_{X_2}(y) = a_1 \times \frac{1}{\theta_1} e^{-y/\theta_1} + a_2 \times \frac{1}{\theta_2} e^{-y/\theta_2}$, and $a_1\theta_1 + a_2\theta_2 = 2$. Then $\theta_1 < 2 < \theta_2$ (or vice-versa).

For an exponential distribution with mean 2, we have $f_Z(y) = \frac{1}{2}e^{-y/2}$ $\frac{f_Y(y)}{f_Z(y)} = \frac{a_1 \times \frac{1}{\theta_1}e^{-y/\theta_1} + a_2 \times \frac{1}{\theta_2}e^{-y/\theta_2}}{\frac{1}{2}e^{-y/2}} = \frac{2a_1}{\theta_1}e^{y(\frac{1}{2} - \frac{1}{\theta_1})} + \frac{2a_2}{\theta_2}e^{y(\frac{1}{2} - \frac{1}{\theta_2})}.$ Since $\theta_1 < 2 < \theta_2$, it follows that $\frac{1}{\theta_2} < \frac{1}{2} < \frac{1}{\theta_1}$. Therefore, $\frac{1}{2} - \frac{1}{\theta_2} > 0$ so that $\lim_{y \to \infty} \frac{f_Y(y)}{f_Z(y)} = \lim_{y \to \infty} \frac{2a_1}{\theta_1}e^{y(\frac{1}{2} - \frac{1}{\theta_1})} + \frac{2a_2}{\theta_2}e^{y(\frac{1}{2} - \frac{1}{\theta_2})} = 0 + \infty$. True.

II. Suppose that $f_Y(y) = \frac{1}{2}e^{-y} + \frac{1}{6}e^{-y/3}$. Then $S_Y(y) = \frac{1}{2}e^{-y} + \frac{1}{2}e^{-y/3}$, so that Y is mixture of exponentials with means of 1 and 3, with equal mixing weights of $\frac{1}{2}$. $e_Y(d) = \frac{\int_d^{\infty} S(y) \, dy}{\int_d^{\infty} S(y) \, dy} = \frac{\frac{1}{2}e^{-d} + \frac{3}{2}e^{-d/3}}{\frac{1}{2}e^{-d/3}}$

$$e_Y(d) = \frac{ba}{S(d)} = \frac{\frac{1}{2}e^{-d} + \frac{1}{2}e^{-d/3}}{\frac{1}{2}e^{-d} + \frac{1}{2}e^{-d/3}}.$$

$$e_{X_1}(d) = 1 \ , \ e_{X_3}(d) = 3 \ \to \ \frac{1}{2} \times e_{X_1}(d) + \frac{1}{2} \times e_{X_3}(d) = 2.$$
 False.

9. Since the two estimates are independent,

$$Var[\mu_C] = Var[w \times \mu_A + (1 - w) \times \mu_B] = w^2 \times Var[\mu_A] + (1 - w)^2 \times Var[\mu_B]$$

= $w^2 \times 400^2 + (1 - w)^2 \times 200^2 = 200,000w^2 - 80,000w + 40,000.$

This is a quadratic expression in w with positive coefficient of w^2 . The minimum can be found by differentiating with respect to w and setting equal to 0:

 $400,000w - 80,000 = 0 \rightarrow w = \frac{1}{5}$ is the value of w that minimizes $Var[\mu_c]$.

Answer B

10. If Y is a mixture of X_1 and X_2 with mixing weights a and 1-a, we can define the parameter $\Theta = \{1,2\}$, with $P[\Theta = 1] = a$, $P[\Theta = 2] = 1 - a$.

Then
$$E[Y] = E[E[Y|\Theta]] = E[Y|\Theta = 1] \times P[\Theta = 1] + E[Y|\Theta = 2] \times P[\Theta = 2]$$

= $E[X_1] \times a + E[X_2] \times (1 - a)$

(this is the usual way the mean of a finite mixture is formulated).

 $Var[Y] = E[Var[Y|\Theta]] + Var[E[Y|\Theta]], \text{ where}$ $E[Var[Y|\Theta]] = Var[X_1] \times a + Var[X_2] \times (1-a) \text{ and}$ $Var[E[Y|\Theta]] = (E[X_1] - E[X_2])^2 \times a \times (1-a).$

The last equality follows from the fact that if Z is a two-point random variable

$$Z = \begin{cases} u & \text{prob. } p \\ v & \text{prob. } 1 - p \end{cases}, \text{ then } Var[Z] = (u - v)^2 \times p \times (1 - p); E[Y|\Theta] \text{ is }$$

a two-point random variable $E[Y|\Theta] = \begin{cases} E[X_1] & \text{prob. } a \\ E[X_2] & \text{prob. } 1-a \end{cases}$.

Therefore $Var[Y] = Var[X_1] \times a + Var[X_2] \times (1-a) + (E[X_1] - E[X_2])^2 \times a \times (1-a).$

B) If X_1, X_2 are Poisson random variables then $E[X_1] = Var[X_1]$ and $E[X_2] = Var[X_2]$, so that $E[Y] = E[X_1] \times a + E[X_2] \times (1-a) = Var[X_1] \times a + Var[X_2] \times (1-a)$.

It follows that $Var[Y] = Var[X_1] \times a + Var[X_2] \times (1-a) + (E[X_1] - E[X_2])^2 \times a \times (1-a) > E[Y].$

C) For a negative binomial random variable with parameters r and β , the mean is $r\beta$ and the variance is $r\beta(1 + \beta)$, so the variance is larger than the mean. If X_1 and X_2 have negative binomial distributions, the $E[X_1] < Var[X_1]$ and $E[X_2] < Var[X_2]$.

Therefore, $E[Y] = E[X_1] \times a + E[X_2] \times (1-a) < Var[X_1] \times a + Var[X_2] \times (1-a)$, and $Var[X_1] \times a + Var[X_2] \times (1-a)$ $< Var[X_1] \times a + Var[X_2] \times (1-a) + (E[X_1] - E[X_2])^2 \times a(1-a) = Var[Y]$; therefore E[Y] < Var[Y].

A) For a binomial random variable with parameters m and q, the mean is mq and the variance is mq(1-q), which is smaller than the mean.

Therefore $E[Y] = E[X_1] \times a + E[X_2] \times (1-a) > Var[X_1] \times a + Var[X_2] \times (1-a)$, and it is possible that when we add $(E[X_1] - E[X_2])^2 \times a \times (1-a)$ to the right side, we get approximate equality.

So it is possible that E[Y] = Var[Y] for a mixture of binomials.

Answer A

11. The mean will be $\frac{1}{2} \times (m_1 + m_2)$ and the variance will be $\frac{1}{2} \times (2m_1^2 + 2m_2^2) - [\frac{1}{2} \times (m_1 + m_2)]^2$. The coefficient of variation is the ratio of standard deviation to mean.

The square of the coefficient of variation is

$$\frac{\frac{1}{2} \times (2m_1^2 + 2m_2^2) - [\frac{1}{2} \times (m_1 + m_2)]^2}{[\frac{1}{2} \times (m_1 + m_2)]^2} = \frac{3m_1^2 - 2m_1m_2 + 3m_2^2}{(m_1 + m_2)^2}$$
$$= \frac{3m_1^2 + 6m_1m_2 + 3m_2^2 - 8m_1m_2}{(m_1 + m_2)^2}$$
$$= 3 - \frac{8m_1m_2}{(m_1 + m_2)^2}.$$

The maximum square of the coefficient of variation is 3, and it occurs at the minimum of $\frac{8m_1m_2}{(m_1+m_2)^2}$, which is 0 (if m_1 or m_2 is 0).

The least upper bound of the coefficient of variation is $\sqrt{3}$.

Answer C

12. Var[Z] = Var[E[Z | Y]] + E[Var[Z | Y]].

 $E[Z | Y] = 2 \times Y$ (a sum of Y independent normal random variables each with mean 2), and $Var[Z | Y] = 2 \times Y$ (a sum of Y independent normal random variables each with a variance of 2).

Thus, $Var[Z] = Var[2 \times Y] + E[2 \times Y] = 4 \times Var[Y] + 2 \times E[Y].$ Var[Y] = Var[E[Y | X]] + E[Var[Y | X]].Since E[Y] = 0.6 with prob. 0.4 (X = 0) and E[Y] = 1.3 with prob. 0.6 (X = 1), it follows that $Var[E[Y | X]] = (0.6)^2 \times 0.4 + (1.3)^2 \times 0.6 - (1.02)^2 = 0.1176.$

Also, if X = 0, the variance of Y is $1^2 \times 0.2 + 2^2 \times 0.2 - (0.6)^2 = 0.64$, and if X = 1, the variance of Y is $1^2 \times 0.3 + 2^2 \times 0.5 - (1.3)^2 = 0.61$. Thus, $E[Var[Y \mid X]] = 0.64 \times 0.4 + .61 \times .6 = 0.622$, and so Var[Y] = 0.1176 + 0.622 = 0.7396.

Then, the variance of Z is $4 \times 0.7396 + 2 \times 1.02 = 4.9984$. Note that Var[Y] can also be found from $E[Y^2] - (E[Y])^2$. Again, E[Y] = 1.02, as above, and now, $E[Y^2] = E[E[Y^2 \mid X]] = 1 \times 0.4 + 2.3 \times 0.6 = 1.78$, so that Var[Y] = 0.7396.

Answer A

13. Given that a local train arrives first, you will get to work 28 minutes after that local train arrives, since you will take it. Your co-worker will wait for first express train. You will get to work before your co-worker if the next express train (after the local) arrives more than 12 minutes after the local. We expect 5 express trains per hour, so the time between express trains is exponentially distributed with a mean of $\frac{1}{5}$ of an hour, or 12 minutes. Because of the lack of memory property of the exponential distribution, since we are given that the next train is local, the time until the next express train after that is exponential with a mean of 12 minutes. Therefore, the probability that after the local, the next express arrives in more than 12 minutes is P[T > 12], where T has an exponential distribution with a mean of 12. This probability is $e^{-12/12} = e^{-1} = 0.368$ (37%).

Answer A



Ready for more practice? Check out GOAL!

GOAL offers additional questions, quizzes, and simulated exams with helpful solutions and tips. Included with GOAL are topic-by-topic instructional videos! Head to ActuarialUniversity.com and log into your account.

Practice Exam 1

1. The XYZ Insurance Company sells property insurance policies with a deductible of \$5,000, policy limit of \$500,000, and a coinsurance factor of 80%. Let X_i be the individual loss amount of the *i*th claim and Y_i be the claim payment of the *i*th claim. Which of the following represents the relationship between X_i and Y_i ?

(A) $Y_i = \begin{cases} 0\\ 0.80 (X_i - 5,000)\\ 500,000 \end{cases}$	$X_i \le 5,000$ $5,000 < X_i \le 625,000$ $X_i > 625,000$
(B) $Y_i = \begin{cases} 0\\ 0.80 (X_i - 4,000)\\ 500,000 \end{cases}$	$X_i \le 5,000$ $4,000 < X_i \le 500,000$ $X_i > 500,000$
(C) $Y_i = \begin{cases} 0\\ 0.80 (X_i - 5,000)\\ 500,000 \end{cases}$	$X_i \le 5,000$ $5,000 < X_i \le 630,000$ $X_i > 630,000$
(D) $Y_i = \begin{cases} 0\\ 0.80 (X_i - 6,250)\\ 500,000 \end{cases}$	$X_i \le 6,250$ $6,250 < X_i \le 631,500$ $X_i > 631,500$
(E) $Y_i = \begin{cases} 0\\ 0.80 (X_i - 5,000)\\ 500,000 \end{cases}$	$X_i \le 5,000$ $5,000 < X_i \le 505,000$ $X_i > 505,000$

- 2. A casino has a game that makes payouts at a Poisson rate of 5 per hour and the payout amounts can be any non-negative integer, 1, 2, 3, ... without limit. The probability that any given payout is equal to i > 0 is $\frac{1}{2^i}$. Payouts are independent. Calculate the probability that there are no payouts of 1,2, or 3 in a given 20 minute period.
 - (A) 0.08 (B) 0.13 (C) 0.18 (D) 0.23 (E) 0.28

3. A compound Poisson claim distribution has $\lambda = 3$ and individual claims amounts distributed as follows:

$$egin{array}{ccc} x & f_X(x) \ 5 & 0.6 \ 10 & 0.4 \end{array}$$

Determine the expected cost of an aggregate stop-loss insurance with a deductible of 6.

- (A) Less than 15.0
- (B) At least 15.0 but less than 15.3
- (C) At least 15.3 but less than 15.6
- (D) At least 15.6 but less than 15.9
- (E) At least 15.9

Use the following information for questions 4 and 5. You are the producer of a television quiz show that gives cash prizes. The number of prizes, N, and prize amounts, X are independent of one another and have the following distributions:

N:
$$P[N = 1] = 0.8$$
, $P[N = 2] = 0.2$

- X: P[X = 0] = 0.2, P[X = 100] = 0.7, P[X = 1000] = 0.1
- 4. Your budget for prizes equals the expected prizes plus $1.25 \times$ standard deviation of prizes. Calculate your budget.
 - (A) 384 (B) 394 (C) 494 (D) 588 (E) 596 (E) 596
- 5. You buy stop-loss insurance for prizes with a deductible of 200. The cost of insurance includes a 175% relative security load (the relative security load is the percentage of expected payment that is added). Calculate the cost of the insurance.
 - (A) 204 (B) 227 (C) 245 (D) 273 (E) 357

6. An actuary determines that claim counts follow a negative binomial distribution with unknown β and r. It is also determined that individual claim amounts are independent and identically distributed with mean 700 and variance 1,300. Aggregate losses have a mean of 48,000 and variance 80 million. Calculate the values for β and r.

 $\begin{array}{ll} ({\rm A}) \ \beta = 1.20, r = 57.19 \\ ({\rm C}) \ \beta = 2.38, r = 28.83 \\ ({\rm E}) \ \beta = 1,664.81, r = 0.04 \end{array} \\ \end{array} \\ \begin{array}{ll} ({\rm B}) \ \beta = 1.38, r = 49.75 \\ ({\rm D}) \ \beta = 1,663.81, r = 0.04 \end{array}$

7. \checkmark Let X_1, X_2, X_3 be independent Poisson random variables with means $\theta, 2\theta$, and 3θ respectively. What is the maximum likelihood estimator of θ based on sample values x_1, x_2 , and x_3 from the distributions of X_1 , X_2 and X_3 , respectively,

(A)
$$\frac{x}{2}$$

(B) \bar{x}
(C) $\frac{x_1 + 2x_2 + 3x_3}{6}$
(D) $\frac{3x_1 + 2x_2 + x_3}{6}$

(E)
$$\frac{6x_1 + 3x_2 + 2x_3}{11}$$

8. **V** For a group of policies, you are given:

 x_3

- (i) Losses follow a uniform distribution on the interval $(0, \theta)$, where $\theta > 25$.
- (ii) A sample of 20 losses resulted in the following:

Interval	Number of Losses
$x \le 10$	n_1
$10 < x \le 25$	n_2
x > 25	n_3

The maximum likelihood estimate of θ can be written in the form 25 + y. Determine y.

(A)
$$\frac{25n_1}{n_2 + n_3}$$

(B) $\frac{25n_2}{n_1 + n_3}$
(C) $\frac{25n_3}{n_1 + n_2}$
(D) $\frac{25n_1}{n_1 + n_2 + n_3}$
(E) $\frac{25n_2}{n_1 + n_2 + n_3}$

9. So The number of claims follows a negative binomial distribution with parameters β and r, where β is unknown and r is known. You wish to estimate β based on n observations, where \bar{x} is the mean of these observations. Determine the maximum likelihood estimate of β .

(A)
$$\frac{\bar{x}}{r^2}$$
 (B) $\frac{\bar{x}}{r}$ (C) \bar{x} (D) $r\bar{x}$ (E) $r^2\bar{x}$

10. The following 6 observations are assumed to come from the continuous distribution with pdf $f(x;\theta) = \frac{1}{2}x^2\theta^3 e^{-\theta x} : 1, 3, 4, 4, 5, 7.$

Find the mle of θ .

- (A) 0.25 (B) 0.50 (C) 0.75 (D) 1.00 (E) 1.25
- 11. An analysis of credibility premiums is being done for a particular compound Poisson claims distribution, where the criterion is that the total cost of claims is within 5% of the expected cost of claims with a probability of 90%. It is found that with n = 60 exposures (periods) and $\bar{X} = 180.0$, the credibility premium is 189.47. After 20 more exposures (for a total of 80) and revised $\bar{X} = 185$, the credibility premium is 190.88. After 20 more exposures (for a total of 100) the revised \bar{X} is 187.5. Assuming that the manual premium remains unchanged in all cases, and assuming that full credibility has not been reached in any of the cases, find the credibility premium for the 100 exposure case.
 - $(A) 191.5 (B) 192.5 (C) 193.5 (D) 194.5 (E) 196.5 \\ (E) 196.5 (E$

12. \checkmark For an insurance portfolio, you are given:

- (i) For each individual insured, the number of claims follows a Poisson distribution.
- (ii) The mean claim count varies by insured, and the distribution of mean claim counts follows gamma distribution.
- (iii) For a random sample of 1000 insureds, the observed claim counts are as follows:

Number of Claims, n	0	1	2	3	4	5
Number of Insureds, f_n	512	307	123	41	11	6

$$\sum nf_n = 750, \qquad \sum n^2 f_n = 1494$$

- (iv) Claim sizes follow a Pareto distribution with mean 1500 and variance 6,750,000.
- (v) Claim sizes and claim counts are independent.
- (vi) The full credibility standard is to be within 5% of the expected aggregate loss 95% of the time.

Determine the minimum number of insureds needed for the aggregate loss to be fully credible.

- (A) Less than 8300
- (B) At least 8300, but less than 8400
- (C) At least 8400, but less than 8500
- (D) At least 8500, but less than 8600
- (E) At least 8600

	Cumulative Incurred Losses			Paid-to-Date	
Accident	Development Year		at Dec 31, AY4		
Year	0	1	2	3	
AY1	2325	3749	4577	4701	4701
AY2	2657	4438	5529		4500
AY3	2913	4995			3500
AY4	3163				2500

Information on Questions 13 and 14 is as follows. You are given the following information on cumulative incurred losses through development years shown.

- 13. Vising an average factor model, calculate the estimated total loss reserve as of Dec. 31, AY4.
 - (A) Less than 7500
 - (B) At least 7500 but less than 7800 $\,$
 - (C) At least 7800 but less than 8100
 - (D) At least 8100 but less than 8400
 - (E) At least 8400
- 14. S of Dec. 31, AY4, calculate

Estimated reserve for AY2 based on an average factor model) – (Estimated reserve for AY2 based on a mean factor model)

- (A) Less than -300
- (B) At least -300 but less than -100
- (C) At least -100 but less than 100
- (D) At least 100 but less than 300
- (E) At least 300

- 15. For a one-period binomial model for the price of a stock with price 100 at time 0, you are given:
 - (i) The stock pays no dividends.
 - (ii) The stock price is either 110 or 95 at the end of the year.
 - (iii) The risk free force of interest is 5%.

Calculate the price at time 0 of a one-year call option with strike price 100.

- (A) Less than 6.00
- (B) At least 6.00 but less than 6.25
- (C) At least 6.25 but less than 6.50
- (D) At least 6.50 but less than 6.75
- (E) At least 6.75
- 16. Vising the following information, determine the incurred losses for the 2017 accident year as reported at Dec. 31, 2018.

Occurrence #1: Occurrence date Feb. 1/16, Report date Apr. 1/16 Loss History:

Date	Total Paid to Date	Unpaid Loss Reserve	Total Incurred
Apr. 1/16	1000	1000	2000
Dec. 31/16	1500	1000	2500
Dec. 31/17	1500	1000	2500
Mar. $1/18$	3000	0	3000

Occurrence #2: Occurrence date May 1/17, Report date July 1/17 Loss History:

Date	Total Paid to Date	Unpaid Loss Reserve	Total Incurred
July 1/17	1000	2000	3000
Dec. $31/17$	3000	1000	4000
Dec. 31/18	5000	0	5000

Occurrence #3: Occurrence date Nov. 1/17, Report date Feb. 1/18 Loss History

Date	Total Paid to Date	Unpaid Loss Reserve	Total Incurred	
Mar. 1/18	0	8000	8000	
Dec. 31/18	5000	5000	10,000	
(A) 0	(B) 3,000	(C) 5,000	(D) 8,000	(E) 15,000

17. 💙 You are given the following calendar year earned premium.

Year	CY2	CY3	CY4
Earned Premium	4200	4700	5000

You are also given the following rate changes

Date	April 1, CY1	September 1, CY2	July 1, CY3
Average Rate Change	+12 %	+6 %	+10 %

Determine the approximate earned premium at current (end of CY4) rates for CY3.

- (A) Less than 5000
- (B) At least 5000 but less than 5100
- (C) At least 5100 but less than 5200
- (D) At least 5200 but less than 5300 $\,$
- (E) At least 5300

** END OF EXAMINATION **

Solutions to Practice Exam 1

1.	С	
2.	D	
3.	С	
4.	Е	
5.	D	
6.	В	
7.	А	
8.	С	
9.	В	
10.	С	
11.	А	
12.	Е	
13.	D	
14.	С	
15.	С	
16.	Е	
17.	С	

- 1. With coinsurance factor α , deductible d, policy limit $\alpha(u-d)$, the amount paid per loss is
 - With combutance factor x_{r} (we are assuming in inflation rate of r = 0) $Y = \begin{cases} 0 & X \le d \\ \alpha(X-d) & d < X \le u. \\ \alpha(u-d) & X > u \end{cases}$

In this problem, the coinsurance factor is $\alpha = .8$, the deductible is d = 5,000, and the policy limit is .8(u-5,000) = 500,000, so that the maximum covered loss is u = 630,000.

The amount paid per loss becomes
$$Y = \begin{cases} 0 & X \le 5,000 \\ 0.80(X - 5,000) & 5,000 < X \le 630,000 \\ 500,000 & X > 630,000 \end{cases}$$

Answer C

2. When a payout occurs, it is 1, 2 or 3 with probability $\frac{1}{2} + \frac{1}{2^2} + \frac{1}{2^3} = \frac{7}{8}$. The number of payouts that are 1,2 or 3 follows a Poisson process with an hourly rate of $5 \times \frac{7}{8} = \frac{35}{8}$. The expected number of payouts that are 1, 2 or 3 in 20 minutes, say N, has a Poisson distribution with mean $\frac{35}{8} \times \frac{20}{60} = \frac{35}{24}$. The probability that there are no payouts of 1, 2, or 3 in a given 20 minute period is the probability that N = 0, which is $e^{-35/24} = .233$.

Answer D

3. The minimum claim amount is 5 if a claim occurs. S must be 0 or a multiple of 5. The stop-loss insurance with deductible 6 pays $(S-6)_+ = S - (S \wedge 6)$,

where
$$S \wedge 6 = \begin{cases} 0 & S = 0 \\ 5 & S = 5 \\ 6 & S \ge 10 \end{cases}$$

 $E[S] = E[N] \times E[X] = (3)[(5)(.6) + (10)(.4)] = 21$ $E[S \land 6] = 5 \times P(S = 5) + 6[1 - P(S = 0,5)].$ $P(S = 0) = P(N = 0) = e^{-3}$ and $P(S=5) = P(N=1) \times P(X=5) = e^{-3} \times 3 \times (.6) = 1.8e^{-3}.$ $E[S \land 6] = 5(1.8e^{-3}) + 6[1 - P(S = 0.5)] = 6[1 - 2.8e^{-3}] = 5.61.$ Then $E[(S-6)_+] = E[S] - E[S \land 6] = 21 - 5.61 = 15.39.$

Answer C

4. This problem involves a compound distribution. The frequency (number of prizes) is N and the severity (prize amount) is X. The aggregate prize amount is $S = X_1 + X_2 + x + X_N$, with mean $E[S] = E[N] \times E[X] = (1.2)(170) = 204$ and variance $Var[S] = E[N] \times Var[X] + Var[N] \times (E[X])^2$ In this case, $Var[N] = E[N^2] - (E[N])^2 = 1.6 - (1.2)^2 = .16$, and $Var[X] = E[X^2] - (E[X])^2 = 107,000 - (170)^2 = 78,100$ Then, $Var[S] = (1.2)(78,100) + (.16)(170)^2 = 98,344.$ The budget is $E[S] + 1.25\sqrt{Var[S]} = 204 + 1.25\sqrt{98,344} = 596.$

Answer E