

SAS A00-225

SAS ADVANCED ANALYTICS PROFESSIONAL CERTIFICATION
QUESTIONS & ANSWERS

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A00-225

[SAS Certified Advanced Analytics Professional Using SAS 9](#)

50-55 Questions Exam – 67% Cut Score – Duration of 110 minutes



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Discover More about the A00-225 Certification

Are you interested in passing the SAS A00-225 exam? First discover, who benefits from the A00-225 certification. The A00-225 is suitable for a candidate if he wants to learn about Advanced Analytics Professional. Passing the A00-225 exam earns you the SAS Certified Advanced Analytics Professional Using SAS 9 title.

While preparing for the A00-225 exam, many candidates struggle to get the necessary materials. But do not worry; your struggling days are over. The A00-225 PDF contains some of the most valuable preparation tips and the details and instant access to useful [A00-225 study materials just at one click](#).

A00-225 SAS Advanced Analytics Professional Certification Details:

Exam Name	SAS Advanced Predictive Modeling
Exam Code	A00-225
Exam Duration	110 minutes
Exam Questions	50 to 55 Multiple choices or short answer questions
Passing Score	67%
Exam Price	\$180 (USD)
Training	<ol style="list-style-type: none"> 1. SAS Academy for Data Science: Advanced Analytics 2. Neural Network Modeling 3. Predictive Modeling Using Logistic Regression 4. SAS Visual Statistics: Interactive Model Building 5. Predictive Modeling Using SAS In-Memory Statistics 6. SAS Enterprise Miner High-Performance Data Mining Nodes 7. Using SAS to Put Open Source Models into Production
Exam Registration	Pearson VUE
Sample Questions	SAS Advanced Analytics Professional Certification Sample Question
Practice Exam	SAS Advanced Analytics Professional Certification Practice Exam

A00-225 Syllabus:

Objective	Details
Neural Networks - 20%	
Describe key concepts underlying neural networks	<ul style="list-style-type: none"> - Use SAS procedures to perform nonlinear modeling <ul style="list-style-type: none"> • Use the NLIN procedure for non-linear regression - Explain advantages and disadvantages of using neural networks compared to other approaches <ul style="list-style-type: none"> • Explain two ways to respond to the black-box objection • Compare and contrast variable selection, degrees of freedom to traditional approaches • Explain advantages of the Widrow-Hoff Delta rule
Use two architectures offered by the Neural Network node to model either linear or non-linear input-output relationships	<ul style="list-style-type: none"> - Define the linear perceptron neural network <ul style="list-style-type: none"> • Define combination functions (linear, additive, equal slopes) • Define activation functions (logistic, tanh, arctan, softmax, exponential, identity) • Explain the difference between activation and link functions - Be able to demonstrate how a linear perceptron is a generalized linear model that is able to model many target distributions <ul style="list-style-type: none"> • Explain the difference between a general and generalized model • Demonstrate the power of the NEURAL procedure in SAS - Construct multilayer perceptrons <ul style="list-style-type: none"> • Define the three layers in a basic multilayer perceptron (input, hidden, output) • Explain how you can obtain a skip-layer network - Construct radial basis function networks <ul style="list-style-type: none"> • Compare ordinary and normalized - Identify advantages of using a radial basis function network over using a multilayer perceptron (invert order)
Use optimization methods offered by the SAS Enterprise Miner Neural	<ul style="list-style-type: none"> - Describe the problem of local minima - Explain the rationale behind the initialization settings - Explain how early stopping and weight decay can be used to help avoid bad local minima

Objective	Details
<p>Network node to efficiently search the parameter space in a neural network</p>	<ul style="list-style-type: none"> - Describe parameter estimation methods and determine best method to use - List the assortment of error functions that are available in the Neural Networks node and determine the appropriate one to use based upon statistical considerations <ul style="list-style-type: none"> • Find the parameter set that minimizes the specified error function • Ordinary least squares • Maximum likelihood /Minimizing Deviance • Robust estimation methods Huber's M-estimation (HUBER) • Determine the appropriate activation and error function combination to apply based on the target data - List the optimization (training) techniques available in the Neural Networks node and determine the appropriate method to use based upon statistical considerations <ul style="list-style-type: none"> • iterative updating • back propagation <ul style="list-style-type: none"> - Conjugate gradient - Quasi-Newton - Levenberg-Marquardt
<p>Construct custom network architectures by using the NEURAL procedure (PROC Neural)</p>	<ul style="list-style-type: none"> - Working with SAS Enterprise Miner, use selected NEURAL procedure statements and PROC DMDB to construct neural networks <ul style="list-style-type: none"> • ARCHI • CONNECT • HIDDEN • INPUT • PRELIM • TARGET • TRAIN - Define Sequential Network Construction (SNC) and use it to build an MLP(Multilayer Perceptron) - Use weight interpretation to select relevant input variables - Define a generalized additive neural network (GANN) and be able to explain the use of the GANN paradigm
<p>Based upon statistical considerations, use either time delayed neural networks, surrogate models to</p>	<ul style="list-style-type: none"> - Given a particular scenario/problem, use the time delayed neural network (TDNN) model to conduct time series analysis - Apply a surrogate model to help understand a neural network's predictions

Objective	Details
augment neural networks	<ul style="list-style-type: none"> • Interpret a neural network with a continuous target • Interpret a neural network with a discrete target
Use the HP Neural Node to perform high-speed training of a neural network	
Logistic Regression - 30%	
Score new data sets using the LOGISTIC and PLM procedures	<ul style="list-style-type: none"> - Use the SCORE statement in the PLM procedure to score new cases - Use the CODE statement in PROC LOGISITIC to score new data - Describe when you would use the SCORE statement vs the CODE statement in PROC LOGISTIC - Use the INMODEL/OUTMODEL options in PROC LOGISTIC - Explain how to score new data when you have developed a model from a biased sample
Identify the potential challenges when preparing input data for a model	<ul style="list-style-type: none"> - Identify problems that missing values can cause in creating predictive models and scoring new data sets - Identify limitations of Complete Case Analysis - Explain problems caused by categorical variables with numerous levels - Discuss the problem of redundant variables - Discuss the problem of irrelevant and redundant variables - Discuss the non-linearities and the problems they create in predictive models - Discuss outliers and the problems they create in predictive models - Describe quasi-complete separation - Discuss the effect of interactions - Determine when it is necessary to oversample data
Use the DATA step to manipulate data with loops, arrays, conditional statements and functions	<ul style="list-style-type: none"> - Use ARRAYS to create missing indicators - Use ARRAYS, LOOP, IF, and explicit OUTPUT statements
Improve the predictive power of categorical inputs	<ul style="list-style-type: none"> - Reduce the number of levels of a categorical variable - Explain thresholding - Explain Greenacre's method - Cluster the levels of a categorical variable via Greenacre's method using the CLUSTER procedure <ul style="list-style-type: none"> • METHOD=WARD option • FREQ, VAR, ID statement

Objective	Details
	<ul style="list-style-type: none"> • Use of ODS output to create an output data set - Convert categorical variables to continuous using smooth weight of evidence
Screen variables for irrelevance and non-linear association using the CORR procedure	<ul style="list-style-type: none"> - Explain how Hoeffding's D and Spearman statistics can be used to find irrelevant variables and non-linear associations - Produce Spearman and Hoeffding's D statistic using the CORR procedure (VAR, WITH statement) - Interpret a scatter plot of Hoeffding's D and Spearman statistic to identify irrelevant variables and non-linear associations
Screen variables for non-linearity using empirical logit plots	<ul style="list-style-type: none"> - Use the RANK procedure to bin continuous input variables (GROUPS=, OUT= option; VAR, RANK statements) - Interpret RANK procedure output - Use the MEANS procedure to calculate the sum and means for the target cases and total events (NWAY option; CLASS, VAR, OUTPUT statements) - Create empirical logit plots with the GPLOT procedure - Interpret empirical logit plots
Apply the principles of honest assessment to model performance measurement	<ul style="list-style-type: none"> - Explain techniques to honestly assess classifier performance - Explain overfitting - Explain differences between validation and test data - Identify the impact of performing data preparation before data is split
Assess classifier performance using the confusion matrix	<ul style="list-style-type: none"> - Explain the confusion matrix - Define: Accuracy, Error Rate, Sensitivity, Specificity, PV+, PV- - Explain the effect of oversampling on the confusion matrix - Adjust the confusion matrix for oversampling
Model selection and validation using training and validation data	<ul style="list-style-type: none"> - Divide data into training and validation data sets using the SURVEYSELECT procedure - Discuss the subset selection methods available in PROC LOGISTIC - Discuss methods to determine interactions (forward selection, with bar and @ notation) - Create interaction plot with the results from PROC LOGISTIC - Select the model with fit statistics (BIC, AIC, KS, Brier score)
Create and interpret graphs (ROC, lift, and gains charts) for model comparison and selection	<ul style="list-style-type: none"> - Explain and interpret charts (ROC, Lift, Gains) - Create a ROC curve (OUTROC option of the SCORE statement in the LOGISTIC procedure) - Use the ROC and ROCONTRAST statements to create an overlay plot of ROC curves for two or more models

Objective	Details
	<ul style="list-style-type: none"> - Explain the concept of depth as it relates to the gains chart
<p>Establish effective decision cut-off values for scoring</p>	<ul style="list-style-type: none"> - Illustrate a decision rule that maximizes the expected profit - Explain the profit matrix and how to use it to estimate the profit per scored customer - Calculate decision cutoffs using Bayes rule, given a profit matrix - Determine optimum cutoff values from profit plots - Given a profit matrix, and model results, determine the model with the highest average profit
<p>Predictive Analytics on Big Data - 40%</p>	
<p>Build and interpret a cluster analysis in SAS Visual Statistics</p>	<ul style="list-style-type: none"> - Assign roles for cluster analysis - Set cluster matrix properties (number, seed, etc) - Select the proper inputs for the k-means algorithm for a given cluster analysis scenario - Choose the number of clusters for a given cluster analysis scenario - Set Parallel coordinate properties for cluster analysis - Interpret a cluster matrix - Interpret a parallel coordinates plot - Display summary statistics for clusters - Interpret summary statistics for clusters - Assign cluster IDs to the data within Visual Statistics - Score observations into clusters based on the results from Visual Statistics
<p>Explain SAS high-performance computing</p>	<ul style="list-style-type: none"> - Identify limitations of traditional computing environments - Describe the characteristics of SAS High-Performance Analytics procedures - Compare SMP and MPP computing modes - Distinguish between HPA and the LASR related operation
<p>Perform principal component analysis</p>	<ul style="list-style-type: none"> - Explain how principal component analysis is performed - List the benefits and problems of principal component analysis - Distinguish between clustering, variable clustering, and principal component analysis - Determine the number of principal components to retain - Compare IMSTAT, Visual Statistics, and High Performance Computing nodes in Enterprise Miner
<p>Analyze categorical targets using logistic regression in SAS Visual Statistics</p>	<ul style="list-style-type: none"> - Assign roles for logistic regression - Assign properties for logistic regression - Filter data used for logistic regression - Interpret logistic regression results (fit summary, residual plots, ROC/Lift charts, etc)

Objective	Details
	<ul style="list-style-type: none"> - Use Group-By variables to perform binary logistic regression
<p>Analyze categorical targets using decision trees in SAS Visual Statistics</p>	<ul style="list-style-type: none"> - Assign roles for decision trees - Assign properties for decision trees - Interpret decision trees results (trees, leaf statistics, assessment, etc) - Identify variable importance with decision trees for use in other analysis techniques - Splitting criteria used by Visual Statistics
<p>Analyze categorical targets using decision trees in PROC IMSTAT</p>	<ul style="list-style-type: none"> - Use the DECISIONTREE statement to create decision trees - Define input variables with the INPUT and NOMINAL options - Create and retrieve saved trees for input data scoring with the SAVE, TREETAB, and ASSESS options - Evaluate the output of ODS tables (DTREE, DTreeVarImpInfo, DTREESCORE, etc) from decision trees - Use the ASSESS statement to create data sets for evaluating the decision tree model - Perform honest assessment on PROC IMSTAT decision trees - Assess decision trees using ODS statistical graphics (SGPLOT)
<p>Analyze categorical targets using logistic regression in PROC IMSTAT</p>	<ul style="list-style-type: none"> - Assign variables to roles for logistic regression in PROC IMSTAT - Create logistic regression in PROC IMSTAT using the LOGISTIC statement - Use selected options of the LOGISTIC STATEMENT (ROLEVAR, INPUTS, SCORE, CODE, SHOWSELECTED, SLSTAY=) - Assess logistic regression models using ODS statistical graphics (SGPLOT) - Perform honest assessment on PROC IMSTAT logistic regression
<p>Build random forest models with PROC IMSTAT</p>	<ul style="list-style-type: none"> - Describe random forests - Use the RANDOMWOODS statement to build a forest of trees - Score data with the RANDOMWOODS score code - List benefits of forests - Interpret random forests - Identify variable importance with forest for use in other analysis techniques
<p>Analyze interval targets with SAS Visual Statistics</p>	<ul style="list-style-type: none"> - Build linear regression models in SAS Visual Statistics - Assign roles for linear regression models - Set properties for linear regression models - Assess a linear regression model (evaluate Fit summary statistics, residual plot, influence plot, summary table, etc)

Objective	Details
	<ul style="list-style-type: none"> - Assess linear model assumption violations and recognize when linear model is inadequate - Build generalized linear models in SAS Visual Statistics - Assign roles for generalized linear models - Set properties for generalized linear models - Assess a generalized linear model (evaluate Fit summary statistics, residual plot, assessment, etc)
<p>Analyze interval targets with PROC IMSTAT</p>	<ul style="list-style-type: none"> - Use GENMODEL and GLM statements - Distinguish between GENMODEL and GLM statements and the results of each procedure - Assign variables to roles for GENMODEL and GLM statements in PROC IMSTAT - Create models with GENMODEL and GLM statements in PROC IMSTAT - Use selected options of the GENMODEL and GLM statements in PROC IMSTAT - Assess models using ODS statistical graphics (SGPLOT) - Perform honest assessment on PROC IMSTAT linear models
<p>Analyze zero inflated models with HPGLM in Enterprise Miner</p>	<ul style="list-style-type: none"> - Identify when it would be appropriate to use mixture distribution - Describe the link functions and distributions available in the HP GLM node - Build a zero inflated generalized linear model in EM - Describe restrictions on roles and levels in input data sources for generalized linear models in EM - Assess a zero inflated generalized linear model (evaluate Fit summary statistics, residual plot, assessment, etc)
<p>Open Source Models in SAS - 10%</p>	
<p>Incorporate an existing R program into SAS Enterprise Miner</p>	<ul style="list-style-type: none"> - Enable R language statements to connect SAS to R - Use the Open Source Integration node in SAS Enterprise Miner <ul style="list-style-type: none"> • Modes of operation (training, output) • Use Predictive Modeling Markup Language (PMML) in Open Source Integration Node - Use Enterprise Miner variable handles to alter an R script <ul style="list-style-type: none"> • Use Enterprise Miner to run a random forest in R
<p>Incorporate an existing Python program into SAS Enterprise Miner</p>	<ul style="list-style-type: none"> - Determine steps to perform in SAS to incorporate a Python model - Determine nodes in Enterprise Miner to incorporate a Python model - Determine the necessary set up requirements for running Python models in SAS

Broaden Your Knowledge with SAS A00-225

Sample Questions:

Question: 1

Why is a decision tree an ideal surrogate model for a neural network?

- a) The if-then rules are easy to interpret.
- b) A decision tree is a black-box.
- c) A decision tree can be used to do variable selection.
- d) A decision tree is a parametric model.

Answer: a

Question: 2

In the Open Source Integration node in SAS Enterprise Miner, which Output Mode(s) creates SAS DATA step score code for the user?

- a) Predictive Modeling Markup Language (PMML)
- b) None
- c) Merge
- d) Both PMML and Merge

Answer: a

Question: 3

After a logistic regression has been created in SAS Visual Statistics, you discover that not all observations were used to create the model. How would you run the model on all of the data?

- a) Include the correct interaction term.
- b) Select Informative Missingness on the properties tab.
- c) Include the correct offset term.
- d) Select Use Variable Selection on the properties tab.

Answer: b

Question: 4

In a forest, what is the out of bag (OOB) sample?

- a) A random partition of the validation data
- b) The partition of training data not used in growing an individual tree
- c) The partition of training data not used in growing any tree
- d) The partition of the validation data most closely resembling the data used to train a tree

Answer: b

Question: 5

McCulloch-Pits neurons used which activation functions?

- a) hyperbolic tangent
- b) logistic function
- c) Elliot function
- d) step function

Answer: d

Question: 6

In order to take advantage of a neural network's ability to model a nonlinear relationship between inputs and outputs, which feature of the network is necessary?

- a) The inclusion of direct connections between the input and output units in the network.
- b) At least one hidden layer with a non-linear activation function.
- c) A non-linear combination function in the output units.
- d) A sigmoidal activation function in the output units.

Answer: b

Question: 7

Which statement is true with respect to the DECISIONTREE statement in PROC IMSTAT?

- a) Only binary target variables are supported.
- b) By default, pruning is based on assessment of a holdout sample.
- c) The C4.5 decision tree methodology is employed to derive the decision tree.
- d) Pruning can be controlled using the GREEDY option.

Answer: c

Question: 8

The softmax activation function is appropriate for which type of target?

- a) Continuous
- b) Binary
- c) Ordinal
- d) Multinomial

Answer: d

Question: 9

What is the primary purpose of weight decay?

- a) Prevent overfitting.
- b) Prevent underfitting.
- c) Avoid bad local minima.
- d) Increase convergence speed.

Answer: a

Question: 10

Which SAS Enterprise Miner node should you use to run a Python script?

- a) Open Source Integration node
- b) Model Import node
- c) SAS Code node
- d) Register Model node

Answer: c

Avail the Study Guide to Pass A00-225 SAS Advanced Analytics Professional Exam:

- Find out about the A00-225 syllabus topics. Visiting the official site offers an idea about the exam structure and other important study resources. Going through the syllabus topics help to plan the exam in an organized manner.
- Once you are done exploring the [A00-225 syllabus](#), it is time to plan for studying and covering the syllabus topics from the core. Chalk out the best plan for yourself to cover each part of the syllabus in a hassle-free manner.
- A study schedule helps you to stay calm throughout your exam preparation. It should contain your materials and thoughts like study hours, number of topics for daily studying mentioned on it. The best bet to clear the exam is to follow your schedule rigorously.
- The candidate should not miss out on the scope to learn from the A00-225 training. Joining the SAS provided training for A00-225 exam helps a candidate to strengthen his practical knowledge base from the certification.
- Learning about the probable questions and gaining knowledge regarding the exam structure helps a lot. Go through the [A00-225 sample questions](#) and boost your knowledge

- Make yourself a pro through online practicing the syllabus topics. A00-225 practice tests would guide you on your strengths and weaknesses regarding the syllabus topics. Through rigorous practicing, you can improve the weaker sections too. Learn well about time management during exam and become confident gradually with practice tests.

Career Benefits:

Passing the A00-225 exam, helps a candidate to prosper highly in his career. Having the certification on the resume adds to the candidate's benefit and helps to get the best opportunities.

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