South Miami Miami-Dade Statistical Data Analyse Analyses

PREPARED FOR

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Summary

This report describes the compilation of various clean data sets into a single file containing daily observations of flow, stage, well depth and water quality at numerous locations in south Florida. The data were used to compute correlations between the various types of measurements in order to assess the degree of association between the sets of measurements. Correlations were computed for the full set of observations as well as for the wet and for the dry seasons. In addition to calculations based on daily values, correlations using monthly mean values were also computed. Both parametric and nonparametric correlations were used to summarize the associations. An example using the data to explore relationships between flow at three stations and well, stage and salinity at neighboring stations is presented.

Table of Contents

List of Tables

List of Figures

1. Introduction and Background

The South Florida Water Management District (District) has provided clean data sets for South Miami-Dade County projects on hydrology, water quality, geographic and regulatory data. The objective of this project is to perform an exploratory statistical analysis of the cleaned data. The data has been collected over different time periods by different agencies and covers a period from 1/1/2000 through 12/31/2009. Most of the data are at a daily scale although some is a more frequent interval and others are at an irregular interval.

This document describes exploratory analyses that were carried out and summarizes the output. Due to the size of the files that were produced, it is not possible to provide printed copies of the output; these were saved to disk. Some summaries are provided in this document as well as descriptions of the output. These include correlations that measure the association between the time series of measurements. The correlations were calculated for measurements on well, flow, stage, rain and water quality. Correlations for daily data as well as monthly averages were calculated. Graphical displays and multivariate methods were used to understand some of the main attributes of the data. Correlation tables and summaries were saved in files as they are too large to produce as printed output. Some summaries are presented below to help understand the analyses that were undertaken. An example exploring the relationship between flow at S20F and well, stage and salinity at several sites is presented to illustrate usage of the data.

2. Scope of work

The following six analyses were carried out:

- 1. Calculation of the correlation between stations of the same data types such as groundwater to groundwater.
- 2. Calculation of the correlations between stations of different types.
- 3. Conduct evaluations of the data sets to investigate importance of lags for correlations between variables of the same type or different types
- 4. Repeat the correlations between the variables of the same type during wet and dry seasons
- 5. Repeat the correlations between the variables of different types during wet and dry seasons
- 6. Evaluate relationships at a given station as a function of neighboring stations

In addition, additional analyses were made to try to identify patterns in the sites based on correlations. These include

- 1. Factor analysis and principal components analysis of the correlation matrix (Spearman) to identify groups of sites that are related in pattern. This was done for different types.
- 2. Where possible, create a biplot displays of sites and times
- 3. Where possible, use regression analysis to identify possible relationships between variables.

The first step in the analysis was to compile the data into a single file. Then SAS was used to perform the analyses associated with the above tasks.

3. Compilation of data sets

The first step in the analysis was the compilation of data sets. The initial data files were converted to .csv files. These files were created using consistent formats for the dates, the names of the variables and the period of the collection. Note that the word "variable" will be used to refer to a measurement at a site of a particular type. In some cases it will refer to a derived value such as the mean of several measurements.

After generating the *.csv* files, these were imported into SAS. A problem in combining the data files into one large file was noted. In some cases, a measurement was made at a particular site for a certain period of time in one study and for a second period of time in a different study. These sites were identified and the data combined into one site/variable. Following this data reduction, a "wide" file was created. The wide file is an array of times by variables. The wide data file consists of over 1000 variables measured at 3652 times. Variable names were created by combining the station name with the type of measurement. For example, S_123_Rain is the rain measurement from station S_123. Stations that included the measurement type were not altered (e.g. ANGEL_WELL_km). A master list of variables was created to use in subsequent analyses.

Because the time series for the water quality data were irregular and sparse, some of the analyses were found to be non-informative. A second data file was created in which all of the data collected for a month were averaged. This resulted in a wide data file of 120 monthly observations. Some of the water quality measurements (salinity, temperature and conductivity) were made at different depths for the same time and station. The value at the greatest depth (referred to as the bottom) was used in these cases.

A summary of the data sets including the quality codes and their frequency as well as summary information on the measurements at each site such as the mean (averaged over time), median, number of times with observations, start and stop dates of sample in given in Appendix 1. Several of the files had some irregularities and these are described in the Appendix 2. For example, site S-122, a flow station, was zero for all times and was deleted.

4. Analyses

The information below gives details on the analyses that were undertaken. The analyses in general produced rather large summary tables that are too large to reproduce in this document. Therefore, only some of the analyses are reported here. Others are provided as EXCEL spreadsheets (csv), PDFs or Word files (doc or rtf). Rather than produce correlation arrays describing relationships between different types of variables a single large array was calculated for items 2 and 5. This array includes information about the correlations between variables of the same type as well as different types. A summary of these files is given Appendix 3.

For the correlation analyses, three correlations were calculated Pearson, Spearman and Kendall. The traditional correlation measure is the Pearson measure. This measure provides a good summary of similarity in pattern of time series when the data are from a normal distribution and the relationship is linear. However, the measure is sensitive to odd observations and nonlinear relationships. The Spearman correlation is a rank-based measure of association. Ranks of the observations are computed for each time series and the correlation between the ranks is calculated. The Spearman measure does not require the assumption of normality for inference and is not sensitive to odd observations. The Kendall measure also does not require the data to be normal for good results and is not sensitive to odd observations. The Kendall correlation measure is a good measure of association for monotonic relationships between series. All three measures range from -1 to 1 with values close to one indicating strong relationships. Because the data consist of time series of observations, statistical inference (i.e. tests) on the correlations were not considered.

The analyses that were undertaken are described below:

1. Calculation of the correlation between stations of the same data types such as groundwater to groundwater.

Three types of correlations were computed: Kendall, Spearman and Pearson. The correlations were calculated for each of the five data types (flow, rain, stage, water quality and well) and stored in comma separated files. The resulting files are arrays with columns associated with measurements at each site. The first three rows of the file contain summary statistics including the mean, standard deviation and sample size. The remainder of the file gives the correlations with other variables of the same type. In the case of water quality the variables are salinity, temperature and conductivity.

The correlation matrices are stored in separate files for each type and correlation measure. For example the file *Rain Kendall.csv* has the Kendall correlation coefficients for all pairs of rain variables. The files are stored in the folder *Report Items 1-6\Item1_correlations.*

2. Calculation of the correlations between stations of different types.

This analysis was done using the full matrix of observations and by calculating the full correlation matrix. The elements of this matrix are the correlations between a measurement at a particular site and all other measurements at the same site or different sites. Again, three correlation matrices were produced using different types of correlations.

The correlations are in three *csv* files for the overall data set: *dayAll_kendall.csv, dayAll_spearman.csv and dayAll_pearson.csv*. The name indicates that the data are from daily measurements using all the data with a specified correlation. The files are located in the folder *Report Items 1-6\Item2_correlations.*

3. Conduct evaluations of the data sets to investigate importance of lags for correlations between variables of the same type or different types

To evaluate lags, a data set was generated with new variables that corresponded to the original measurements plus lags from1 to 5 days (the lag 1 measurement is the measurement from the previous day). For each lag, a prefix was added to the variable name that was L1 through L5 corresponding to the number of days. Correlations were then computed for this matrix of values. This was carried out by writing a SAS macro that looped through the variables and calculated correlations. For each variable that was considered, correlations were evaluated for all other variables, at lags of 1 through 5. The correlation with the original variable (i.e. no lag) was also computed. The 10 greatest (in absolute value) correlations were selected and output into a table. The file *LagCorrs.docx* is a Word file containing the ten greatest (absolute) correlations for each variable. Three tables were computed, for Kendall, Spearman and Pearson type correlations. Because some pairs of variables had small sample sizes, some correlations were equal to 1.0. These were dropped from the table to attempt to give useful results.

A sample table is given below for the variable stage at station _3B_SE_B_Stage_D (Table 1). The columns are as follows: TYPE is the type of correlation that was calculated, NAME is the name of the variable that was most correlated with the variable of interest. The next column's name is the name of the variable of interest and the correlations are given in the column. The last column gives the absolute value of the correlation. The name of the best variable is the variable name with a prefix added for the lag. Thus, for example, _3B_SE_B_Stage_D refers to stage at

station $3B$ _{SE}B. In the table below, _3BS1W1_H_Stage_D (no lag) had the greatest absolute Pearson correlation with the stage data at 3B_SE_B, followed by the well variable at site 3BS1W2_G. The next two variables are also at lag 0. The fifth variable is the stage variable at 3B_SE_B. Note that sample size is not listed and some of the sample sizes could be small. Also, the periods where data are observed will overlap for the variables in the table with the variable of interest however the overlap is potentially small and could be different for different pairs. The correlations should only be interpreted as measures of association between the measurements and not as cause-effects measures.

Table 1. Example of output for lag analysis. Obs refers to the rank of the correlation pair, _TYPE_ is the type of correlation, _NAME_ is the name of the variable that was correlated with the variable of interest, $3BSIW1$ H Stage D is the variable of interest with the correlation coefficient given below, absc is the absolute value of the correlation.

The complete set of correlations is given in the file *lagcorrrs.doc* (located in the Item 3 folder).

4. Repeat the correlations between the variables of the same type during wet and dry seasons

The basic data set was divided into two data sets based on season. The wet season was defined as the period from May 1 through October 14 and the dry season was from October 15 through April 30. Once the data set was generated, correlations were computed for each period using SAS. Again, three types of correlations were computed. The results are in the folder: *Item 4 Correlations WetDry by Type*. The analyses were run both on the full data and on the monthly mean data. For each type there will be two folders, one for the full data and one for the monthly data (e.g. the folder *flow* has three files for the full data with a correlation matrix for the wet season and one for the dry season in each spreadsheet. *Monthly Flow* is the folder that has similar results using monthly means.)

5. Repeat the correlations between the variables of different types during wet and dry seasons

The correlations for the complete set of variables for item 5 are contained in the files *pearsonWD.csv, kendallWD.csv* and *spearmanWD.csv,* located in the folder *Item 5 Correlations Wet Dry all.* In addition, there is a folder *correlations with Monthly means* that contains the correlations based on monthly data.

6. Evaluate relationships at a given station as a function of neighboring stations

To evaluate relationships with neighbors the metadata file that was provide with the latitudes and longitudes of the sites was used to calculate distances between sites. Some sites did not have location information and this was obtained through web search. A SAS macro was written to do the following: the wide data set was transposed to be of the form of variables by time. The latitude and longitude was added to the data from each site. For a given site of interest, the distance to other sites was calculated and the seven nearest neighbors were obtained. Correlations were then calculated for each of these variables. This was then repeated for all of the variables.

The output from the analysis is a table of correlations for each site and is described in Table 2. The columns of the table correspond to the variables from the stations closest to the station of interest. Three rows are given for each table, corresponding to the correlations using Spearman, Kendall and Pearson correlations. The column labeled *ctype* gives the correlation type. The site name is listed in the column labeled as *site_name*. An example using ANGEL_WELL_km is given below.

Table 2. Table of correlations for sites that are closest to a given site. The columns are: *Obs* is the observation number in the file, the next seven columns give the sites that are closest, ctype is the type is correlation and site_name identifies the variable of interest. The type of variable was not included to allow the table to fit in the margins. The correlations are only measures of associated and do not represent measures of causation.

The results of this analysis are in the file *neighbors correlations.pdf* in the folder *Report Items 1-6\Item 6_Correlations Neighbors.* In addition, the monthly data was used to repeat the analysis and these correlations are provided in *neighborsYM.pdf*, located in the same folder. A pdf file was used to allow indexing.

Regression analysis

The data from different stations were also evaluated using regression methods. The exploratory approach that was taken used a stepwise regression procedure to find the top variables (five or fewer) that best explained another variable. When this approach was applied generally, the method often did not produce good results because of sample size issues. When multiple variables are considered in a regression analysis, if one site has a missing value then all the sites are deleted. Thus, for example, if one site was measured in the wet season and another in the dry season then the sample size is zero after missing values are deleted and regression cannot be used. Since some sites only have a few observations, the sample size for the regression was greatly reduced. Given the large number of sites that were considered for a regression, the missing data problem greatly reduces the sample size and can lead to uninformative models or the inability to fit a model. To deal with this issue, sites with smaller sample sizes were eliminated. Data sets were generated using different minimum sample sizes and stepwise regression analyses were run. For each variable, the model selected the top variables related to the variable of interest. The minimum sample size of 3500 was selected for reported analysis and output was stored in an EXCEL file (*RegsN_3500* in the *regression* folder). A sample of the table using three variables is given below in Table 3.

Table 3. Summary of regression analysis. The two tables below are partial tables from the full EXCEL spreadsheet. _*DEPVAR_*is the name of the variable that is analyzes, *_RMSE_* is the square root of the mean square error, *Intercept* gives the value of the intercept in the model independents is the number of variables selected. Next is the list of variables that could be or are in the model. In the second table,*_IN_* is the number of variables in the model, *_P_* is the number of parameters in the regression model, _E*Df_* is the degrees of freedom for the model, *_RSQ_* is the r-square for the model and _AIC_ is the Akaike's Information Criterion. In the table a blank indicates the variable was not selected, otherwise the coefficient is given. The regressions represent possible relationships and should not be interpreted as causal relationships.

The table lists the dependent variable, the root mean square error of the model *(_RMSE_* or standard deviation), the number of independent variables in the final model, *p* is the number of parameters, *df* the degrees of freedom of the model, Rsq is the R^2 of the model. The rest of the table is the list of the parameter values. The remaining columns contain all the variables that could potentially have been in the model. Most of the entries are blank indicating the variable is not in the model. If there is a -1 then this indicates the dependent variable. For example, the first row describes the model for stage at station S179_H. The model includes 5 variables and resulted in an R^2 of 0.98, the intercept is 0.075 and the variable S121_C_Flow_SK has a coefficient of -0.0037. S20F_H_Stage_J has no coefficient, indicating that it is not important, given the other variables in the model. The other variables that are not in the model are not listed in the above table because of space. The full set of models is given in the *csv* table *RegsN_3650A.* The full model for flow at S179_H is:

S179 H_Stage_J = 0.076 - 0.0037 S121 C_Flow_SK - 0.00015 S179 S_flow - 0.0003 S21A_S_flow -0.081 S165 T Stage $D + 1.051$ S166 T Stage D.

An overall summary of the files that were generated is described in Table 4.

Table 4. Table of deliverables. This summarizes the files that are included in the report but not printed due to size.

5. Summary of some results

The interpretation of the correlation matrices and other analyses is difficult due to the number of variables in the analysis. In some cases, there are some additional analyses that might help summarize the data and relationships between the variables. Two such analyses are based on principal components analysis/factor analysis and heat maps.

Principal components analysis (PCA) is a useful way to summarize a correlation matrix by reducing the matrix using eigenvector and eigenvalues. It is typically applied to a matrix calculated from a full data set (no missing values) but may also be applied to a matrix of calculated correlations. In the examples below, the Spearman rank correlation matrix was used. The basic idea behind PCA is to form a combination or weighted average (called a *component)* of the variables that best represents the information in the data. If all the variables are closely related then a simple average of the variables would provide a good representation of the data. In general, there are likely to be several of these averages that represent different aspects of the data.

The PCA solution can be rotated to try to group variables together. These groups of variables define what are referred to as "factors" and the analysis is referred to as factor analysis. The analysis requires a selection of the number of components prior to rotation. To select the number of factors the plot of the eigenvalues versus the eigenvalue number (scree plot) was produced (see, for example, Figure 1). The number of factors with eigenvalues above one was used as a criterion, selecting fewer factors when eigenvalues were close to one. Variables or sites important to a factor were selected based on the correlations between the site measurements and the factors. These eigenvectors from the PCA were rotated using a varimax rotation to group together variables and correlations with the rotated factors computed. To facilitate summarization, these are multiplied by 100 and rounded, then sorted. Values that were "small" (as determined by SAS) were replaced with blank values to aid in interpretation.

Another way to summarize results is through a biplot display (Figure 2). This graphical display summarizes sites and times in one display. The center of the plot corresponds to the mean of the data. To interpret the display one would look for sites that are close together to represent sites that are similar in measurement pattern and times that are not near the center to identify the times that have more extreme measurements. In addition, the times may be projected onto the sites to give an approximation to the value of measurement at the site (for that time). The program uses the correlation matrix (Pearson) and the SAS program that was used requires the number of usable observations (times) to be larger than the number of variables (sites). While a useful display, interpretations may be difficult when there are missing values.

A third way to summarize the information in the correlations between variables is through a "heat map" (Figure 3). The heat map is a graphical display that tries to summarize relationships by assigning

different colors to correlations of different size. In an ideal situation one might have two sets of variables that would yield high within block correlations and low between block correlations. The map attempts to put variables that have similar, large correlations into blocks that have the same colors. In the case of two blocks, this would yield a heat map with two blocks of one color on the diagonal and off-diagonal blocks of a different color.

Note that the PCA and heatmaps are based on Spearman (rank) correlations while the biplot display is based on the Pearson (parametric) correlation. This is because the PCA analysis can be run on a correlation matrix while the biplot requires a data matrix. Because of this, the sample sizes between pairs of variables are critical and are likely to affect interpretations. Variables with missing values are omitted in the biplot display and this fact results in patterns that might be hard to interpret or connect to the PCA. For example, with the rain data, there are several sites that did not record rain during peak rain periods. In these cases, all sites during that period are not included. The result is that only one date appears to be extreme in the plot and the plot is not representative of the entire period.

The above analyses were applied to data that satisfied conditions to allow calculations. Specifically, variables that were constant were omitted, variables that were perfectly correlated with other variables were omitted, and if a value was missing from one site, the time was omitted from all sites. Heat maps are only feasible for smaller sets of variables (under 150 variables).

The plots and table below are for rain and were included because the number of variables are small relative to other types (stage, well and water quality), hence are easy to display. The types stage, well and flow are summarized in separate documents, located in the *heatmap* folder. The files are *well heatmap.doc, flow heatmap.doc and stage heatmap.doc*. These files are for correlation matrices computed using daily data; there are also files for the analysis on a monthly basis. The other graphical displays and summary analyses that are based on PCA/factor analyses are in the folder *factor analysis* and there will be two files for each type (daily and monthly).

The results from the analysis of the correlation matrices generally suggest that while there are a large number of stations associated with each type, there are some sets of sites with a moderate degree of similarity in measurement. This is indicated by the relatively high amount of variance explained by a few components and would be expected as the measurements are over time.

Figure 1. Scree plot and variance explained plot for the rain data. The plot gives the eigenvalues of the correlation matrix versus the number of factors. Look for an "elbow" or values to drop below 1. The variance explained plot displays the proportion of the total variance for each component and the cumulative proportion of the variance explained.

Table 5. Sets of variables identified by the rotated factor pattern for the rain data. Variables with similar values flagged with a star are put into a group. The numerical values are correlations between the factor and the variable and have been multiplied by 100, rounded and reported as two digits. The cutoff value is determined based on the root mean square error of the loadings.

Printed values are multiplied by 100 and rounded to the nearest integer. Values greater than 0.474122 are flagged by an '*'.

Figure 2. Principal components biplot for the rain data (complete cases). The axes correspond to the first (x-axis) and second (y-axis) principal component with the proportion of the variance of standardized variables explained in parentheses. The red names correspond to the different sites; the blue numbers correspond to the day, starting from 1/1/2000.

Figure 3. Heat map for rain variables using Spearman correlations. Colors are associated with different sizes of correlation using different colors. Note that in the case of rain, no correlations were below 0.25 so only three colors are needed.

6. Example analysis: S20F flow and salinity

To illustrate how the compiled data set may be used to aid in understanding and exploring relationships among the different data types at specific sites, consider the relationship between flow at S20F and two neighboring stations and salinity, well and stage. The main question of interest is whether or not the flow out of S20F causes changes in salinity, stage and well levels. This is a complex question as factors other than those measured may be required for a full assessment. Here a simpler question is addressed: Is there evidence that the salinity, stage and well levels are related to flow. To address this, correlations between flow variables S20F_S_flow, S21A_S_flow, S179 S flow and well, stage and salinity are calculated. The variables considered are the well variables BBCW6GW1_Well_D, BBCW6GW2_Well_D and G_3356_Well, and the stage variables S179 H_Stage_J, S20F_H_Stage_J, and S21A_H_Stage_D. Salinity at six locations close to the flow station (BISC16_Sal_BOT, BISC24_Sal_BOT, BISC34_Sal_BOT, BISC14_Sal_BOT, BISC32_Sal_BOT, and BISC22_Sal_BOT) are considered as well as four sites farther away from the S20F site, but still potentially influenced by S20F flows (BISC113_SALD_bot, BISC124_SALD_bot, BISC123_SALDbot, BISC122_SALD_bot). In addition, two sites located north of S20F that are unlikely to be influenced by S20F flows, BISC129 and BISC104, are used to estimate the degree of correlation that is likely to be associated with other factors. These sites are expected to have correlations with S20F flows that are close to zero. If the correlations are non-zero, it possibly indicates that flow and salinity are partly connected through another factor. Correlations are calculated using both daily and monthly values.

Florida City

G-3356

BBCW6GW1

BBCW6GW

 O BISC124

BISC113

 O BISC123

BISC122

Figure 4. Maps of sites used in the analysis. The first map displays all the locations (some are hidden) while the second focuses on the locations closer to S20F.

To display the relationships scatterplots are used (Figures 5-9). In Figures 5 and 6, the three flow variables are plotted along with the log transformed flow for S20F versus the well and stage data. The data exhibit nonlinear relationships and there are a variety of extreme values, both for flow and well depth. The nonlinear pattern is expected as well and stage measurements have maximum values (so an increasing or decreasing pattern followed by a flattening pattern in expected). The log transformation of flow does improve the linearity of the relationship and there are quite a few zero values that make modeling the relationship difficult. This suggests that parametric correlations are not likely to adequately capture the associations and that nonparametric correlation is more appropriate. Kendall's correlation will be used to evaluate the strength of association between flows and the other variables.

Figure 5. Scatterplot of well data versus flow data. Ls20F is the log of S20F flow.

Figure 6. Scatterplot of stage data versus flow data. Ls20F corresponds to the log transformed flow at S20.

Tables 6 and 7 give the Kendall correlations for the flow measurements and measurements for well, stage and salinity for daily data with separate calculations for wet and dry seasons. The well values are moderately correlated for BBCW6GW1_Well_D BBCW6GW2_Well_D and G_3356_Well. Correlations are strongest with S20F and weak for S179. The correlation is also moderate between the flow variables and S179_H_Stage_J. The other correlations with stage are negative and low. The wet season tends to have slightly lower correlations with flow and stage.

Table 6. Summary statistics and Kendall correlation coefficients for well and stage variables with flow. The summary statistics are for the flow variables.

Scatterplots of salinity versus flow are given in Figures 7-9. Plots for the sites close to S20F (near sites) BISC16, 24, 34, 14, 22 and 32 are in Figure 7. A general decline in salinity with flow is observed in the plots with S20F and S21A, less so with S179. The pattern is not linear although there is a general decline in salinity with flow. There are several dates with extreme flows. Interestingly the salinity is not always low for these high flow periods (for example, BISC34 has moderate salinity for a period in August-September 2005, even though the flow is high). A similar pattern is present in the plots for BISC14, 22, 32. Figure 8 plots flow and salinity for BISC113, 122, 123 and 124 (far sites). The plots show little pattern. Even though the plots are of daily observations, there are not many observations for most of the sites. Figure 9 plots the data for the two northern sites and the scatterplots for these sites look similar to those of the three far sites. Note that the sites with fewer observations were not measured during the highest flows and the range of salinity is smaller. It is also evident from the plots that sites near were sampled more frequently than the other sites.

Figure 7. Scatterplot of salinity data versus flow for near sites.

Figure 8. Scatterplot of salinity data versus flow for far sites.

Figure 9. Scatterplot of salinity data versus flow for two northern sites.

Kendall correlation coefficients for the salinity data are given in Table 7. For the dry period there is little difference in the correlations except perhaps for BISC113 and BISC124. Correlations tend to be moderate in size for S20F and S21A. Correlations for S179 flow with salinity are quite low. For the wet period the pattern is similar although correlations are higher for S179 flow with salinity relative to the dry period.

Sample sizes for the variables are given in Table 8 and indicate considerable differences in sample size. The sample sizes for the well and stage variables tend to be close to or the same as the flow variables. Sample sizes for the salinity variables are moderate for the stations close to S20F but quite small for the other stations. Because of the sample size issue and autocorrelations, monthly averages were calculated and used in a second analysis.

Plots of the monthly means for well and stage data are given in Figures 10 and 11 and Table 9 gives the correlation coefficients. Again a nonlinear pattern is present, partly due to a flattening of the relationship for high flows with higher levels present in the wet period. Patterns for the three well sites are somewhat similar.

	Obs season	TYPE_	NAME_	S20F_S_flow	S21A_S_flow	S179_S_flow
	1 Dry	MEAN		152.42	85.43	13.30
	2 Dry	STD		194.84	102.38	78.46
	3 Dry	$\mathbf N$		1983.00	1983.00	1983.00
	4 Dry	CORR	BISC113_SALD_bot	-0.29	-0.30	0.11
	5 Dry	CORR	BISC124_SALD_bot	-0.24	-0.26	0.09
	6 Dry	CORR	BISC123_SALD_bot	-0.45	-0.44	-0.08
	7 Dry	CORR	BISC122_SALD_bot	-0.48	-0.50	-0.06
	8 Dry	CORR	BISC16_Sal_BOT	-0.44	-0.44	-0.20
9	Dry	CORR	BISC24_Sal_BOT	-0.47	-0.44	-0.20
	10 Dry	CORR	BISC34_Sal_BOT	-0.45	-0.38	-0.27
	11 Dry	CORR	BISC14_Sal_BOT	-0.46	-0.44	-0.18
	12 Dry	CORR	BISC32_Sal_BOT	-0.49	-0.43	-0.14
	13 Dry	CORR	BISC22 Sal BOT	-0.48	-0.45	-0.20
	14 Dry	CORR	BISC104_SALD_bot	-0.36	-0.38	0.07
	15 Dry	CORR	BISC129_SALD_bot	-0.53	-0.51	-0.25
	16 Wet	MEAN		275.47	159.22	72.17
	17 Wet	STD		312.04	200.17	196.30
	18 Wet	${\bf N}$		1670.00	1670.00	1670.00
	19 Wet	CORR	BISC113_SALD_bot	-0.41	-0.51	-0.25
	20 Wet	CORR	BISC124_SALD_bot	-0.27	-0.41	-0.14
	21 Wet	CORR	BISC123_SALD bot	-0.44	-0.50	-0.31
22	Wet	CORR	BISC122_SALD_bot	-0.39	-0.42	-0.29
23	Wet	CORR	BISC16_Sal_BOT	-0.42	-0.38	-0.31
24	Wet	CORR	BISC24_Sal_BOT	-0.49	-0.47	-0.37
25	Wet	CORR	BISC34_Sal_BOT	-0.48	-0.46	-0.36
26	Wet	CORR	BISC14_Sal_BOT	-0.51	-0.47	-0.38
27	Wet	CORR	BISC32_Sal_BOT	-0.54	-0.51	-0.39
	28 Wet	CORR	BISC22_Sal_BOT	-0.53	-0.50	-0.39

Table 7. Kendall correlation coefficients for salinity and flow.

Table 8. Sample sizes for variables.

Figure 10. Scatterplot of well data versus flow data for monthly data.

The scatterplots for the stage variables is in Figure 11. For S179, there is an indication of relationships with flows at S20F and S21A. Again the pattern is nonlinear with an increase in stage with flow followed by a flattening for higher flows. There is less of a pattern for the other stage sites. Correlations for the well and stage variables are given in Table 9. Correlations between flow and well levels at BBCW6GW2 and G_3356 are moderately strong and positive as expected from the plots. Stage at S179 also exhibits a strong correlation. The correlation for the other stage sites is weaker and negative. Again correlations are weaker with S179 than with the other sites.

Figure 11. Scatterplot of stage data versus flow for monthly values.

Table 9. Kendall correlations for stage and well variables using monthly data. Summary statistics are for the flow variables.

Plots of monthly flow and salinity are given in Figures 12-14 and associated Kendall correlations are in Table 10. The plots and correlations indicate a general decline in salinity levels at all locations as flow increases. Note however that the axes do not have the same scales. Declines are over a greater range in the sites that are close to S20F than at sites that are farther away. In addition, the range is greater in the wet season than in the dry season.

Correlations between salinity and flow are moderately large and negative, ranging from -0.390 to 0.733. Correlations tend to greater at sites closer to S20F than those that are farther away, especially in the dry season. The greater correlations occur for sites BISC14, 24, 34, 16, 22, and 32. Correlations with sites BISC 113, 122, 123, and 124 are smaller in the dry period but are roughly the same in the wet period. Note also that sites that are not expected to be correlated with S20F have correlations around -0.45 (BISC104 and 129). These correlations suggest that a relationship with flow may be partly due to another factor, perhaps at a larger spatial scale (possibly due to weather patterns or seasonal factors).

Figure 12. Scatterplot of salinity data versus flow for close sites using monthly data

Figure 13. Scatterplots of salinity data versus flow for far stations using monthly data.

Figure 14. Scatterplots of salinity versus flow for northern stations using monthly data.

Table 10. Correlations between flow and salinity variables using monthly data. Summary statistics and sample sizes are for the flow variables.

 $\boxed{\text{Obs}}$ **season** $\boxed{\text{TVPE}}$ $\boxed{\text{NAME}}$ **S20F_S_flow S21A_S_flow S179_S_flow**

The correlations are based on data with differing sample size and this may be important if the sampling periods are quite different. Figure 15 below plots the data for salinity and flow at S20F over time for the different stations. Note that one group of stations is mostly observed in the 2004-2008 period while the other group is missing during some of that period. This suggests that care is needed in comparing the correlations across stations.

Figure 15. Time series plots of monthly data for flow and salinity.

The above analyses show how the data and correlations can be used to investigate relations between measurements of different types. The analyses were run using both daily data and monthly averages. Note that in interpreting the results the actual times of sampling are important and might limit interpretation. As indicated in the tables, the sample sizes are different for different pairs of measurements indicating that the times of sampling are not matched. Correlations can only be compared when the times are matched or the times may be viewed as a random sample of times. For example, if station BISC16 is sampled only in the wet season and station BISC32 only in the dry season, the number of samples is likely to be the same but sampling dates completely mismatched. An analysis comparing the correlations would not be useful in that case.

The analysis suggests that there is a relationship between flow at S20F and salinity levels at nearby stations. The relationship is strongest at the stations closer to the S20F location. However, the strength of the relationship is difficult to assess as there may be other factors involved in the relationship. This is suggested by the fact that salinity at stations farther from S20F is also correlated with flow at S20F, although the correlation is slightly weaker. The reason for this correlation is unclear but suggests that care needs to be taken in interpretation of relationships. The above analysis is viewed as exploratory and more complex analyses may be required to try to provide a better estimate of the relationship, especially if daily data are to be used.

Appendix 1. Data and summaries

The following data was supplied on DVDs from the District:

- 1. July 19 2010 DVD SoMiami Dade final as document and PDF Stage - 47 EXCEL files Well - 2 files: G-1362 USGS_daily_data_Scrub (a large file) 2. Data DVD Flow_station - 23 EXCEL files Rain – one EXCEL file Stage – one file with all stage data (date is before) Well data – 32 EXCEL files Final salinity graphs Original salinity graphs SoMiamiDadeScrubFinal – Access file dated 7/8/2010 3. Scrub documents SoMiamiDade_final (doc and pdf) SoMiamiDadeScrubFinalRpt_FinalsTable Revised (large file of data quality data) 4. Wells 40 EXCEL files of well water levels
- 5. Additional Flow data on 32 sites was added in an EXCEL file

Summaries of the data files created and combined from the supplied files are described below. Frequency tables provide counts of codes for each data file. No code refers to the number of observations that have no code. Summaries include the basic summary statistics as well as the start and end dates for the site measurements. Freq is the number of days with measurements.

Well data

Well19USGS data

South Florida Well data

Stage data

Stage July19 data

Rain data

п

Flow data

Flow data for additional sites

FlBay Salinity data

Diel Salinity data

Biscayne Salinity data

Appendix 2. Data preparation, decisions and comments.

Well data

Problem: G-3786_well_km has a missing value as the first value. This will cause the column to be read as a nominal variable rather than a continuous variable. A value of zero replaced the missing value. A code of DEL was used to indicate that this value should be deleted. In the SAS program observations with code=DEL were deleted.

Problem: G-613-B_well_km – line 3106 was "query returned 3104 records". This line was deleted.

Problem: G_1486_well_km: the date 20011022 occurs twice, once in record 639 and once in 662. The values are the same, the first one is deleted.

Problem: G_551_well: the date code is of a different variety (EXCEL style). I added a column with the dates in the regular format and renamed. The last line has entry "query returned 3". This was deleted.

Well Data USGS

The drawdown files have date in the format "yyyymmdd hour". The other files have daily dates (no hour) so this created a problem with input. The column for date was copied and year, month, day, and hour were extracted and used to create day. This was not a date formatted column but will suffice for calculations.

USGS data from 7-19-2010

The sheet for 254917080143301 (next to last sheet) has data on several additional sites. This does not seem to have been processed. All other sheets consist of two sets of columns, one for the original data and the other for the cleaned data. However, this sheet seems to have six sites in one set and one in the other. The sites are 254943080121501, 254950080171202, 254950080180801, 254951080194901, 255008080161801 (a total of 21,960 records).

Flow Data

Two of the sites have data past 12-31-. The files were truncated to be consistent with other data.

One site S_122 had the same value (zero) for all times and was omitted.

Bay data

Some of the data sets have several observations for each data. These are values at different depths (for example, salinity at three depths). Means were computed to produce one observation per time.

Appendix 3. Summary of data files, program files and output files. YM indicates monthly data.

