

Biomarkers Consortium Project Use of Targeted Multiplex Proteomic Strategies to Identify Plasma-Based Biomarkers in Alzheimer’s Disease

Data Primer

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Background:

The data described within this document represents the work of the Biomarkers Consortium Project “**Use of Targeted Multiplex Proteomic Strategies to Identify Plasma-Based Biomarkers in Alzheimer’s Disease**”. This project was submitted to the Biomarkers Consortium Neuroscience Steering Committee by a subgroup of the Alzheimer’s Disease Neuroimaging Initiative (ADNI) Industry Scientific Advisory Board (ISAB) for execution and was managed by a Biomarkers Consortium Project Team that includes members from academia, government and the pharmaceutical industry (for the list of Project Team members, see Appendix I); funding for this project was provided through an overage of funds raised by the Foundation for NIH for the ADNI partnership, as well as Pfizer Inc. This project is intended to be the first part of a multi-phased effort seeking to utilize samples collected by ADNI to qualify multiplex panels in both plasma and cerebrospinal fluid (CSF) to diagnose patients with Alzheimer’s Disease (AD) and monitor disease progression.

The study described herein analyzed a subset of plasma samples from the ADNI cohort in a 190 analyte multiplex immunoassay panel. The panel, referred to as the human discovery map, was developed on the Luminex xMAP platform by Rules-Based Medicine (RBM) to contain proteins previously reported in the literature to be altered as a result of cancer, cardiovascular disease, metabolic disorders and inflammation. In addition, RBM partnered with Satoris to include plasma proteins believed to be involved in cell signaling and previously reported to change in patients with Alzheimer’s disease (Ray et al., *Nat. Med.*, 2007). The analysis of plasma samples on the human discovery map is available on a commercial fee-for-service basis. The current document describes the technology and experimental design of the plasma multiplex biomarker pilot study.

Description of Multiplex Technology:

The Luminex xMAP technology uses a flow-based laser apparatus to detect fluorescent polystyrene microspheres which are loaded with different ratios of two spectrally distinct fluorochromes (Figure 1A). Using a precise ratio of the fluorochromes, up to 100 different beads can be generated such that each contains a unique color-coded signature. The beads serve as a solid phase matrix that can then be coated with either ligand or capture antibodies (Figure 1B) and then standard sandwich or competitive assay formats applied to detect the analytes. Signal is typically amplified via a reporter streptavidin-phycoerythrin conjugate. The beads are read one at a time as they pass through a flow cell on the Luminex laser instrument using a dual laser system (Figures 1C and D). One laser records the color code for individual beads (e.g. analyte ID) and the other quantitates the reporter signal (e.g. biomarker concentration). In theory, up to 100 different analytes can be measured per well per 250 ul of sample. However, dynamic range, matrix interference and cross-reactivity limit the number of analytes that can be multiplexed in one well. The actual RBM panel consists of several panels with between 3 and 24 multiplexed analytes. The combination of analytes per panel is proprietary to RBM. In addition, the dilution of samples per plate is also proprietary information.

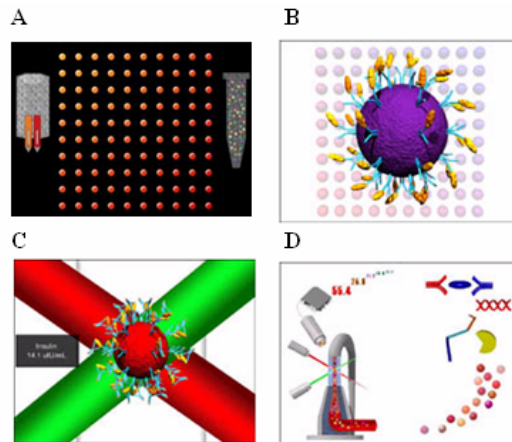


Figure 1: Luminex xMAP platform and basis of technology

RBM has attempted to validate each of the analytes on the 190 analyte panel up to clinical laboratory improvement amendment (CLIA) standards, but the assays themselves are not CLIA approved. Each analyte has an individual standard curve with between 6-8 reference standards. Each plate is run with 3 levels of QCs (low, medium and high) for each analyte. Assays are qualified based on least detectable dose (LDD - see below), precision, cross-reactivity, dilutional linearity, and spike recovery (assessment of accuracy). Cross validation to alternative methods is reported for some assays where feasible. In addition dynamic range of analytes in a healthy population is reported. The assays themselves should be considered exploratory and are not in full compliance with diagnostic standards for assays. For example, reference calibrators are diluted in a buffer and not in matrix (i.e. plasma or serum) and measurement bias is a component of the platform. Linearity of dilution and stability were not evaluated. In addition, the magnitude of batch-batch variation is not defined. RBM uses the following criteria for assay qualification:

Least Detectable Dose

The least detectable dose (LDD) is the concentration of target analyte that produces a signal that can be distinguished from that produced by a blank with 99% confidence. It is determined from the average and standard deviation of the signal for a minimum of 20 replicate determinations of the standard curve blank for each assay. Three standard deviations are added to the average of the signal, and this value is converted to concentration as interpolated from the dose response curve. The LDD is considered the most reliable lowest point for the individual assays.

Lower Assay Limit

The lower assay limit (LAL) is the lowest calibrator for individual assays. The LAL can be below the LDD. On the RBM panels, the LAL are often on the flat part of the calibration curve and are associated with significant error.

Precision

Precision is defined by the agreement between replicate measurements of the same material when measured within Run (intra-assay CV) and over a series of Runs (inter-assay or Total CV). It is determined by measuring 3 levels of controls in duplicate over a minimum of 5 Runs and provides information concerning random error expected in a test result caused by factors that vary under normal laboratory operating conditions such as pipeting, timing, mixing, and temperature.

Cross-reactivity

Cross-reactivity is the ability of an assay to differentiate and quantify the analyte of interest in the presence of other similar analytes in the sample that could have a positive or negative effect on the assay value. It is determined by testing high concentrations of each MAP analyte across all multiplexes. However, true specificity against highly related proteins is not well described in some cases.

Linearity

Linearity is the ability of the assay to obtain test results which are proportional to the concentration of analyte in a sample when serially-diluted to produce values within the dynamic range of the assay. It is determined by testing high positive or spiked samples serially-diluted in standard curve diluent. The average % recovery throughout the dilution series is then calculated as observed vs. expected concentration.

Spike Recovery

Spike recovery is performed as an assessment of accuracy which is often not possible for biological products due to the unavailability of pure “gold” standards. It is used to account for interference caused by compounds introduced from the physical composition of the sample or sample matrix that may affect the accurate measurement of the analyte. It is performed by spiking different amounts of standard spanning the assay range into standard curve diluent (control spike) and known samples. The average % recovery is calculated as the proportion of spiked standard in the sample (observed) to that of the control spike (expected) following analysis.

Correlation

Agreement of RBM multiplexed assay values to other methods is assessed by testing samples in an alternate FDA approved commercial immunoassay system, when available. This comparison of methods is performed to estimate inaccuracy or systematic error. Data from the two methods are graphed in a comparison plot and the correlation coefficient is determined.

Dynamic Range

The dynamic range is defined as the range of standard used to produce the standard curve. It is initially realized during assay development when standards are analyzed in a wide range above and below the expected concentrations using full-log dilutions. The standards are subsequently retested using reduced serial dilutions that target the useful part of the standard curve.

Matrix Interferences

Matrix interference assays are performed to determine whether the presence of interferants commonly found in serum/plasma samples introduce any systematic error in RBM MAPs. Samples spiked with high levels of hemoglobin, bilirubin, or triglyceride, as well as unspiked samples, are tested in each multiplex. % recovery is calculated as observed (spiked sample) result versus expected (unspiked sample) result.

RBM provides reports of analytes with the LDD and range for that particular run. Values that are below LDD are typically reported as LOW. High values may be reported as >top of analyte range number. If there is not sufficient volume, RBM will report as quantity not sufficient (QNS).

Listing of the Multiplex Analytes, LDD and Range:

Each analyte on the panel has a validation report that is available through RBM. Validation reports and dynamic range for serum and plasma in young healthy normal patients are known and can be obtained from Rules-Based Medicine. Table 1 lists the analytes, concentration units, LDD, LAL and range in young healthy volunteer plasma. In addition, Table 1 lists summary statistics from the QCs run during the analysis of the ADNI plasma subset (see below for more detailed discussion).

Table 1: List of Analytes, units, LDD, LAL, Dynamic range in healthy volunteers and summary QC statistics of each analyte. Analytes with one QC (red) with a CV above 25% are highlighted in yellow.

| Analyte | Units | Lower Assay Limit | Least Detectable Dose | RBM Low Plasma Range | RBM High Plasma Range | Level 1 qc | | | Level 2 qc | | | Level 3 qc | | |
|--|--------------|-------------------|-----------------------|----------------------|-----------------------|--------------|---------------|-------------|--------------|---------------|-------------|---------------|--------------|-------------|
| | | | | | | mean.1 | sd.1 | cv.1 | mean.2 | sd.2 | cv.2 | mean.3 | sd.3 | cv.3 |
| Alpha-1-Microglobulin (A1Micro) | ug/ml | 0.044 | 0.014 | 6.2 | 16 | 4.36 | 0.649 | 14.9 | 35.9 | 6.53 | 18.2 | 48.1 | 8.99 | 18.7 |
| Alpha-2-Macroglobulin (A2Macro) | mg/mL | 0.0044 | 0.013 | 0.13 | 1.0 | 0.291 | 0.0212 | 7.29 | 0.728 | 0.0567 | 7.79 | 2.79 | 0.501 | 17.9 |
| Alpha-1-Antichymotrypsin (AACT) | ug/ml | 2.2 | 0.75 | pending | pending | 26.3 | 2.74 | 10.4 | 158 | 11.2 | 7.04 | 920 | 68.5 | 7.44 |
| Alpha-1-Antitrypsin (AAT) | mg/mL | 0.0038 | 0.0095 | 1.2 | 3.1 | 0.167 | 0.037 | 22.2 | 0.513 | 0.0522 | 10.2 | 1.55 | 0.141 | 9.06 |
| Angiotensin-Converting Enzyme (ACE) | ng/ml | 0.090 | 0.20 | 36 | 279 | 2.65 | 0.196 | 7.4 | 13.5 | 0.99 | 7.35 | 108 | 9.04 | 8.35 |
| Adrenocorticotrophic Hormone (ACTH) | ng/mL | 0.035 | 0.29 | <LOW> | 2.2 | 1.31 | 0.153 | 11.7 | 4.06 | 0.44 | 10.8 | 8.13 | 0.722 | 8.87 |
| Adiponectin | ug/mL | 0.016 | 0.028 | 1.6 | 14 | 0.133 | 0.0377 | 28.3 | 0.517 | 0.0583 | 11.3 | 1.67 | 0.154 | 9.26 |
| Alpha-Fetoprotein (AFP) | ng/mL | 0.33 | 0.34 | <LOW> | 6.7 | 1.72 | 0.353 | 20.6 | 26.1 | 3.81 | 14.6 | 163 | 16.7 | 10.2 |
| Agouti-Related Protein (AGRP) | pg/mL | 22 | 125 | pending | pending | 184 | 37.4 | 20.3 | 615 | 68.4 | 11.1 | 1959 | 193 | 9.84 |
| Angiotensin-2 (ANG-2) | ng/mL | 0.029 | 0.32 | pending | pending | 25.9 | 1.79 | 6.91 | 85.7 | 4.62 | 5.4 | NA | NA | NA |
| Angiotensinogen | ng/mL | 0.39 | <0.5 | 0.55 | 670 | 181 | 8.33 | 4.59 | 429 | 21.7 | 5.06 | 772 | 41.4 | 5.36 |
| Apolipoprotein A-I (Apo A-I) | mg/mL | 0.00052 | 0.0055 | 0.19 | 0.89 | 0.0363 | 0.00526 | 14.5 | 0.127 | 0.014 | 11 | 0.436 | 0.0488 | 11.2 |
| Apolipoprotein A-II (Apo A-II) | ng/ml | 0.76 | 3.1 | pending | pending | 35.1 | 8.66 | 24.7 | 119 | 21.3 | 17.8 | 387 | 60.8 | 15.7 |
| Apolipoprotein A-IV (Apo A-IV) | ug/ml | 0.024 | 0.19 | pending | pending | 1.03 | 0.0933 | 9.02 | 3.17 | 0.191 | 6.05 | 8.77 | 0.508 | 5.79 |
| Apolipoprotein B (Apo B) | ug/ml | 8.2 | 13 | pending | pending | 140 | 13.8 | 9.87 | 744 | 64.2 | 8.63 | 2798 | 260 | 9.3 |
| Apolipoprotein C-I (Apo C-I) | ng/ml | 0.81 | 1.8 | pending | pending | 21.9 | 2.72 | 12.4 | 75.2 | 10.1 | 13.5 | 244 | 23.5 | 9.63 |
| Apolipoprotein C-II (Apo C-II) | ug/mL | 0.34 | 3.3 | 28 | 224 | 14.5 | 2.67 | 18.4 | 42.9 | 3.92 | 9.13 | 124 | 12.4 | 9.97 |
| Apolipoprotein D (Apo D) | ug/ml | 1.4 | 26 | pending | pending | 49.5 | 8.6 | 17.4 | 125 | 13.6 | 10.9 | 399 | 27.2 | 6.8 |
| Apolipoprotein E (Apo E) | ug/ml | 0.24 | 1.3 | pending | pending | 7.05 | 0.937 | 13.3 | 13.1 | 1.22 | 9.36 | 33.1 | 2.96 | 8.96 |
| Apolipoprotein H (Apo H) | ug/mL | 3.3 | 2.6 | 131 | 430 | 51.9 | 5.88 | 11.3 | 82.8 | 7.48 | 9.03 | 177 | 14.6 | 8.28 |
| Amphiregulin (AR) | pg/mL | 200 | 168 | pending | pending | 5267 | 1348 | 25.6 | 37380 | 8623 | 23.1 | 131253 | 32208 | 24.5 |
| AXL Receptor Tyrosine Kinase (AXL) | ng/mL | 0.051 | 0.051 | pending | pending | 1.5 | 0.107 | 7.13 | 4.29 | 0.318 | 7.42 | 11.3 | 0.841 | 7.43 |
| Beta-2-Microglobulin (B2M) | ug/mL | 0.0048 | 0.067 | 1.2 | 6.2 | 0.153 | 0.0154 | 10.1 | 0.31 | 0.0259 | 8.36 | 0.735 | 0.0638 | 8.68 |
| Brain-Derived Neurotrophic Factor (BDNF) | ng/mL | 0.013 | 0.023 | 0.32 | 16 | 0.859 | 0.0831 | 9.68 | 2.99 | 0.289 | 9.65 | 8.02 | 1.24 | 15.4 |
| B Lymphocyte Chemoattractant (BLC) | pg/ml | 0.65 | 55 | pending | pending | 68.1 | 6.63 | 9.74 | 205 | 20.2 | 9.86 | 524 | 81.9 | 15.6 |
| Bone Morphogenetic Protein 6 (BMP-6) | ng/mL | 0.034 | 0.55 | pending | pending | 1.75 | 0.25 | 14.3 | 5.63 | 0.631 | 11.2 | 15.9 | 0.99 | 6.24 |
| Brain Natriuretic Peptide (BNP) | pg/ml | 25 | 102 | pending | pending | 820 | 316 | 38.5 | 4327 | 443 | 10.2 | 7603 | 833 | 11 |
| Betacellulin (BTC) | pg/mL | 11 | 100 | pending | pending | 593 | 45.1 | 7.61 | 1707 | 99.2 | 5.81 | 5127 | 335 | 6.53 |
| Complement C3 (C3) | mg/mL | 0.0033 | 0.0046 | 0.76 | 2.1 | 0.149 | 0.0181 | 12.2 | 0.453 | 0.043 | 9.49 | 1.37 | 0.0976 | 7.14 |
| Cancer Antigen 125 (CA-125) | U/mL | 3.8 | 8.7 | <LOW> | 12 | 60.5 | 6.32 | 10.4 | 513 | 42.2 | 8.24 | 5213 | 489 | 9.38 |
| Cancer Antigen 19-9 (CA-19-9) | U/mL | 0.43 | 1.6 | <LOW> | 9.2 | 16.3 | 3.6 | 22 | 106 | 9.73 | 9.15 | 510 | 43.3 | 8.49 |
| Calbindin | ng/ml | 2.2 | 2.9 | <LOW> | 5.0 | 15 | 2.27 | 15.1 | 54.7 | 6.76 | 12.4 | 213 | 27.2 | 12.8 |
| Calcitonin | pg/mL | 0.91 | 12 | <LOW> | 12 | 38.9 | 3.22 | 8.26 | 185 | 13.7 | 7.37 | 1001 | 93.4 | 9.33 |
| CD 40 antigen (CD40) | ng/mL | 0.0095 | 0.022 | 0.17 | 1.5 | 0.455 | 0.0259 | 5.68 | 1.57 | 0.118 | 7.5 | 6.9 | 0.51 | 7.39 |
| CD40 Ligand (CD40-L) | ng/mL | 0.0080 | 0.013 | <LOW> | 1.1 | 0.175 | 0.0239 | 13.7 | 1.06 | 0.16 | 15.1 | 3.84 | 0.565 | 14.7 |
| CD5 (CD5L) | ng/ml | 43 | 40 | pending | pending | 301 | 44.3 | 14.8 | 841 | 78.1 | 9.28 | 2153 | 162 | 7.54 |
| Carcinoembryonic Antigen (CEA) | ng/mL | 0.14 | 0.19 | <LOW> | 4.8 | 3.68 | 0.328 | 8.91 | 20.9 | 1.58 | 7.55 | 128 | 6.51 | 5.08 |
| Chromogranin-A (CgA) | ng/mL | 4.2 | 10 | pending | pending | 60 | 6.59 | 11.1 | 256 | 22 | 8.56 | 755 | 90.4 | 12 |

Table 1 Continued

| Analyte | Units | Lower Assay Limit | Least Detectable Dose | RBM Low Plasma Range | RBM High Plasma Range | Level 1 qc | | | Level 2 qc | | | Level 3 qc | | |
|--|--------|-------------------|-----------------------|----------------------|-----------------------|------------|---------|------|------------|---------|------|------------|--------|--------|
| | | | | | | mean.1 | sd.1 | cv.1 | mean.2 | sd.2 | cv.2 | mean.3 | sd.3 | cv.3 |
| | | | | | | | | | | | | | | |
| Creatine Kinase-MB (CK-MB) | ng/mL | 0.019 | 0.16 | <LOW> | 1.1 | 0.627 | 0.0793 | 12.6 | 3.85 | 0.577 | 15 | 28.9 | 4.93 | 17.1 |
| Clusterin (CLU) | ug/ml | 0.77 | 1.3 | pending | pending | 52.4 | 3.16 | 6.03 | 164 | 19.6 | 12 | 568 | 50.6 | 9.08 |
| Ciliary Neurotrophic Factor (CNTF) | pg/mL | 7.4 | 92 | pending | pending | 96.6 | 10.6 | 11 | 407 | 27.8 | 6.83 | 1369 | 88.1 | 6.43 |
| Complement Factor H | ug/ml | 8.3 | 13 | pending | pending | 413 | 33.2 | 8.05 | 1195 | 80.7 | 6.75 | 3174 | 210 | 6.6 |
| Cortisol (Cortisol) | ng/ml | 2.7 | 2.0 | 48 | 250 | 165 | 26.3 | 15.9 | 440 | 87.2 | 19.8 | 1250 | 263 | 21 |
| C-peptide | ng/ml | 0.021 | 0.23 | pending | pending | 1.7 | 0.0756 | 4.45 | 6.92 | 0.336 | 4.86 | 26.1 | 2.5 | 9.58 |
| C-Reactive Protein (CRP) | ug/mL | 0.010 | 0.0065 | 0.18 | 35 | 0.234 | 0.0504 | 21.5 | 0.728 | 0.176 | 24.2 | 2.16 | 0.333 | 15.4 |
| Connective Tissue Growth Factor (CTGF) | ng/ml | 0.36 | 3.3 | 2.8 | 7.5 | 0.991 | 0.406 | 41 | 5.74 | 0.652 | 11.4 | 19.3 | 3.06 | 15.8 |
| Cystatin-C | ng/ml | 16 | 76 | 608 | 1170 | 10.2 | 0.947 | 9.31 | 34.9 | 2.47 | 7.1 | 89.4 | 8.52 | 9.63 |
| Epidermal Growth Factor (EGF) | pg/mL | 1.7 | 3.0 | <LOW> | 505 | 46 | 4.9 | 10.6 | 246 | 18.8 | 7.64 | 1261 | 115 | 9.13 |
| Epidermal Growth Factor Receptor (EGFR) | ng/mL | 0.050 | 0.042 | pending | pending | 0.674 | 0.115 | 17 | 7.73 | 0.724 | 9.36 | 29.7 | 3.56 | 12 |
| Epithelial-Derived Neutrophil-Activating Protein 78 (ENA-78) | ng/mL | 0.010 | 0.033 | 0.069 | 5.3 | 0.688 | 0.0446 | 6.48 | 2.38 | 0.182 | 7.65 | 5.69 | 0.691 | 12.1 |
| EN-RAGE | ng/mL | 0.35 | | 4.6 | 592 | 10.2 | 1.74 | 17.1 | 24.9 | 2.84 | 11.4 | 51.2 | 4.63 | 9.05 |
| Eotaxin-1 | pg/mL | 4.8 | 19 | <LOW> | 177 | 101 | 15.1 | 15 | 595 | 36.3 | 6.1 | 2383 | 253 | 10.6 |
| Eotaxin-3 | pg/mL | 98 | 70 | <LOW> | 177 | 6657 | 776 | 11.7 | 22693 | 2109 | 9.29 | 59533 | 5694 | 9.66 |
| Erythropoietin (EPO) | pg/mL | 1.7 | 14 | <LOW> | 284 | 97.1 | 12 | 12.3 | 487 | 28.1 | 5.77 | 2516 | 320 | 12.7 |
| Eprexigulin (EPR) | pg/mL | 27 | 17 | pending | pending | 209 | 40.3 | 19.3 | 1381 | 125 | 9.06 | 5361 | 438 | 8.18 |
| E-Selectin | ng/mL | 0.080 | 0.25 | pending | pending | 3.87 | 0.276 | 7.14 | 18.8 | 0.862 | 4.58 | 80.5 | 4.24 | 5.27 |
| Endothelin-1 (ET-1) | pg/mL | 12 | 14 | <LOW> | 26 | 63.8 | 14.6 | 22.9 | 485 | 31.3 | 6.45 | 2799 | 201 | 7.17 |
| Fatty Acid-Binding Protein, heart (FABP, heart) | ng/mL | 0.17 | 0.71 | <LOW> | 10 | 97.1 | 10.3 | 10.6 | 581 | 46 | 7.92 | 89.3 | 6.43 | 7.2 |
| Factor VII | ng/mL | 0.44 | 0.57 | 106 | 443 | 2.45 | 0.311 | 12.7 | 41.7 | 3.13 | 7.52 | 387 | 29.4 | 7.6 |
| FASLG Receptor (FAS) | ng/mL | 0.40 | 1.2 | pending | pending | 49.2 | 2.78 | 5.66 | 161 | 5.46 | 3.39 | 509 | 22.6 | 4.44 |
| Fas Ligand (FasL) | pg/mL | 1.7 | 16 | pending | pending | 54.1 | 4.61 | 8.52 | 174 | 15.5 | 8.91 | 543 | 36.2 | 6.67 |
| Fetuin-A | ug/ml | 4.7 | 3.0 | pending | pending | 254 | 18.2 | 7.17 | 709 | 58.5 | 8.25 | 2339 | 196 | 8.36 |
| Fibroblast Growth Factor 4 (FGF-4) | pg/mL | 5.8 | 248 | pending | pending | 717 | 76.2 | 10.6 | 2541 | 190 | 7.47 | 7619 | 475 | 6.23 |
| Fibroblast Growth Factor basic (FGF-basic) | pg/mL | 16 | 249 | <LOW> | 2000 | 203 | 102 | 50.2 | 2357 | 192 | 8.15 | 36707 | 2536 | 6.91 |
| Fibrinogen | mg/mL | 0.017 | 0.014 | 2.2 | 8.0 | 0.0956 | 0.0197 | 20.6 | 0.295 | 0.034 | 11.5 | 0.89 | 0.114 | 12.8 |
| Ferritin (FRTN) | ng/mL | 0.90 | 1.5 | 5.0 | 552 | 28.5 | 4.27 | 15 | 78.6 | 6.77 | 8.61 | 217 | 23.9 | 11 |
| Follicle-Stimulating Hormone (FSH) | mIU/mL | 0.13 | 2.7 | 2.0 | 98 | 2.03 | 0.237 | 11.7 | 9.91 | 0.866 | 8.74 | 69.5 | 5.14 | 7.39 |
| Granulocyte Colony-Stimulating Factor (G-CSF) | pg/mL | 2.5 | 3.4 | <LOW> | 37 | 48.3 | 7.27 | 15.1 | 285 | 34.2 | 1.2 | 1259 | 134 | 10.6 |
| Growth Hormone (GH) | ng/mL | 0.020 | 0.092 | <LOW> | 4.4 | 0.414 | 0.0392 | 9.48 | 2.32 | 0.178 | 7.68 | 16.6 | 0.91 | 5.48 |
| Glucagon-like Peptide 1, total (GLP-1 total) | pg/ml | 0.53 | 7.5 | <LOW> | 812 | 18.7 | 3.63 | 19.4 | 125 | 19.3 | 15.5 | 904 | 80.9 | 8.94 |
| Glucagon | pg/ml | 12 | 200 | <LOW> | 1477 | 418 | 76.9 | 18.4 | 2335 | 219 | 9.4 | 7880 | 2021 | 25.7 |
| Granulocyte-Macrophage Colony-Stimulating Factor (GM-CSF) | pg/mL | 5.7 | 39 | <LOW> | 152 | 132 | 17.7 | 13.3 | 329 | 47 | 14.3 | 867 | 68.1 | 7.86 |
| Growth-Regulated alpha protein (GRO-alpha) | pg/mL | 3.5 | 103 | pending | pending | 130 | 10.4 | 8.02 | 433 | 31.3 | 7.23 | 1285 | 84.9 | 6.61 |
| Glutathione S-Transferase alpha (GST-alpha) | ng/ml | 0.42 | 3.1 | 6.7 | 62 | 16.6 | 2.38 | 14.4 | 55.3 | 10.4 | 18.8 | 194 | 43.1 | 22.2 |
| Haptoglobin | mg/mL | 0.0098 | 0.028 | 0.047 | 7.6 | 0.0867 | 0.00963 | 11.1 | 0.324 | 0.0325 | 10 | 1.05 | 0.0995 | 9.46 |
| Heparin-Binding EGF-Like Growth Factor (HB-EGF) | pg/mL | 29 | 13 | pending | pending | 237 | 67.1 | 28.3 | 6831 | 978 | 14.3 | 26440 | 2935 | 11.1 |
| Chemokine CC-4 (HCC-4) | ng/mL | 0.0077 | 0.024 | pending | pending | 0.203 | 0.0103 | 5.1 | 0.637 | 0.0337 | 5.3 | 2.03 | 0.105 | 5.15 |
| Hepatocyte Growth Factor (HGF) | ng/mL | 0.061 | 0.20 | pending | pending | 6.49 | 0.697 | 10.7 | 22.1 | 2.59 | 11.7 | 91.9 | 17.3 | 18.9 |
| Heat Shock Protein 60 (HSP-60) | ng/ml | 2.0 | 18 | pending | pending | 103 | 12.2 | 11.8 | 299 | 55.2 | 18.5 | 967 | 147 | 15.2 |
| T Lymphocyte-Secreted Protein 1-309 (I-309) | pg/mL | 6.1 | 95 | pending | pending | 584 | 53.9 | 9.23 | 1985 | 128 | 6.42 | 6289 | 325 | 5.17 |
| Intercellular Adhesion Molecule 1 (ICAM-1) | ng/mL | 0.087 | 2.1 | 42 | 213 | 2.48 | 0.673 | 27.1 | 13.7 | 1.1 | 8.01 | 80.3 | 11.5 | 14.3 |
| Interferon gamma (IFN-gamma) | pg/mL | 0.88 | 3.2 | <LOW> | 9.5 | 52.5 | 5.01 | 9.55 | 261 | 23 | 8.8 | 1343 | 76.3 | 5.68 |
| Immunoglobulin A (IgA) | mg/mL | 0.0073 | 0.043 | 0.58 | 5.6 | 0.336 | 0.0481 | 14.3 | 0.842 | 0.0684 | 8.12 | 2.44 | 0.261 | 10.7 |
| Immunoglobulin E (IgE) | ng/mL | 1.3 | 3.9 | <LOW> | 770 | 2.65 | 1.2 | 45.4 | 57.3 | 13.1 | 22.9 | 418 | 41.7 | 9.98 |
| Insulin-like Growth Factor-Binding Protein 2 (IGFBP-2) | ng/mL | 0.27 | 1.2 | pending | pending | 6.75 | 0.692 | 10.2 | 19.3 | 1.22 | 6.35 | 55.5 | 2.92 | 5.27 |
| Insulin-like Growth Factor 1 (IGF-1) | ng/mL | 0.86 | 6.0 | <LOW> | 177 | 50.1 | 6.63 | 13.2 | 143 | 12.8 | 8.99 | 749 | 54.8 | 7.32 |
| Immunoglobulin M (IGM) | mg/mL | 0.014 | 0.080 | 0.24 | 3.3 | 0.211 | 0.063 | 29.8 | 0.707 | 0.104 | 14.7 | 2.11 | 0.253 | 12 |
| Interleukin-1 alpha (IL-1 alpha) | ng/mL | 0.00098 | 0.0071 | <LOW> | 0.35 | 0.0151 | 0.00136 | 8.96 | 0.0478 | 0.00355 | 7.43 | 0.144 | 0.0091 | 6.32 |
| Interleukin-1 beta (IL-1 beta) | pg/mL | 0.35 | 1.1 | <LOW> | 8.7 | 5.13 | 0.828 | 16.2 | 28.9 | 3.06 | 10.6 | 142 | 11.2 | 7.85 |
| Interleukin-10 (IL-10) | pg/mL | 2.9 | 6.3 | 1.8 | 38 | 185 | 13.7 | 7.4 | 632 | 40.1 | 6.35 | 2677 | 224 | 8.38 |
| Interleukin-11 (IL-11) | pg/mL | 17 | 56 | pending | pending | 831 | 56.2 | 6.77 | 2224 | 213 | 9.58 | 6048 | 348 | 5.76 |
| Interleukin-12 Subunit p40 (IL-12p40) | ng/mL | 0.096 | 0.23 | <LOW> | 2.7 | 5.35 | 0.504 | 9.42 | 19.4 | 1.45 | 7.5 | 77.5 | 5.53 | 7.13 |
| Interleukin-12 Subunit p70 (IL-12p70) | pg/mL | 21 | 26 | <LOW> | 165 | 1528 | 119 | 7.78 | 5530 | 362 | 6.54 | 20580 | 1230 | 5.98 |
| Interleukin-13 (IL-13) | pg/mL | 1.4 | 16 | <LOW> | 133 | 38.5 | 4.7 | 12.2 | 85.9 | 9.23 | 10.7 | 236 | 14.5 | 6.12 |
| Interleukin-15 (IL-15) | ng/mL | 0.025 | 0.58 | <LOW> | 4.6 | 0.506 | 0.189 | 37.3 | 1.71 | 0.237 | 13.9 | 6.21 | 0.599 | 9.66 |
| Interleukin-16 (IL-16) | pg/mL | 10 | 16 | 232 | 3380 | 84.6 | 9.42 | 11.1 | 330 | 18.6 | 5.64 | 1217 | 73.9 | 6.07 |
| Interleukin-18 (IL-18) | pg/mL | 7.0 | 25 | 72 | 1020 | 356 | 23.4 | 6.57 | 1314 | 147 | 11.2 | 6163 | 635 | 10.3 |
| Interleukin-1 receptor antagonist (IL-1ra) | pg/mL | 10 | 44 | 17 | 622 | 238 | 42.5 | 17.8 | 1373 | 192 | 14 | 6619 | 492 | 7.44 |
| Interleukin-2 (IL-2) | pg/mL | 7.7 | 19 | <LOW> | 61 | 156 | 15.1 | 9.71 | 549 | 50.6 | 9.22 | 2021 | 172 | 8.51 |
| Interleukin-25 (IL-25) | pg/mL | 1.9 | 13 | <LOW> | 124 | 664 | 58 | 8.72 | 2588 | 219 | 8.46 | 7207 | 539 | 7.48 |
| Interleukin-3 (IL-3) | ng/mL | 0.0022 | 0.049 | <LOW> | 1.2 | 0.0297 | 0.00622 | 20.9 | 0.0934 | 0.0153 | 16.3 | 0.276 | 0.0241 | 8.75 |
| Interleukin-4 (IL-4) | pg/mL | 7.6 | 34 | <LOW> | 103 | 647 | 66.8 | 10.3 | 2234 | 181 | 8.1 | 6834 | 510 | 7.46 |
| Interleukin-5 (IL-5) | pg/mL | 0.83 | 4.0 | <LOW> | 62 | 5.53 | 1.48 | 26.7 | 15.6 | 1.88 | 12.1 | 28.6 | 5.8 | 20.3 |
| Interleukin-6 (IL-6) | pg/mL | 0.95 | 1.8 | <LOW> | 25 | 34.8 | 4.55 | 13.1 | 184 | 18 | 9.79 | 1019 | 84.5 | 8.29 |
| Interleukin-6 receptor (IL-6r) | ng/mL | 0.013 | 0.030 | pending | pending | 1.22 | 0.0561 | 4.6 | 5.99 | 0.343 | 5.73 | 25.1 | 0.99 | 3.94 |
| Interleukin-7 (IL-7) | pg/mL | 3.6 | 59 | 3.7 | 125 | 628 | 52.6 | 8.38 | 2087 | 109 | 5.23 | 6871 | 509 | 7.41 |
| Interleukin-8 (IL-8) | pg/mL | 2.1 | 3.8 | <LOW> | 59 | 71.5 | 6.5 | 9.09 | 246 | 20.9 | 8.51 | 907 | 57.3 | 6.32 |
| Insulin | uIU/mL | 0.051 | 0.66 | <LOW> | 34 | 4.74 | 0.36 | 7.6 | 16.6 | 1.06 | 6.36 | 29.8 | 1.52 | 5.1 |
| Interferon gamma Induced Protein 10 (IP-10) | pg/ml | 9.0 | 30 | pending | pending | 444 | 40.8 | 9.19 | 2075 | 113 | 5.45 | 9697 | 934 | 9.63 |
| Kidney Injury Molecule-1 (KIM-1) | ng/ml | 0.0080 | 0.017 | 0.053 | 0.57 | 0.113 | 0.0139 | 12.3 | 0.337 | 0.0324 | 9.6 | 1.19 | 0.119 | 9.96 |
| Leptin | ng/mL | 0.026 | 0.12 | 0.41 | 41 | 0.8 | 0.0614 | 7.68 | 3.95 | 0.318 | 8.05 | 22.7 | 1.91 | 8.45 |
| Luteinizing Hormone (LH) | mIU/mL | 0.23 | 0.80 | 4.2 | 151 | 6.04 | 0.642 | 10.6 | 28.7 | 2.43 | 8.47 | 212 | 15.1 | 7.11 |
| Lectin-Like Oxidized LDL Receptor 1 (LOX-1) | ng/mL | 0.64 | 0.82 | pending | pending | 6.27 | 0.858 | 13.7 | 93.3 | 12.6 | 13.5 | 391 | 38.8 | 9.91 |
| Apolipoprotein(a) (Lp(a)) | ug/mL | 1.3 | 3.3 | 10 | 2888 | 9.1 | 1.94 | 21.3 | 30.7 | 4.2 | 13.7 | 95.3 | 10.4 | 10.9 |
| Lymphactin | ng/mL | 0.029 | 0.23 | <LOW> | 0.57 | 1.87 | 0.09 | 4.82 | 5.79 | 0.379 | 6.54 | 15.7 | 1.49 | 9.45</ |

Table 1 Continued

| Analyte | Units | Lower Assay Limit | Least Detectable Dose | RBM Low Plasma Range | RBM High Plasma Range | Level 1 qc | | | Level 2 qc | | | Level 3 qc | | |
|--|--------|-------------------|-----------------------|----------------------|-----------------------|--|---------|------|------------|--------|------|------------|--------|------|
| | | | | | | mean.1 | sd.1 | cv.1 | mean.2 | sd.2 | cv.2 | mean.3 | sd.3 | cv.3 |
| | | | | | | Monocyte Chemotactic Protein 1 (MCP-1) | pg/mL | 4.8 | 40 | 35 | 401 | 126 | 10.8 | 8.58 |
| Monocyte Chemotactic Protein 2 (MCP-2) | pg/ml | 2.1 | 15 | pending | pending | 54.8 | 3.63 | 6.62 | 236 | 10.7 | 4.52 | 690 | 41 | 5.94 |
| Monocyte Chemotactic Protein 3 (MCP-3) | pg/mL | 0.57 | 17 | pending | pending | 25.7 | 2.5 | 9.73 | 88.1 | 5.77 | 6.54 | 278 | 18.6 | 6.68 |
| Monocyte Chemotactic Protein 4 (MCP-4) | pg/ml | 7.7 | 30 | pending | pending | 305 | 34.6 | 11.4 | 1517 | 134 | 8.84 | 8102 | 469 | 5.78 |
| Macrophage Colony-Stimulating Factor 1 (M-CSF) | ng/mL | 0.019 | 0.10 | pending | pending | 0.423 | 0.0498 | 11.8 | 2 | 0.156 | 7.79 | 9.67 | 0.511 | 5.28 |
| Malondialdehyde-Modified Low-Density Lipoprotein (MDA-LDL) | ng/mL | 51 | 55 | pending | pending | 334 | 182 | 54.6 | 5014 | 1390 | 27.7 | 10568 | 2856 | 27 |
| Macrophage-Derived Chemokine (MDC) | pg/mL | 4.0 | 12 | 162 | 774 | 148 | 8.04 | 5.45 | 528 | 30.9 | 5.86 | 1621 | 167 | 10.3 |
| Macrophage Migration Inhibitory Factor (MIF) | ng/mL | 0.0040 | 0.017 | pending | pending | 0.137 | 0.0145 | 10.6 | 0.431 | 0.0365 | 8.48 | 1.26 | 0.0986 | 7.82 |
| Monokine Induced by Gamma Interferon (MIG) | pg/ml | 38 | 100 | pending | pending | 7513 | 834 | 11.1 | 31953 | 4435 | 13.9 | 161667 | 13010 | 8.05 |
| Macrophage Inflammatory Protein-1 alpha (MIP-1 alpha) | pg/mL | 7.7 | 19 | <LOW> | 89 | 80.1 | 16 | 20 | 514 | 33.1 | 6.45 | 2803 | 156 | 5.58 |
| Macrophage Inflammatory Protein-1 beta (MIP-1 beta) | pg/mL | 5.9 | 21 | 25 | 595 | 261 | 17.4 | 6.69 | 1139 | 102 | 8.92 | 15373 | 1335 | 8.69 |
| Macrophage Inflammatory Protein-3 alpha (MIP-3 alpha) | pg/ml | 3.7 | 16 | pending | pending | 486 | 38.9 | 8.01 | 2161 | 120 | 5.55 | 7231 | 378 | 5.22 |
| Matrix Metalloproteinase-1 (MMP-1) | ng/ml | 0.11 | 0.37 | pending | pending | 3.09 | 0.309 | 10 | 15.2 | 0.862 | 5.67 | 75.9 | 3.63 | 4.79 |
| Matrix Metalloproteinase-10 (MMP-10) | ng/ml | 0.010 | 0.010 | pending | pending | 1.54 | 0.0828 | 5.38 | 7.87 | 0.446 | 5.67 | 38.6 | 1.99 | 5.16 |
| Matrix Metalloproteinase-2 (MMP-2) | ng/mL | 1.7 | 15 | 183 | 3070 | 36.1 | 4.72 | 13.1 | 251 | 25.9 | 10.3 | 1506 | 130 | 8.66 |
| Matrix Metalloproteinase-3 (MMP-3) | ng/mL | 0.016 | 0.066 | <LOW> | 1.8 | 0.27 | 0.0455 | 16.9 | 2.05 | 0.185 | 8.99 | 15 | 1.13 | 7.56 |
| Matrix Metalloproteinase-7 (MMP-7) | ng/ml | 0.020 | 0.50 | pending | pending | 0.212 | 0.0283 | 13.4 | 1.11 | 0.064 | 5.75 | 5.83 | 0.358 | 6.14 |
| Matrix Metalloproteinase-9 (MMP-9) | ng/mL | 0.89 | 10 | <LOW> | 1050 | 15.3 | 7.42 | 48.7 | 88.1 | 9.86 | 11.2 | 788 | 52.2 | 6.62 |
| Matrix Metalloproteinase-9, total (MMP-9, total) | ng/ml | 0.72 | 0.46 | pending | pending | 2.99 | 1.1 | 36.6 | 23.3 | 1.79 | 7.7 | 119 | 8 | 6.7 |
| Myeloid Progenitor Inhibitory Factor 1 (MPIF-1) | ng/mL | 0.010 | 0.39 | pending | pending | 1.33 | 0.135 | 10.1 | 6.81 | 0.424 | 6.22 | 34.5 | 3.09 | 8.95 |
| Myeloperoxidase (MPO) | ng/mL | 22 | 32 | <LOW> | 1110 | 809 | 193 | 23.8 | 2285 | 404 | 17.7 | 5643 | 611 | 10.8 |
| Myoglobin | ng/mL | 0.22 | 2.0 | 3.6 | 37 | 3.94 | 0.481 | 12.2 | 14.5 | 1.41 | 9.68 | 45.3 | 3.01 | 6.65 |
| Neutrophil Gelatinase-Associated Lipocalin (NGAL) | ng/ml | 19 | 400 | 89 | 375 | 17.6 | 2.47 | 14 | 52.3 | 7.71 | 14.8 | 154 | 10.7 | 6.98 |
| Nerve Growth Factor beta (NGF-beta) | ng/mL | 0.016 | 0.056 | pending | pending | 1.09 | 0.102 | 9.32 | 3.55 | 0.177 | 4.98 | 8.8 | 0.64 | 7.28 |
| Neuronal Cell Adhesion Molecule (NCAM) | ng/mL | 0.073 | 0.065 | pending | pending | 4.05 | 0.233 | 5.74 | 11.9 | 0.704 | 5.9 | 21.1 | 2.26 | 10.7 |
| Osteopontin | ng/ml | 0.17 | 1.2 | 4.1 | 25 | 399 | 29.8 | 7.45 | 1507 | 80.8 | 5.36 | 5482 | 374 | 6.81 |
| Plasminogen Activator Inhibitor 1 (PAI-1) | ng/mL | 1.0 | 1.1 | 10 | 87 | 11.8 | 1.24 | 10.4 | 36.5 | 3.78 | 10.3 | 108 | 8.64 | 7.99 |
| Prostatic Acid Phosphatase (PAP) | ng/mL | 0.017 | 0.011 | 0.058 | 0.54 | 0.255 | 0.029 | 11.4 | 1.34 | 0.14 | 10.5 | 6.33 | 0.735 | 11.6 |
| Pregnancy-Associated Plasma Protein A (PAPP-A) | mlU/mL | 0.0028 | 0.011 | <LOW> | 0.48 | 0.0353 | 0.0116 | 32.8 | 0.283 | 0.0415 | 14.7 | 1.56 | 0.172 | 11 |
| Pulmonary and Activation-Regulated Chemokine (PARC) | ng/mL | 4.3 | 0.070 | pending | pending | 0.659 | 0.0431 | 6.54 | 1.97 | 0.149 | 7.53 | 4.93 | 0.237 | 4.82 |
| Platelet-Derived Growth Factor BB (PDGF-BB) | pg/ml | 43 | 139 | pending | pending | 447 | 111 | 24.9 | 7920 | 609 | 7.69 | 28353 | 2594 | 9.15 |
| Placenta Growth Factor (PLGF) | pg/ml | 13 | 9.3 | pending | pending | 149 | 19.6 | 13.2 | 1836 | 126 | 6.86 | 5773 | 535 | 9.27 |
| Pancreatic Polypeptide (PPP) | pg/ml | 0.68 | 5.0 | 38 | 1531 | 11.5 | 1.15 | 10 | 59.6 | 4.66 | 7.81 | 502 | 40.6 | 8.09 |
| Prolactin (PRL) | ng/ml | 0.023 | 0.50 | 0.88 | 42 | 0.768 | 0.0731 | 9.52 | 3.95 | 0.27 | 6.83 | 31 | 1.93 | 6.22 |
| Progesterone | ng/ml | 0.55 | 0.72 | <LOW> | 60 | 30.1 | 9.16 | 30.8 | 112 | 24.3 | 21.6 | 397 | 65.1 | 16.4 |
| Proinsulin, Intact | pM | 1.00 | 0.60 | pending | pending | 20.3 | 1.29 | 6.35 | 75.7 | 4.11 | 5.43 | 253 | 9.55 | 3.78 |
| Proinsulin, Total | pM | 4.2 | 2.5 | pending | pending | 86.4 | 4.72 | 5.46 | 322 | 19.7 | 6.12 | 1075 | 49.4 | 4.59 |
| Prostate-Specific Antigen, Free (PSA-f) | ng/mL | 0.0070 | 0.028 | <LOW> | 1.6 | 0.205 | 0.0226 | 11 | 0.713 | 0.0752 | 10.5 | 1.91 | 0.158 | 8.28 |
| Peptide YY (PYY) | pg/mL | 17 | <50 | <LOW> | 396 | 1488 | 138 | 9.27 | 4623 | 454 | 9.82 | 13680 | 1215 | 8.88 |
| Receptor for advanced glycosylation end products (RAGE) | ng/mL | 0.084 | 0.10 | pending | pending | 0.943 | 0.0672 | 7.12 | 3.54 | 0.302 | 8.53 | 16 | 1.41 | 8.84 |
| T-Cell-Specific Protein RANTES (RANTES) | ng/mL | 0.019 | 0.033 | 2.6 | 83 | 0.828 | 0.127 | 15.4 | 2.91 | 0.32 | 11 | 8.98 | 0.963 | 10.7 |
| Resistin | ng/ml | 0.020 | 0.030 | 0.85 | 10.0 | 0.0445 | 0.00554 | 12.5 | 0.601 | 0.057 | 9.49 | 4.6 | 0.417 | 9.08 |
| S100 calcium-binding protein B (S100-B) | ng/mL | 0.070 | 0.30 | pending | pending | 5.41 | 0.668 | 12.4 | 19.3 | 1.58 | 8.2 | 102 | 11.3 | 11.1 |
| Serum Amyloid P-Component (SAP) | ug/mL | 0.074 | 0.16 | 15 | 50 | 1.26 | 0.25 | 19.9 | 3.83 | 0.359 | 9.39 | 11 | 1.02 | 9.28 |
| Stem Cell Factor (SCF) | pg/mL | 15 | 43 | <LOW> | 281 | 826 | 53.8 | 6.52 | 2647 | 235 | 8.88 | 7111 | 648 | 9.11 |
| Secretin | ng/mL | 0.021 | <0.015 | <LOW> | 45 | 0.2 | NA | NA | 5.96 | 0.974 | 16.3 | 24.7 | 2.06 | 8.35 |
| Serum Glutamic Oxaloacetic Transaminase (SGOT) | ug/mL | 0.11 | 2.5 | 3.9 | 28 | 7.35 | 1.61 | 21.9 | 23.9 | 2.76 | 11.5 | 87.3 | 16.4 | 18.8 |
| Sex Hormone-Binding Globulin (SHBG) | nmol/L | 1.4 | 2.3 | 12 | 106 | 2.19 | 0.399 | 18.2 | 7.23 | 1.04 | 14.4 | 23.3 | 1.88 | 8.05 |
| Superoxide Dismutase 1, Soluble (SOD-1) | ng/mL | 0.059 | 0.66 | pending | pending | 6.17 | 0.913 | 14.8 | 19.5 | 1.51 | 7.71 | 64.2 | 23 | 35.6 |
| Sortilin | ng/mL | 0.085 | 0.18 | pending | pending | 2.95 | 0.3 | 10.1 | 9.63 | 0.329 | 3.41 | 22.3 | 2.53 | 11.3 |
| Thyroxine-Binding Globulin (TBG) | ug/mL | 0.053 | 0.33 | 40 | 104 | 2.73 | 0.422 | 15.5 | 8.32 | 0.616 | 7.4 | 26.9 | 2.45 | 9.1 |
| Thymus-Expressed Chemokine (TECK) | ng/mL | 1.9 | 18 | pending | pending | 365 | 43.7 | 12 | 3708 | 1092 | 29.4 | NA | NA | NA |
| Testosterone, Total | ng/ml | 0.029 | 0.17 | 0.11 | 6.5 | 2.05 | 0.862 | 42.1 | 7.74 | 1.38 | 17.8 | 30.2 | 5.76 | 19.1 |
| Tissue Factor (TF) | ng/mL | 0.52 | 0.31 | <LOW> | 2.4 | 2.52 | 0.254 | 10.1 | 15 | 1.46 | 9.76 | 138 | 15.9 | 11.6 |
| Trefoil Factor 3 (TFF3) | ug/ml | 0.00010 | 0.00019 | 0.040 | 0.49 | 0.805 | 0.117 | 14.5 | 5.11 | 0.589 | 11.5 | 75.5 | 7.61 | 10.1 |

Table 1 Continued

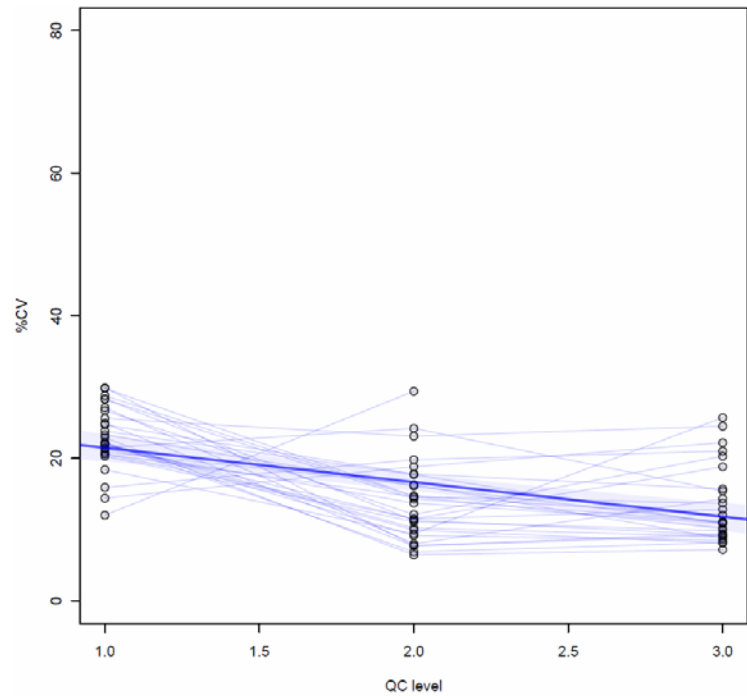
| Analyte | Units | Lower Assay Limit | Least Detectable Dose | RBM Low Plasma Range | RBM High Plasma Range | Level 1 qc | | | Level 2 qc | | | Level 3 qc | | |
|---|--------|-------------------|-----------------------|----------------------|-----------------------|------------|-------|------|------------|-------|------|------------|------|------|
| | | | | | | mean.1 | sd.1 | cv.1 | mean.2 | sd.2 | cv.2 | mean.3 | sd.3 | cv.3 |
| Transforming Growth Factor alpha (TGF-alpha) | pg/mL | 12 | 6.7 | pending | pending | 118 | 26.3 | 22.2 | 2211 | 152 | 6.85 | 9117 | 745 | 8.18 |
| Transforming Growth Factor beta-3 (TGF-beta-3) | pg/mL | 9.1 | 20 | pending | pending | 370 | 24.1 | 6.53 | 1274 | 120 | 9.42 | 4025 | 379 | 9.41 |
| Tamm-Horsfall Urinary Glycoprotein (THP) | ug/ml | 0.00015 | 0.00019 | 0.0084 | 0.052 | 4.51 | 3.49 | 77.3 | 47.3 | 7.13 | 15.1 | 153 | 34 | 22.2 |
| Thrombospondin-1 | ng/mL | 57 | 5.7 | pending | pending | 1017 | 154 | 15.2 | 3465 | 663 | 19.1 | 10341 | 1489 | 14.4 |
| Tissue Inhibitor of Metalloproteinases 1 (TIMP-1) | ng/mL | 0.28 | 4.1 | 59 | 192 | 9.96 | 1.04 | 10.4 | 21.7 | 1.71 | 7.87 | 52.7 | 3.22 | 6.1 |
| Thrombomodulin (TM) | ng/ml | 0.12 | 0.10 | pending | pending | 0.888 | 0.266 | 29.9 | 9.81 | 0.761 | 7.75 | 51.8 | 4.89 | 9.43 |
| Tenascin-C (TN-C) | ng/mL | 7.4 | 14 | pending | pending | 53.2 | 8.17 | 15.4 | 490 | 40 | 8.16 | 2363 | 173 | 7.32 |
| Tumor Necrosis Factor alpha (TNF-alpha) | pg/mL | 1.0 | 7.0 | <LOW> | 27 | 6.1 | 2.3 | 37.8 | 37.3 | 9.62 | 25.8 | 241 | 20.3 | 8.44 |
| Tumor Necrosis Factor beta (TNF-beta) | pg/mL | 1.5 | 18 | <LOW> | 120 | 74.8 | 11.9 | 15.9 | 286 | 26.4 | 9.25 | 1088 | 93.5 | 8.59 |
| Tumor Necrosis Factor Receptor-Like 2 (TNFR2) | ng/mL | 0.16 | 0.22 | 3.1 | 79 | 5.39 | 0.579 | 10.8 | 15.5 | 1.25 | 8.02 | 42.3 | 4.3 | 10.2 |
| Thrombopoietin | ng/mL | 0.11 | 4.4 | <LOW> | 6.2 | 5.25 | 0.534 | 10.2 | 17.6 | 1.12 | 6.37 | 56.8 | 4.09 | 7.2 |
| TNF-Related Apoptosis-Inducing Ligand Receptor 3 (TRAIL-R3) | ng/mL | 0.51 | 0.24 | pending | pending | 9.6 | 0.411 | 4.28 | 30.1 | 2.03 | 6.74 | 85.3 | 4.46 | 5.23 |
| Serotransferrin (Transferrin) | mg/dl | 9.5 | 93 | pending | pending | 671 | 54.1 | 8.07 | 2267 | 174 | 7.67 | 8009 | 780 | 9.74 |
| Thyroid-Stimulating Hormone (TSH) | uIU/mL | 0.017 | 0.021 | 0.18 | 3.7 | 0.464 | 0.049 | 10.6 | 2.53 | 0.209 | 8.26 | 15.8 | 1.26 | 8.01 |
| Transferrin (TTR) | mg/dl | 0.25 | 0.49 | pending | pending | 11 | 2.57 | 23.3 | 41.8 | 6.73 | 16.1 | 150 | 16.6 | 11 |
| Vascular Cell Adhesion Molecule-1 (VCAM-1) | ng/mL | 1.6 | 1.5 | 284 | 1310 | 33.4 | 4.21 | 12.6 | 104 | 10.5 | 10.1 | 299 | 26.8 | 8.97 |
| Vascular Endothelial Growth Factor (VEGF) | pg/mL | 7.8 | 17 | 91 | 1790 | 437 | 33.6 | 7.69 | 1793 | 123 | 6.86 | 8321 | 724 | 8.7 |
| Vitronectin | ug/ml | 6.7 | 4.0 | pending | pending | 433 | 95.1 | 22 | 745 | 74 | 9.93 | 1523 | 124 | 8.11 |
| Vitamin K-Dependent Protein S (VKDPS) | ug/ml | 0.12 | 0.39 | pending | pending | 4.88 | 0.367 | 7.52 | 14.3 | 1.29 | 9.01 | 37.1 | 2.76 | 7.46 |
| von Willebrand Factor (vWF) | ug/mL | 0.70 | 2.8 | 5.3 | 74 | 25.5 | 7.34 | 28.8 | 76.3 | 11.1 | 14.6 | 234 | 32.1 | 13.7 |

Analyte Quality Control (QC) results from the 2010 ADNI Plasma Analysis:

QCs were generated by spiking blank human plasma with extracts of cell cultures expressing the individual analytes. Low, medium and high QCs were run on each plate for almost all the analytes. The total ADNI plasma cohort was run on 15 plates. The QC results for each analyte are included in Table 1. QCs were performed in duplicate, but samples were run in singlicate. As a result the first QC result from each plate was used to derive the summary QC statistics for each analyte. It should be noted that analytes with one or more QC CV values above 25% should be treated with caution and are highlighted in yellow. Figure 1 highlights (A) the 28 analytes with QC CVs within the 20-30% range and (B) the 14 analytes with QC CVs >30%. In addition, analytes with numerous sample values close to or below LDD should be treated with caution.

A)

Endothelin-1 (ET-1)
 Transforming Growth Factor alpha (TGF-alpha)
 Cancer Antigen 19-9 (CA-19-9)
 Vitronectin
 Agouti-Related Protein (AGRP)
 Alpha-1-Antitrypsin (AAT)
 Platelet-Derived Growth Factor BB (PDGF-BB)
 Fibrinogen
 Alpha-Fetoprotein (AFP)
 Apolipoprotein(a) (Lp(a))
 Interleukin-3 (IL-3)
 Thrombomodulin (TM)
 Adiponectin
 Intercellular Adhesion Molecule 1 (ICAM-1)
 Transthyretin (TTR)
 Serum Glutamic Oxaloacetic Transaminase (SGOT)
 Myeloperoxidase (MPO)
 Glucagon
 Heparin-Binding EGF-Like Growth Factor (HB-EGF)
 Glutathione S-Transferase alpha (GST-alpha)
 Immunoglobulin M (IGM)
 Cortisol (Cortisol)
 von Willebrand Factor (vWF)
 Apolipoprotein A-II (Apo A-II)
 Interleukin-5 (IL-5)
 C-Reactive Protein (CRP)
 Thymus-Expressed Chemokine (TECK)
 Amphiregulin (AR)



B)

Matrix Metalloproteinase-9, total (MMP-9, total)
 Superoxide Dismutase 1, Soluble (SOD-1)
 Pregnancy-Associated Plasma Protein A (PAPP-A)
 Brain Natriuretic Peptide (BNP)
 Interleukin-15 (IL-15)
 Fibroblast Growth Factor basic (FGF-basic)
 Matrix Metalloproteinase-9 (MMP-9)
 Connective Tissue Growth Factor (CTGF)
 Progesterone
 Tumor Necrosis Factor alpha (TNF-alpha)
 Immunoglobulin E (IgE)
 Testosterone, Total
 Malondialdehyde-Modified Low-Density Lipoprotein (MDA-LDL)
 Tamm-Horsfall Urinary Glycoprotein (THP)

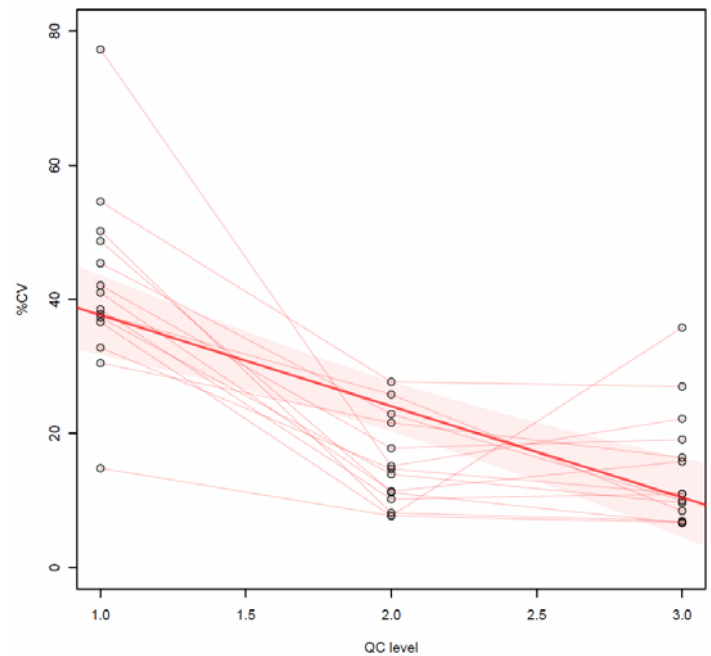


Figure 1: Summary of QC data for (A) the 28 analytes with QCs within the 20-30%CV range and (B) the 14 analytes with QCs >30%CV range.

Methodology:

One thousand sixty five 500ul EDTA plasma samples from the ADNI cohort were selected from a larger set of samples and shipped to RBM for analysis. Plasma samples were obtained in the morning following an overnight fast. For the majority of samples, the time from collection to freezing was within 120 minutes. Processing and aliquoting were per ADNI lab SOPs. The data reported on the ADNI website lists data for 1062 of the 1065 samples. One patient was excluded from the final analysis due to a screen failure. In addition, one patient had a redundant sample submitted for analysis. Only one set of data from that data point was reported. These 3 samples were excluded from further analysis and from the ADNI website. The final samples reported represented baseline and one year collections from 396 MCI, 112 AD and 58 control subjects. Samples were selected based upon availability of additional biomarker endpoints. For example, samples from AD subjects with associated CSF Abeta42 and/or Pittsburgh Compound B (PIB) one year data were included in the AD subset. Samples from normal subjects selected for inclusion were from subjects who had baseline Abeta 42 levels above the median baseline Abeta42 level for the normal cohort. Not all patients selected had a plasma sample available at the one year follow-up. Table 2 represents the demographics of the population selected. The selection process introduced a bias in ApoE4 prevalence in the control population, presumably due to association between ApoE4 levels and CSF Abeta42 levels (i.e. patients with low CSF Abeta42 levels were more likely to have one or more ApoE4 allele).

Table 2: Demographics of the plasma multiplex biomarker cohort

| | Control | MCI | AD |
|-----------------------|--------------|--------------|--------------|
| N baseline (12 m) | 58 (54) | 396 (345) | 112 (97) |
| Age | 75.3 (62-90) | 74.9 (55-90) | 75.4 (55-89) |
| Gender M/F (baseline) | 30/28 | 256/140 | 65/47 |
| ApoE4% (baseline) | 9% | 53% | 68% |
| MMSE (range) | 28.9 (25-30) | 27.0 (23-30) | 23.6 (20-27) |

It should be noted that age was calculated based upon date of birth and upon date of sample draw from baseline visit. Samples were randomized for processing at RBM and RBM was blind to the clinical information. A Statistical Analysis Plan (see Appendix II) was prepared prior to analysis.

It is recognized that the ADNI cohort is not optimized to identify novel diagnostic markers of disease for AD. As a result, a pilot study using plasma samples from the University of Pennsylvania cohort was completed to test the feasibility of the technology and to generate hypotheses that could be tested in the ADNI cohort. The preliminary findings are presented as hypothesis in the statistical analysis plan (Appendix II).

RBM made a comment on RBM sample numbers 191 (RID 750), 348 (RID 1182), 413 (RID 194), 435 (RID 221), 845 (RID 604) and 1041 (RID 1057). Analyte [Fibrinogen]: Result suggests sample type is serum. Caution is advised when interpreting result as analyte concentrations may vary between serum and plasma. An outlier rule was developed for the dataset and is described in the SAP.

What is posted on the ADNI Website and cautionary notes to data analysis:

There are two datasets posted on the ADNI website relating to the plasma multiplex pilot from the Biomarkers Consortium Project. The first dataset coded ADNI Plasma Multiplex Raw Data includes the original raw data from the run to be intended as reference. The second dataset title ADNI Plasma QC Multiplex data is the cleaned, quality controlled data according to methodology described in the statistical analysis plan. The third file is a detailed listing of what variables were changed in the QC process. It is recommended that raw data not be used to derive summary statistics as most of the analytes are not normally distributed and there are some analytes with quite a few LOW reported. Summary statistics should not be run on data that is not normally distributed. It is recommended that analytes with numerous LOW values listed or with majority of values listed between the LDD/LAL be treated with caution as deriving reliable results may be challenging. Consultancy with a trained statistician is highly recommended prior to reporting results based upon multiple comparisons and upon repeated measures (i.e. baseline and 12 months).

Table 3: Column header definitions in the ADNI Plasma Multiplex Imputation log

| Variable Name | Description and Coding |
|----------------------|--|
| ID | record ID |
| RID | ADNI subject ID |
| sampleID | Date of QC |
| Sample_Received_date | Date sample received at UPenn |
| Visit_Code | Visit Designator (bl = baseline; m12 = 1 year) |
| analyte | Name of Analyte with Units |
| LDD | Least Detectable Dose (see primer for details) |
| avalue | Recorded Value |
| analval | Numeric Value after possible imputation (see primer for details) |
| belowLDD | Is analval < LDD? Note: this flag pertains to both recorded value and imputed value (0=no ; 1=yes) |
| readLOW | Is recorded value <LOW> or numeric? (0=numeric; 1=<LOW> - see primer for details) |
| outlier | Is recorded value an outlier? (0=no; 1=yes) - outliers imputed to 5SD from mean |

Table 4. Column header definitions in the ADNI Plasma QC Multiplex data

| Variable Name | Description |
|----------------------|--|
| ID | record ID |
| RID | ADNI subject ID |
| sampleID | Sample ID from UPenn |
| Sample_Recieved_date | Date sample received at UPenn |
| Visit_Code | Visit Designator (bl = baseline; m12 = 1 year) |

References:

Ray S, Britschgi M, Herbert, C et al., (2007) Classification and prediction of clinical Alzheimer’s diagnosis based on plasma signaling proteins. Nat Med 13(11):1359-62

APPENDIX I

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APPENDIX II
Biomarkers Consortium Project
Use of Targeted Multiplex Proteomic Strategies to Identify
Plasma-Based Biomarkers in Alzheimer's Disease

Statistical Analysis Plan

Biomarkers Consortium Project
Use of Targeted Multiplex Proteomic Strategies to Identify
Plasma-Based Biomarkers in Alzheimer’s Disease

Statistical Analysis Plan

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1 Introduction

The Analysis Plan described within this document represents the work of the Biomarkers Consortium Project “**Use of Targeted Multiplex Proteomic Strategies to Identify Plasma-Based Biomarkers in Alzheimer’s Disease**”. This project was submitted to the Biomarkers Consortium Neuroscience Steering Committee by a subgroup of the Alzheimer’s Disease Neuroimaging Initiative (ADNI) Industry Scientific Advisory Board (ISAB) for execution and was managed by a Biomarkers Consortium Project Team that includes members from academia, government and the pharmaceutical industry; funding for this project was provided through an overage of funds raised by the Foundation for NIH for the ADNI partnership, as well as Pfizer Inc. This project is intended to be the first part of a multi-phased effort seeking to utilize samples collected by ADNI to qualify multiplex panels in both plasma and cerebrospinal fluid (CSF) to diagnose patients with Alzheimer’s Disease (AD) and monitor disease progression.

Biomarker tools for early diagnosis and disease progression in Alzheimer’s disease (AD) remain key issues in AD drug development. Identification and validation of cost-effective, non-invasive methods to identify early AD and to monitor treatment effects in mild-moderate AD patients could revolutionize current clinical trial practice. Treatment prior to the onset of dementia may also ensure intervention occurs before irreversible neuropathology.

The aim of the project is to determine the ability of a multiplex plasma based immunoassay panel to discriminate among disease states and to monitor disease progression over a one year period. The multiplex panel is based upon luminex immunoassay technology and a 151 analyte panel has been developed by Rules Based Medicine (RBM) to measure a range of inflammatory, metabolic, lipid and other disease relevant endpoints. Prior studies using an older version of the RBM panel (an 89 analyte version) suggested some analytes on the panel differed between AD and controls. The panel has been expanded to include analytes from a recent article describing plasma based biomarkers of AD.

The analyses described in this statistical analysis plan should be regarded as exploratory and meant for hypothesis and model generation, rather than for hypothesis confirmation and model validation. Findings will need to be confirmed and expanded upon in subsequent studies using other, independent data sets.

2 Study Design and Objectives

2.1 Study Design

Samples from baseline and one year ADNI plasma sample set will also be assessed (N=229 Controls, 192 AD, 398 for amnesic mild cognitive impairment (MCI)). Of the 398 MCI subjects, 153 subjects have progressed to dementia as of March 2010. This statistical analysis plan addresses the analysis of data from a subset of these samples.

Previously, a small pilot study was run using University of Pennsylvania banked plasma samples from AD (N~98) and control (N~72) subjects. Data from the pilot study have already been analyzed (**Hu et al., in preparation**). Assessing the utility of the expanded RBM panel will incorporate some of the findings from the pilot study.

2.2 Study Objectives

- To determine whether baseline levels or changes from baseline levels for individual analytes are associated with patient demographics (age, gender) or disease status.
- To determine whether baseline levels for a combination of analytes derived from either a biological pre-selection based method and/or from a statistically based/machine learning language approach will provide a panel with distinctly different profiles for the ADNI normal controls (NC), MCI or AD.
- To determine whether baseline levels for a combination of analytes derived from either a biological pre-selection based method and/or from a statistically based/machine learning approach will provide a panel that discriminates pre-demented subjects who will progress to dementia in one year and/or two years.
- To determine whether change from baseline levels for a combination of analytes (derived as above) predict cognitive decline in AD or correlate with disease progression.

3 Univariate Analysis

Univariate analyses will be performed first. The results of the univariate analyses may be used to inform and select analytes to be used in the pathway analyses and multivariate predictive model-building. Results from the univariate and multivariate sets of analyses will be compared for overlap and a final panel selected based on optimal overlap.

3.1 Classification Endpoints

Clinical diagnosis at time of enrollment/collection will be used to classify AD, MCI and control groups. Clinical diagnosis of amnesic MCI followed by diagnosis of AD will be used to classify pre-demented progressors.

3.2 Data Quality Control (QC)

Up to 190 analytes may be measured in the plasma updated RBM panel. Plasma data will be analyzed separately and compared for each analyte dependent upon sample availability. The data will be prepared for all analysis as follows:

- Review of the quality control samples data for each run to determine the variability characteristics of the spiked plasma (or serum) QC samples. Characteristics examined for the LOW, MEDIUM and HIGH QC samples for each biomarker will include mean, standard deviation (SD) and the percent coefficient of variation (%CV) for each analyte to determine not only the variability at each concentration but whether or not there is a major change in variability across the concentration range for each analyte.
- Analytes with more than 10% missing data will not be analyzed further. Missing data are generally indicated by “QNS” (quantity not sufficient for analysis) by RBM.
- Analytes with more than 10% recorded as “LOW” will not be included in the multivariate analysis. These analytes will be assessed to compare the proportion of measurable samples in each disease status category. If proportions differ substantially among disease status categories for some analytes, alternative approaches may be explored for incorporating such analytes in the multivariate analyses described below.
- For expression values preceded by a “<” or “>” modifier, the numeric portion of the value will be used for all subsequent data preparation and analyses.
- For each analyte, the distribution of measured values within each diagnostic group will be examined. If the distributions are not normal, the team will seek appropriate transformations (e.g., Box-Cox transformations (**Box and Cox, 1964**)) so the transformed markers approximate normality. All subsequent data preparation and analyses will then be conducted on the transformed values.
- Analytes with less than 10% missing/“LOW” values will have the non-numeric values imputed as follows:
 - Values recorded as “LOW” will be imputed to LLD/2
 - Missing values will be imputed to be the mean of the non-missing values for that analyte.
 - Samples with imputed values for more than 25% of the analytes will be excluded from the analysis
- Multidimensional scaling and/or Mahalanobis distances will be used to detect sample outliers and misclassified subjects.
- For univariate analysis, outliers that are more than 5 STD from mean will be assigned the value of the nearest non-outlier point. For outliers apparent in

multivariate reviews, outliers will be imputed using a nearest neighbor or other appropriate algorithm.

The imputation and outlier definition strategy defined above is only one of many possible strategies that could be used. If resources permit a limited number of alternative strategies may be used to assess the robustness of the analytical conclusions obtained using the strategy defined above.

As part of data QC, patient, visit, and sample identifiers will be checked for uniqueness and logical consistency. Graphical displays will be used to check for systematic patterns related to batch, run date, sample quality measures, and QC sample characteristics.

Cleaning, outlier detection, and distribution displays of all samples will be performed prior to merging phenotype data with the biomarker data. Misclassification assessment will be performed prior to statistical analysis.

3.3 General approach

Analysis of variance (ANOVA) and analysis of covariance (ANOCOVA) models will be used to compare mean analyte levels among groups of interest. These ANOVA/ANOCOVA models will include the diagnosis/disease status group and other covariates including age, gender and apoE4 genotype/status, as well as possible interactions among these factors. The interactive effect between group and other covariates will be tested. Depending on the outcome of these tests, the differences between groups will be tested either by the main effect of diagnosis or the effect of diagnosis at a fixed level of other covariates (i.e., apoE4 status) or through the adjusted least square means.

A major analytic concern in these tests is the control of overall type I error rate due to the relatively large number of CSF and plasma proteins tested in this aim. The team will address this concern using false discovery rate (FDR) methodology.

3.4 Hypotheses to Be Tested

The following univariate hypotheses will be addressed for each analyte:

HO1i: Analyte i is not associated with age [age treated as a continuous variable]

HO2i: Analyte i is not associated with gender

HO3i: Analyte i is not associated with ApoE status

HO4i: Analyte i is not associated with disease status or change in disease status (adjusted for age, gender, and/or ApoE status as necessary)

An initial set of analyses will look at whether the mean baseline level of each individual marker differs among disease groups (normal, MCI, AD) via an ANOVA or ANCOVA and t-test analysis. “Disease status” will be based on the clinical calls recorded in the ADNI database. Additional analyses may be conducted using disease status defined using one or more alternative definitions based on cognitive and/or functional tests.

False discovery (FDR) corrections will be applied to p-values and will be reported along with raw p-values. When adjusting for and evaluating the impact of multiple tests, a distinction will be made between:

- a set of proteins defined a priori as being of particular biological interest based on review of pathway annotations (see Table 1)
- the remaining assayed proteins.

FDR corrections based on the Benjamini-Hochberg method will be calculated separately for the two sets.

A second set of analyses will be performed using data only from MCI subjects. ANOVA/ANCOVAs similar to the above will be run to assess whether mean baseline levels of the analytes differ among MCI non-converters and converters.

A third set of analyses will be run to determine whether change from baseline analyte levels at one year are associated with change in disease status.

A fourth set of analyses will be run to determine whether any of the analytes correlate with significant changes in Clinical Dementia Rating Scale-Sum of the Boxes (CDR-SB) or Auditory-Verbal Learning Test (AVLT).

A fifth set of analyses will determine whether levels of any of the analytes are associated with low CSF abeta/high tau, high amyloid brain burden and significant brain atrophy.

Analyses to examine relationships between analyte levels and use of acetyl cholinesterase inhibitors or other medications by subjects may also be performed.

4 Pathway Analysis of Biomarkers

Although statistical machine learning-based approaches can generate a short list of discriminatory proteins, such analyses reveal little about biological relevance. In addition to machine learning approaches, the current proposal will use a systems biology approach to better understand pathway relationships between identified proteins. The Project Team will use pathway mining tools, such as those offered by Ingenuity and Pathway studio, to find the functional connections between the markers from plasma samples. This will provide direct evidence to support key hypotheses. To further increase the biological

relevance of the protein markers in the predictive models, biomarkers will be selected based on their presence in distinct biological pathways.

In addition empirical characterizations of marker data such as pair-wise correlations or higher-order relations (e.g. principal components analysis (PCA)) will be used. This analysis will derive an initial short list that will then be analyzed using multivariate and machine learning language approaches.

5 Multiple Marker Models

Multivariate statistical methods and multiple machine learning approaches will be used to identify the optimal combination of a group of proteins to predict disease status. The problem of classification and prediction has received a great deal of attention in mining “-omics” data. In the case of this project, the task will be to classify and predict the diagnostic category of a sample on the basis of protein quantitative profiles. The main type of statistical problem is the identification of “marker” genes that characterize the difference between diagnosis groups (e.g. AD, MCI) – the so called “variable/feature selection” problem. One challenge is to find the optimal combination of uncorrelated proteins. This factor not only is very important to improve prediction accuracy but also contributes to the merits of a good classifier: the simplicity and insight gained into the predictive structure of the data.

In all multivariate model building, feature selection will be done using data only from the training set. Feature selection based on a completely independent data set is not feasible for this project due to sample size and the fact that this is the first study to use this version of the RBM panel.

Multiple marker analysis will be used to build relationships to the disease groups. The candidate models include: logistic regression, linear discriminant analysis, nearest shrunken centroid, random forests, support vector machines and partial least squares. The technique of **Xiong et al. (2004)** may be applied to search for the linear combination of informative proteins that optimally discriminates between the diagnostic groups. Models generated by the various methods will be compared and the “best” model will be chosen based on model fit, robustness, and parsimony considerations.

Models will be fit with two sets of covariates, 1) assay results only and 2) assays results plus additional patient information including gender, age, and ApoE4 allele status. Other biomarkers such as amyloid PIB load, hippocampal atrophy, baseline mini-mental state examination (MMSE), and/or baseline Alzheimer's Disease Assessment Scale-Cognitive Subscale 11 (ADAS-cog 11) may also be used. For a specific model, differences in performance between models fit using the two classes of predictors variables should be characterized to understand the predictive ability of the assays beyond that of routine clinical information on the patients. If possible, formal inference should be made regarding the statistical significance of including the assay variables above and beyond that of the clinical data. Analysis will focus on the following:

- good characterizations of error rates; poor fitting models should not be interpreted.
- any feature selection routines should be extensively cross-validated (see **Ambroise and McLachlan, 2002**)
- measures of marker importance should be biased towards those that use uncertainty (e.g. logistic regression slope tests) as opposed to those that do not (e.g. random forest variable importance, etc).

The multivariate results will be compared to the single marker analysis and (especially) biological relevance.

In addition to the modeling efforts outlined above and described in detail in sections 5.1 and 5.2, an additional set of analyses will be based upon confirming a predictive model developed from the University of Pennsylvania pilot dataset. In brief, a predictive model using pilot data for the 24 analytes shown in Table 1 was identified as providing good discrimination between AD vs Control.

Table 1. Analytes in Predictive Model from UPenn Pilot Study

| Analyte | Analyte | Analyte |
|--------------------------------|--------------------------|-----------------------|
| Alpha-1 Microglobulin ↑ | FAS ↑ | Resistin ↑ |
| Angiotensin 2 ↑ | HGF ↑ | Stem Cell Factor ↑ |
| Apolipoprotein E ↓ | IGF-BP2 ↑ | Tenascin C ↑ |
| Beta2-Microglobulin ↑ | IL-10 ↑ | Thrombomodulin ↑ |
| B Lymphocyte Chemoattractant ↑ | NT-proBNP ↑ | TIMP-1 ↑ |
| Cortisol ↑ | Osteopontin ↑ | VCAM-1 ↑ |
| E-Selectin ↓ | Pancreatic Polypeptide ↑ | VEGF ↑ |
| FABP ↑ | PAPP-A ↑ | Von Willebrand Fact ↑ |

The model based on pilot data for these 24 analytes will be tested using the ADNI dataset for its predictive abilities to discriminate AD from Control, MCI from Control and MCI rapid progressors to dementia (within 2 years) vs slow or stable MCI. The algorithm for the pilot study prediction model will be provided by the model developers.

5.1 Analyte Filtering

Several approaches to filtering and feature selection may be examined. Results of the univariate analyses described above may be used to define a starting set of markers for the analysis. Results of the pathway analysis may also be used to define a starting set. In addition, pre-filtering of markers in an unsupervised fashion prior to building models based on empirical measures may also be applied.

5.2 Model Building Approach

For each type of model, predictive model building will be based on an iterative resampling approach.

For each of the K resampling iterations, the steps will include:

- Splitting the data into training and test sets
- Applying an unsupervised filter on the predictors based on data in the training set only.
- Building and tuning the predictive model on the current training set
- Predicting the current test set
- Calculating and saving the performance (classification accuracy, Kappa)
- End resampling iteration
- Assess performance of the model over the K sets of performance metrics

In the above algorithm, the resampling schemes can include cross-validation, the bootstrap and repeated training/test set splits (see Appendix). Methods for unsupervised feature selection can include filters on variance of individual predictors, high pair-wise predictor correlations, etc.

In addition to the iterative resampling approach above, an additional set of predictive models may be built based on the training-test 60-40 split data sets and 10-fold cross-validation set defined by the ADNI Biostatics Core. This set of models will permit easier comparisons with other modeling efforts performed using the ADNI Biostat core-defined approach.

6 Power Calculations

The sample size for this project and resulting analyses are based upon and limited by the availability of samples from both the pilot and ADNI samples. Additional post-hoc analysis will be completed based upon variability characteristics of the current study to understand power requirements for subsequent analysis of future datasets, in discussion with the Project Team.

7 References

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8 Appendix I

In order to make sure that there is balance among the pre-specified cross-validation groups, we present a table of the 10 partitions vs. the baseline-year1 status of ADNI subjects. No info from month 6 or other visits were used.

| | AD-AD | AD-MCI | AD-NA | MCI-AD | MCI-MCI | MCI-NA | MCI-NL | NL-MCI | NL-NA | NL-NL |
|---|-------|--------|-------|--------|---------|--------|--------|--------|-------|-------|
| a | 16 | 0 | 3 | 7 | 29 | 5 | 0 | 0 | 0 | 21 |
| b | 16 | 1 | 2 | 10 | 25 | 3 | 1 | 0 | 1 | 23 |
| c | 16 | 0 | 1 | 6 | 32 | 2 | 1 | 0 | 1 | 21 |
| d | 17 | 0 | 3 | 6 | 29 | 4 | 1 | 0 | 1 | 21 |
| e | 15 | 0 | 5 | 5 | 28 | 8 | 0 | 1 | 2 | 21 |
| f | 15 | 0 | 2 | 4 | 29 | 3 | 3 | 2 | 1 | 20 |
| g | 17 | 0 | 5 | 6 | 30 | 2 | 0 | 0 | 2 | 21 |
| h | 18 | 0 | 2 | 6 | 27 | 4 | 1 | 0 | 3 | 19 |
| i | 16 | 1 | 3 | 6 | 31 | 3 | 0 | 0 | 2 | 23 |
| j | 15 | 0 | 4 | 8 | 27 | 3 | 2 | 0 | 3 | 20 |

The following table restricts to the primary groups of interest.

| | AD-AD | MCI-AD | MCI-MCI | NL-NL |
|---|-------|--------|---------|-------|
| a | 16 | 7 | 29 | 21 |
| b | 16 | 10 | 25 | 23 |
| c | 16 | 6 | 32 | 21 |
| d | 17 | 6 | 29 | 21 |
| e | 15 | 5 | 28 | 21 |
| f | 15 | 4 | 29 | 20 |
| g | 17 | 6 | 30 | 21 |
| h | 18 | 6 | 27 | 19 |
| i | 16 | 6 | 31 | 23 |
| j | 15 | 8 | 27 | 20 |