

21 **Abstract**

22

23 Acoustic telemetry was combined with a project that uses sonar and drift gillnetting methods to 24 estimate Chinook salmon *Oncorhynchus tshawytscha* escapement in the Nushagak River, 25 Alaska. The sonar project uses dual-frequency identification sonars (DIDSONs) to count passing 26 fish and drift gillnetting to apportion sonar estimates to species. These estimates are indices 27 because the river's width (~ 300 m) and uneven bottom topography allow for only a third of the 28 river to be sampled. This range is enough to fully enumerate sockeye salmon *O. nerka*, the 29 dominate species, but not Chinook salmon, which are known to migrate beyond the sampling 30 range. Acoustic telemetry was used to determine what proportion of Chinook salmon traveled 31 within the sampling range of the sonar project. We inserted acoustic tags into Chinook salmon 32 ~13 km downriver and deployed an array of acoustic receivers at the sonar site to track tagged 33 fish. From 2011 to 2014, 799 Chinook salmon were tagged. The tagged fish used the entire river 34 width while migrating through the acoustic array exhibiting a wide variety of behaviors that 35 included moving straight through the array, making multiple up and down trips, holding, and 36 crossing over from one side of the river to the other. On average, 57% of tagged fish traveled 37 through regions sampled by the sonar with annual percentages of 65% (2011), 54% (2012), 64% 38 (2013), and of 47% (2014). These proportions were used to expand the sonar-derived indices to 39 in-river abundance estimates.

40

41 **Keywords**

43 acoustic telemetry, Chinook salmon, dual-frequency identification sonar (DIDSON), drift 44 gillnetting, migration behavior

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46 **1. Introduction**

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48 Acoustic telemetry has been widely used to track movements of juvenile and adult fish in a 49 variety of environments including lakes (Hayden et al., 2014), estuaries (Childs et al., 2008), 50 oceans (Chittenden et al., 2009; Starr et al., 2005), and rivers (Heublein et al., 2009; Mathes et 51 al., 2010; McMichael et al., 2010). Dual-frequency identification sonar (DIDSON; Belcher et al., 52 2002) and ARIS (Adaptive Resolution Imaging Sonar; i.e., the DIDSON replacement) have been 53 successfully evaluated for assessing passage rates of migrating adult sockeye salmon 54 *Oncorhynchus nerka* (Holmes et al., 2006; Maxwell and Gove, 2007) and other fish species (Egg 55 et al., 2018). DIDSON or ARIS imaging sonars are widely used to assess fish escapement (Buck, 56 2013; El Mejjati et al., 2010; English et al., 2016; Maxwell et al., 2011; Miller et al., 2013; Pipal 57 et al., 2012), as well as fish composition and species-specific movement patterns (Crossman et 58 al., 2011; Grote et al., 2014). If multiple species co-migrate and are similar in size, then 59 estimating salmon abundance using DIDSON requires a method to apportion the sonar counts to 60 species. The Alaska Department of Fish and Game (ADF&G) operates a sonar project ~50 km 61 upriver from the mouth of the Nushagak River to estimate escapement into the watershed of 62 sockeye, chum *O. keta*, and Chinook salmon *O. tshawytscha*. The project combines sonar 63 (DIDSON) to estimate fish passage and drift gillnetting methods (test fishing) to apportion the 64 sonar estimates to species (Buck et al., 2012; Buck, 2013). These methods are satisfactory for 65 estimating chum and sockeye salmon. However, Chinook salmon escapement estimates are

66 considered indices and not abundance estimates because this species is known to migrate beyond 67 the sampled regions (Miller, 2000). The stability of the indices had not been assessed. This is a 68 concern because commercial and sport fishery management plans based on this Chinook salmon 69 index have been in place since 1992 (Nushagak-Mulchatna King Salmon Management Plan 5 70 AAC 06.361).

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72 Mark-recapture studies have been used to ground-truth salmon estimates from sonar projects 73 (Mora et al., 2015; Rakowitz et al., 2009; Rawding and Liermann, 2011; Reimer and Fleischman, 74 2016). Although mark-recapture studies provide an abundance estimate for comparison, it does 75 not provide specific information on where salmon are traveling in the river and what changes 76 might be made to the sonar project to improve it. Acoustic telemetry had the potential to provide 77 this additional information. However, large, shallow rivers are a difficult environment for 78 acoustics (Faulkner and Maxwell, 2015). Surface and boundary layers interfere with signal 79 propagation and cause multi-pathing of the signal, and uneven bottom topography produces 80 acoustic shadow zones. We first tested an acoustic telemetry system in the Kenai River, Alaska, 81 a smaller river on the road system that is easier to access, to determine how well the system 82 would work in a riverine environment. A single hydrophone was deployed on one bank and then 83 moved to the opposite bank. Acoustic tags were placed at stationary positions for a period of 84 time and then pulled alongside a boat through the test region. The acoustic tags were detectable 85 across the river and up and downriver as far as 200 m in each direction. These results convinced 86 us to proceed with the study at the Nushagak River.

109 a variance has been estimated (Reynolds et al., 2007). Three gillnet mesh sizes (20.6 cm (8.125 110 in), 29 meshes deep; 15.2 cm (6.0 in), 45 meshes deep; and 13.0 cm (5.125 in), 45 meshes deep) 111 each 18.3 m long (10 fathoms) were drifted through regions directly below each sonar strata. 112 Buoys marked the end range of each stratum. Daily estimates of Chinook salmon were obtained 113 by counting all fish images in DIDSON 10-min/h files, expanding the counts to an estimate of 114 daily passage by counting strata, and apportioning the strata estimates using proportional catch 115 per unit effort (CPUE) of each species in each stratum. A DIDSON Chinook salmon count as it 116 is used in this paper is a simplification that refers to a count obtained from the expanded and then 117 apportioned sonar estimates.

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119 2.2. Tag insertion

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121 Acoustic tags were inserted into Chinook salmon at a site that was presumed to be far enough 122 upriver from the mouth (37 km) to avoid tagging fish whose ultimate destination might not be 123 the Nushagak River and far enough downriver of the sonar site (13 km) to allow fish to 124 normalize their swimming behavior before reaching the detection zone. At the insertion site 125 (Figure 1), the river flows through one unobstructed main channel (Tag Insertion Site 1) with a 126 small side channel (Tag Insertion Site 2). The sites experienced tidal fluctuations of \sim 2 m. 127 128 A bathymetry map of the 2 channels was produced in 2011 to determine the depths across the 129 river and whether the site was adequate for drifting gillnets. A Simrad EK60 echo sounder with a 130 4° 200 kHz single-beam transducer (ping rate 5 pings/s, pulse duration 0.128 ms, power 250 W)

131 was used to obtain depth data. The unit was pole-mounted to a boat with the transducer placed

132 ~0.25 m below the water surface. A Trimble DSM212H Global Positioning System (GPS) unit

133 provided positioning information at a rate of 10 Hz with differential corrections received

134 from the U.S. Coast Guard Differential GPS station in Kodiak. Hypack version 2011 was used to 135 follow survey lines, bottom-track the acoustic data, and correct depth information for the vertical 136 mounting offset and changes in water surface elevation that occurred over the course of the 137 survey. Water surface elevation was read from a staff gauge at half-hour intervals during the 138 surveys. Depth values were referenced to the water surface elevation at the beginning of each 139 survey. Bathymetry data were processed using ESRI ArcGIS version 10.2 with the Spatial 140 Analyst extension for raster-based spatial analyses and the Geostatistical Analyst extension for 141 the interpolation of the bathymetric data using kriging without anisotropy. Map data were 142 projected in WGS 1984 UTM Zone 4N coordinates. The resulting geostatistical surface was 143 converted to a raster. The final map was generated from a triangulated irregular network built 144 from 0.05 m contour lines extracted from the raster.

145

146 The main channel was a long, straight stretch ~275 m wide. The bathymetry showed the river 147 bottom dropping off smoothly from both shores with no significant debris obstructions or sand 148 bars (Figure 2). The shallow side channel was deepest at the downriver end, had a low flow rate, 149 and was inaccessible by boat during low tide. Although the side channel did not appear to be a 150 significant migratory route, it was included as a drift station so that all Chinook salmon 151 migrating the Nushagak River would be available for capture.

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153 A tagging schedule was implemented to tag fish in proportion to their abundance across a 6- 154 week period based on historical run timing. Fish were captured using drift gillnets with effort 155 concentrated around the high tides. Three fishing zones were established at Site 1 (zone1: right-156 bank, zone 2: mid-channel, and zone 3: left-bank). Zone 4 was established at Site 2 close to

157 where it rejoined the main channel upriver. Two gillnet mesh sizes were used (20.6 cm (8.125 158 in), 29 meshes deep and 15.2 cm (6.0 in), 45 meshes deep), both 18.3 m long (10 fathoms). 159 These nets were identical to the two nets that account for the overwhelming majority of Chinook 160 captured for apportionment at the sonar site. The nets were mono twist filament webbing dyed a 161 translucent green, identical to the nets used at the sonar site for apportionment (Buck, 2013). The 162 primary difference between the netting at the upper and lower river sites was that the upper river 163 test fishing included a smaller mesh size (13.0 cm) geared for sockeye salmon that was omitted 164 from the acoustic tag study.

165

166 To minimize stress on fish, nets were pulled in as soon as a fish was detected, limiting each haul 167 to 1 or 2 fish. The short drift time reduced the amount of time a fish had to become tangled and 168 reduced the stress of capture. A live tank held the captured salmon prior to tagging. The tank was 169 emptied and refilled multiple times daily to freshen the water. Once the net was pulled in, the 1 170 or 2 fish were processed in approximately 1.5–2 min and released. Captured Chinook salmon in 171 poor shape were released without a tag. The biggest factor in determining 'poor shape' was fish 172 energy. Lethargic fish or those with damaging hook or net marks were deemed poor. Few fish, 173 an estimated 2%, were rated as poor.

174

175 Lotek, Inc. model MM-TP 16–25 MAP acoustic tags were inserted into the gullet of Chinook 176 salmon using a long plastic tube (Figure 3). Fish length from mid-eye to tail fork (MEF) was 177 measured, a scale sample was collected, and the tag identification number (ID) was recorded. The 178 acoustic tags were 16 mm (diameter) x 58 mm (length), weighed 27 g in air, and transmitted a 76 179 kHz pulse every 2 s continuously. We discussed the possibility of producing tags that would emit 180 a sound pulse at one of the DIDSON frequencies. However, a pulse emitted from a tagged fish 181 would not be synchronized with the DIDSON's listening range. The DIDSON determines the 182 range of a returned signal, i.e., an echo, based on the time it takes the sound pulse to travel to the 183 end of the range setting and return to the DIDSON receiver. A transmitted pulse from a tagged 184 fish, if detected, would not represent the actual range of the fish relative to DIDSON. In addition, 185 active pings are easier to detect than echoes and may be detected whether the tagged fish 186 physically passed through the DIDSON beam or not.

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188 2.3. Tag detection

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190 An array of receivers was installed at the sonar site to detect the acoustic tags. The sonar project 191 and acoustic tag array were operated concurrently. The site is 50 km upriver from the mouth of 192 the Nushagak River and 4 km downriver from the village of Portage Creek. Here, the river is 300 193 m wide and flows within a single channel. While the river height fluctuates by ~0.4 m due to 194 tides, no flow reversal occurs. Water level typically drops across the summer as snow melt 195 declines, although temporary surges follow periods of excessive rain.

196

197 Bathymetry maps were produced in 2011 and 2012 at the sonar site to provide information on the 198 best placement for the acoustic array and determine whether changes in the bottom topography 199 occurred between years. In 2011, the bathymetry equipment described for the lower river tag 200 insertion sites was used. In 2012, we used the vertical beam of a Sontek River Surveyor M9 201 Acoustic Doppler Current Profiler (ADCP; dual 4-beam 3.0 MHz/1.0 MHz) to collect depth data 202 and an Ashtech Mobile Mapper 100 GPS with GLONASS (Global Navigation Satellite System)

203 to georeference the data. Survey lines were spaced 15 m apart. The relative difference in depth 204 between the 2011 and 2012 surveys was calculated with ArcGIS Spatial Analyst Extension (map 205 algebra with cell size 10 m x 10 m). The mode of the depth differences between the two surveys 206 was used to empirically reference the 2012 data to the reference elevation determined for 2011.

207

208 At the array site, the river channel is characterized by a gradual slope along the right bank, 209 steeper slope along the left bank, relatively flat center, and submerged mid-river sandbar (Figure 210 4). Profiles extracted from the 2011 bathymetry data show smooth sloping shores along both 211 sides of the river where the sonars are deployed. The 2012 bathymetry map (Figure 5) was 212 similar to the 2011 map with differences caused by a buildup of substrate along the right shore 213 and erosion along the left shore (Figure 6).

214

215 Prior to the first year of the study, a series of feasibility tests were done at the Nushagak River 216 sonar site to determine whether the acoustic tags would be detectable across the river. The tests 217 included static tests where a tag was moored at a surveyed location within the array footprint, 218 tow tests where a series of 3 tags at different depths were suspended from a boat or buoy and 219 drifted downriver through the array, accuracy tests where position estimates of the towed tag 220 were compared with GPS tracks, and cross-channel detection tests where two boats were located 221 on opposite shores and receivers were mounted closer to shore and farther from shore to 222 determine the best locations. For tests 1–3, 4 Lotek WHS 3050 wireless acoustic data-logging 223 receivers (DLs) were deployed, two along each side of the river 200 m apart. Beacon tags (MM-224 16-50, high power, 76 kHz) were attached to three of the DLs. Beacons were similar to fish tags 225 except they had a larger battery, slower burst rate (30 s intervals), and no pressure or temperature 226 sensors. The beacon tags were used to synchronize the array and determine whether each DL 227 could detect the other beacons. Detection ranges up to 600 m were observed, with 300 m more 228 typical. Detection ranges varied with weather conditions and other factors including the 229 orientation of the tag and DL. Lotek, Inc. produced a report with descriptions and results of each 230 test, which is included in its entirety in Maxwell et al. (2019). Following deployment each year 231 of the study, the array was retested for blind spots by drifting test tags through the array at 232 various depths. The DL's were moved as needed to improve detection.

233

234 Based on the feasibility tests, Lotek, Inc. recommended deploying 6 DLs. In 2011, 6 DLs were 235 deployed, 7 in 2012 and 2013, and 8 in 2014. A beacon tag was attached to each DL, with an 236 additional beacon attached to a buoy placed at the end range of the right-bank sonar. It was 237 thought that this offshore beacon might be in a better position to be detected by the DLs and 238 improve the synchronization of the array. In 2011–2013 arrays consisted of two lines of tripods 239 close to each shore with no mid-river deployment due to heavy boat traffic in this region. In 240 2014, an additional DL was deployed on the mid-river sandbar. We felt this posed minimal risk 241 to boat traffic while potentially improving tag detection. The DLs were attached to tripods with 242 the transducers pointed down to place them at deeper depths. The tripods were carried out from 243 the bank at low tide and set in water ~1.5 m deep (i.e., maximum chest wader depth). The 244 latitude and longitude of each tripod was recorded with the Ashtech Mobile Mapper. We ran a 245 cable to shore from one DL on each bank to download data without having to move the tripods. 246 These were periodically checked to assess tagging mortality in-season and to determine when the 247 DLs could be pulled at the seasons' end without missing tags. In 2011, all DLs were pulled from 248 the river multiple times during the field season to check batteries and ensure the hard drives did

249 not overfill. After the first year, we learned that the battery life and hard drive space of the DLs 250 were sufficient for the entire field season, so downloads were reduced to the cabled DLs. Not 251 moving the DLs in-season improved our ability to process the tag data. Eventually, we cabled all 252 DLs to avoid moving them in and out of the water in season. 253 254 255 2.4. Tag processing 256 257 Tag data were processed using software packages from Lotek, Inc. To determine which tagged 258 fish had been detected, WHS Reader version 2.1 was used to convert the tag data to text files. 259 The text files were condensed by tag ID and DL to produce the date and time of the first and last 260 detection and number of rows of data (i.e., the number of detections). Two positioning software 261 programs were used to produce position estimates for the tagged fish, Asynchronous Logger 262 Positioning Software (ALPS) and U-Map. Data processing from 2011–2013 used different 263 versions of ALPS software updated each year by Lotek, Inc. to fix bugs that were encountered. 264 The ALPS vers. 2.2, 2.3, and 2.4 required inputting Universal Transverse Mercator (UTM)

266 of losing fish at this early stage to filtering, we set the dilution of precision (DOP) filtering

265 coordinates for the DLs, sound speed in water, and filtering parameters. To reduce the possibility

267 parameter to 20 (twice the default) and the conditioning number (CN) and H-R (a reliability

268 number which attempts to quantify the reliability of a position estimate) to default values. These

269 filtering parameters were not input by the user in U-Map. Sound speed was calculated from the

270 river temperature based on Simmonds and MacLennan (2005).

271

272 In the first year of the study, a synchronization problem in the ALPS algorithm caused the 273 program to stop processing following a break in the array, which caused many detected fish to 274 be eliminated from the output. According to Lotek, Inc., the DLs needed time to synchronize, so 275 running longer periods of data through the ALPS program should have been a better approach. 276 Lotek, Inc.'s programmers were unable to tell us how long a time period was needed for the 277 synchronization of the beacons to occur. This was problematic in 2011 when we pulled the DLs 278 from the water multiple times during the field season to ensure the DL's storage was not filling 279 up, but unfortunately this caused numerous breaks in the dataflow. We reprocessed the data 280 through ALPS selecting a variety of time periods ranging from multiple weeks to a few days and 281 determined that a period of one week or less resulted in the output of many previously omitted 282 tags. To determine whether fish were omitted from the data processing or not detected, we 283 compared the number of unique tag IDs output from ALPS with the output from the WHS 284 Reader program and reprocessed data for periods when numerous detections of specific tags 285 occurred in the WHS Reader output, but no position estimates were produced in ALPS. This 286 resulted in fewer missed tags. Since we were able to obtain the missed tags using this method, 287 this data was not reprocessed in later years when new software became available. 288 Although beacons were attached to every DL, only a single beacon was needed to synchronize 289 the array. For each tag ID, we processed data using one beacon at a time and selected the output 290 from the beacon that produced the most complete and coherent track. The offshore beacon 291 attached to the buoy was never selected as the 'best' beacon to synchronize the array. This was 292 likely due to the beacon's movement, which swirled around due to current flow. Beacons 293 attached to the DLs were more stable. In 2014, Lotek, Inc.'s new program, U-Map vers. 1.2.2. 294 was used to process the tag data. U-Map fixed the synchronization problem and automatically

295 selected the best beacon for each tag, so it was not necessary to reprocess data using each 296 beacon. U-Map required inputting the DL's UTM coordinates, river temperature, and salinity. 297 Sound speed was automatically calculated. U-Map did not include user-configurable filtering 298 parameters. The files containing the position estimates were concatenated using script files 299 written in TIBCO Spotfire SPLUS (version 8.1).

300

301 Each fish track (unique tag ID) was plotted and viewed sequentially during the filtering process. 302 Preliminary filtering was done using SPLUS. We first eliminated obvious errors such as position 303 estimates that were well beyond the boundaries of the array or incorrectly placed on land. For 304 tracks with excessive scattering or multipathing, we removed points with low DOP, CN, or H-R 305 values. Multipathing, which occurs when a ping emitted from an acoustic tag follows an indirect 306 path to the DL, was sometimes observed as double or even triple pathways with the track 307 appearing to jump back and forth between parallel locations. For some tracks, further restricting 308 the filtering parameters reduced the number of extraneous positions making the direct path more 309 obvious. Secondary filtering was done using ESRI ArcMap version 10.2. A more precise point-310 to-point filter was applied to remove points outside of the dominant track. To help identify and 311 delete obvious outliers, points were converted to lines with each line connecting consecutive 312 detections from an individual fish. Fish that generated no coherent track or a track <100 m long 313 were removed from the dataset. In the final step, fish tracks were smoothed. Edited points with 5 314 consecutive records spanning <2.5 min were smoothed with a 7-point running average of *x* and *y*. 315 The smoothed coordinates were plotted, reviewed, and edited to remove any additional missed 316 outliers.

318 2.5. Accuracy of position estimates

319

320 Many environmental factors affect whether a ping is detected and the detection quality or 321 accuracy of a position estimate. Acoustic signals may bounce off structure in the river such as 322 weirs, sonar mounts, boats, and other fish, which can cause multipathing or even loss of 323 detection. Uneven bottom topography and surface and boundary layers may also interfere with 324 signal propagation. Anything that alters the direct path from the signal to the detector will cause 325 error in a position estimate. To examine this error, we processed position estimates from known, 326 stationary targets—the beacons. Using U-Map, processing the beacon data was like processing 327 fish tags except that it was necessary to remove the DL paired with the beacon being analyzed 328 from the list of DLs in the array. Beacon position estimates were plotted, and a simple spatial 329 filter was applied to remove position estimates outside the of array and obvious outliers. A 330 bootstrap procedure was used that randomly selected 400 points without replacement from a 331 single beacon's dataset. The percentage of points within 5 and 10 m from the GPS-measured 332 beacon location was determined, and the process was repeated 1,000 times. A standard deviation 333 was calculated from the bootstrapped data. This process was repeated for each beacon. Heat plots 334 were made by randomly selecting 2,000 position estimates from one beacon's dataset without 335 replacement, rounding the northing and easting coordinates to the nearest 1 m, and plotting a 336 frequency matrix. For a combined plot of data from all beacons, we randomly selected 20,000 337 position estimates from the database and plotted the frequency matrix with an overlay of the 338 shoreline, sonar beams, and beacon coordinates.

339

340 2.6. Fish depth

342 Fish depths (*FD*) were superimposed as a point layer onto bathymetry maps using the 2011 map 343 for the 2011 fish depths and the 2012 map for all remaining years' data. Since we expect that 344 most Chinook salmon travel along the river bottom when moving upriver to reduce energy loss 345 due to current flow (Hinch and Rand, 2000), aligning fish depth with the bathymetry provided 346 another means of assessing the accuracy of fish positions. The depth of each fish position was 347 obtained using pressure output from the acoustic tags (*pT*) according to an equation supplied by 348 Lotek, Inc.,

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340\,
$$

349
$$
FD = \frac{0.3 \times pT}{1.43}.
$$
 (1)

350 The pressure range of the acoustic tags was 15 psi divided into 50 steps for a conversion of 15/50 351 = 0.3 psi/step. The 1.43 conversion is based on water density (ρ) at 5° Celsius, the gravitational 352 constant (*g*), and the conversion factor (*C*) of pressure in psi to Pascals; i.e., *ρg/C*.

353

354 2.7. Tag fish proportions

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16 356 To generate a potential DIDSON count (*p*), a tagged fish had to pass through the footprint of a 357 DIDSON beam regardless if it passed while the DIDSON was recording. An algorithm written in 358 SPLUS and/or a visual assessment was used to determine whether a given track passed inside or 359 outside of a DIDSON beam footprint. The layout of the sonar beams and acoustic receivers is 360 shown in Figure 7A. Uncertainty in the position estimates and detection issues made 361 classification more difficult. To handle the uncertainty, we drew three probability regions around 362 each beam footprint (Figure 7B). In the cross-river dimension, the first probability region 363 extended from shore to 5 m short of the DIDSON end range. Tagged fish that traveled through

364 this region were assigned a *p* of 1 or -1 depending on their direction of movement and a bank 365 assignment; i.e., left bank (*LB*) or right bank (*RB*), for fish passing through one of the DIDSON 366 beam footprints or *Os* if it passed outside of either footprint. For example, a tagged fish traveling 367 upriver through this first region along the right bank was assigned [*0 LB, 1 RB, 0 Os*] for a 368 potential DIDSON count of 1. The second region extended from the end of the first region to 5 m 369 offshore of the DIDSON end range for a *p* assignment of 0.5. The assignment for a fish traveling 370 along the left bank through this region would be [*0.5 LB, 0 RB, 0.5 Os*], an equal chance of 371 passing inside or outside of the beam footprint. The third region (*p*=0.25) extended from the end 372 of the second region to 5 m farther offshore. Fish passing through the second or third regions 373 were classified as edge fish.

374

375 Some fish traveled through both DIDSON beams during a single upriver trip. For a given fish 376 that traveled through the first region of the LB beam footprint, crossed the river, and then 377 traveled through the first region of the RB beam footprint, a *p* value of 2 would be assigned 378 because the DIDSON would have counted 2 fish. If a fish traveled through the first region along 379 LB and then went through the third region along RB, the assignment [1 LB, 0.25 RB, 0 OS] 380 would result in a *p* value of 1.25.

381

382 Truncated fish tracks (short tracks) occurred when a track moving along the shoreline ended 383 prior to reaching a beam footprint. Short tracks were likely the result of environmental 384 conditions interfering with detection. For these tracks there were multiple possibilities: the fish 385 may have continued through the sonar beam, moved offshore, reversed direction and headed 386 downriver, or was captured by a fisherman. To handle short tracks, the probability regions were

387 extended 10-m in the upriver-downriver dimension (Figure 7). A fish was assigned a zero 388 probability of going through a sonar beam unless the fish entered a probability region before 389 detection was lost. A fish traveling along the left bank that entered the outermost probability 390 region before detection was lost was given an assignment of [*0.25 LB, 0 RB, 0.75 Os*].

391

392 Many tagged fish made a single, upriver trip through the array (ST fish), but several made 393 multiple up and downriver trips. Fish tracks with ≥ 1 h between 2 successive observations were 394 divided into multiple tracks and classified as multiple-trip (MT) fish. For these fish, each trip 395 was assessed in the same manner as the ST fish except that downriver trips yielded negative 396 potential counts and assignments for multiple trips were summed. For example, a fish observed 397 traveling upriver through the LB beam footprint and then downriver through the RB beam 398 footprint would be assigned a *p* value of 0 [1 LB, -1 RB, 0 OS]. If a fish traveled upriver through 399 the LB beam footprint, downriver through the RB beam footprint, and then back upriver in the 400 middle of the river, the *p* value would be 1 [1 LB, -1 RB, 1 OS]. A special case was presented by 401 implied trips. Tagged fish first detected moving downriver through the array were assumed to 402 have traveled upriver unobserved or the track was rejected by filters. Also, a fish that made two 403 consecutive upriver trips had to have made an unobserved (implied) downriver trip. We assumed 404 these fish had an equal probability of traveling through the *RB*, *LB*, or *Os* so implied upriver trips 405 were assigned [*0.33 LB, 0.33 RB, 0.33 Os*] and implied downriver trips were assigned [-*0.33 LB,* 406 *-0.33 RB, -0.33 Os*]. Fish trips were classified as upriver, downriver, both if the fish moved 407 upriver and downriver within a single trip, or undetermined. Whether implied or observed, 408 multiple assignments for a given fish were summed to produce a single *p* value per fish.

409 Probability assignments for each fish were put into Assignment Tables by year (Maxwell et al., 410 2019).

411

412 To obtain an overall proportion of fish that traveled through a DIDSON beam footprint, we first 413 tallied the right \hat{R}_i and left \hat{L}_i bank potential counts for each tagged fish *i* by year *y* from the 414 assignment tables,

415
$$
\hat{R}_y = \sum_{i=1}^{n_y} (\hat{R}_i)
$$
 and $\hat{L}_y = \sum_{i=1}^{n_y} (\hat{L}_i)$ (2-3)

416 where *n* is the number of tagged, filtered fish. Next, we calculated yearly proportions \hat{P}_y from 417 summed right and left bank potential DIDSON counts, yearly variances *Var* (\hat{P}_y) , and a total (all 418 vears) variance $Var(\hat{P})$:

$$
\hat{P}_y = \frac{\left(\hat{R}_y + \hat{L}_y\right)}{n_y},\tag{4}
$$

420
$$
Var(a_y) = \frac{a_y \bullet (1 - a_y)}{(n_y - 1)}
$$
, and (5)

421
$$
Var(a) = \frac{1}{16} \sum_{y=1}^{4} Var(a_y)
$$
 (6)

422 where $a = \hat{P}$. Yearly proportions were averaged to obtain a mean proportion (\hat{P}_m) .

423

424 2.8. In-river abundance estimates

426 In-river abundance estimates \hat{A}_y were obtained by expanding apportioned Chinook salmon

427 escapement indices from the sonar project \hat{S}_y for each year using \hat{P}_y ,

$$
\hat{A}_y = \frac{\hat{S}_y}{\hat{P}_y} \tag{7}
$$

429 with variances $Var(\hat{A}_y)$,

430
$$
Var(I_y) \approx b_y^2 Var(\frac{1}{a_y}) + \left(\frac{1}{a_y}\right)^2 Var(b_y) - Var(\frac{1}{a_y})Var(b_y)
$$
 (8)

431

where $I = \hat{A}$, $a = \hat{P}$, and $b = \hat{S}$. The $Var(\frac{1}{\hat{P}})$ *ay* 432 where $I = \overline{A}$, $a = \overline{P}$, and $b = \overline{S}$. The $Var(\overline{A})$ was approximated using the Delta method

433 (Seber, 1982),

434
$$
Var(\frac{1}{a_y}) \approx a_y^{-4}Var(a_y).
$$
 (9)

435 The $Var(\hat{S}_y)$ obtained from the sonar project incorporates variance in the sonar estimates and

436 test-fishing catch per unit effort (CPUE). The total sonar variance $Var(\hat{S})$ was calculated using

437 equation (5) with $a = \hat{S}$. For the total variance *Var* (\hat{A}), the total sonar variance *Var*(\hat{S}) and

438 mean values of *a* and *b* were used in equations (8) and (9).

439

440 Yearly and total standard errors (*SE*) and coefficients of variation (*CV*) were calculated from the 441 variances:

443
$$
SE_y = \sqrt{Var(\hat{I}_y)}
$$
 and $CV_y = \frac{SE_y}{\hat{I}_y}$. (10-11)

445 2.8. Fish length analyses

446

447 We analyzed length data to determine whether a bias occurred in fish lengths. A length bias was 448 possible between tagged fish detected nearshore versus mid-river; i.e., we might expect that 449 larger fish would travel mid-river while smaller fish would travel closer to shore. A second 450 potential bias examined was between tagged Chinook salmon captured in the nearshore zones 451 versus Chinook salmon captured at the upriver site. A potential bias between the two projects 452 would stem from either site differences or differences between netting operations. At the tagging 453 site, the small mesh net (13.0 cm) was omitted because it typically tangles rather than gills 454 Chinook salmon. Tangled salmon often fall out of the net as it is pulled in. Another difference 455 between netting operations was the length of the drift. Nets were pulled at the tagging site as 456 soon as a fish was detected, while at the sonar site, nets were pulled after timed 2.5-min drifts 457 were completed. Since both projects recorded MEF fish lengths, a length comparison was 458 possible.

459

460 Length frequency distributions from the sonar test-fishing and tagging projects were plotted as 461 density plots and compared using Kolmogorov-Smirnov Goodness-of-Fit Tests (K-S tests). Two 462 hypotheses were tested: 1) length frequencies from tagged fish that passed within the sonar 463 footprint (inside fish) were similar to length frequencies from tagged fish that passed outside the 464 sonar footprint (outside fish); and 2) length frequencies from the inside tagged fish were similar 465 to length frequencies from fish captured at the upriver test-fishing site (sonar fish). For this

- 487 1. Merged tagged fish from the assignment tables with their corresponding fish lengths.
- 488 2. Separated tagged fish into small and large fish datasets.

23 489 3. Tallied potential right \hat{R}_i and left \hat{L}_i bank counts for each fish *i* by year *y* from the small 490 fish dataset *s* using equations (2–3) where $\hat{R} = \hat{R}_s$, $\hat{L} = \hat{L}_s$, and $n = n_s$. 491 4. Summed the right and left bank potential DIDSON counts by year and calculated the 492 small fish proportion using equation (4) where $\hat{P} = \hat{P}_s$, $\hat{R} = \hat{R}_s$, $\hat{L} = \hat{L}_s$, and $n = n_s$. 493 Yearly variances $Var(\hat{P}_{sy})$ and a total variance $Var(\hat{P}_{s})$ were calculated using equations 494 (5) and (6) where $a = \hat{P}_s$ and $n = n_s$. 495 5. Repeated steps 3 and 4 using the large fish dataset *l* to obtain a large fish proportion \hat{P}_b 496 where $\hat{P} = \hat{P}_l$, $\hat{R} = \hat{R}_l$, $\hat{L} = \hat{L}_l$, and $n = n_l$. Calculated yearly variances *Var* (\hat{P}_k) and a 497 total variance $Var(\hat{P}_l)$ using equations (5) and (6) where $a = \hat{P}_l$ and $n = n_l$. 498 For the sonar data, we: 499 6. Extracted Chinook salmon lengths from the sonar's mixed-species Age-Sex-Length 500 (ASL) database. 501 7. Calculated the proportions of small $S\hat{P}_{sy}$ and large $S\hat{P}_{ly}$ Chinook salmon by year, 502 *y sy sy Sn Sn* $S\hat{P}_{\rm sv} = \frac{3n_{\rm sy}}{5}$ and *y ly ly Sn* 503 $S\hat{P}_{sv} = \frac{Sn_{sv}}{S}$ and $S\hat{P}_{lv} = \frac{Sn_{lv}}{S}$ (12–13) 504 505 where *Snsy* is the number of small Chinook salmon in the ASL database, *Snly* is the 506 humber of large Chinook salmon, and Sn_y is the total number. Yearly variances $Var(S\hat{P}_{sy})$ 507 and *Var* (\hat{SP}_{l}) and total variances *Var* (\hat{SP}_{s}) and *Var* (\hat{SP}_{l}) were calculated using

508 equations (5) and (6) where $a = S\hat{P}_s$ and $n = S n_s$ for small fish and $a = S\hat{P}_l$ and $n = S n_l$ 509 for large fish.

510 8. Apportioned yearly sonar estimates \hat{S}_y into small \hat{S}_s and large \hat{S}_b Chinook salmon,

511
$$
\hat{S}_{sy} = (S\hat{P}_{sy})(\hat{S}_{y})
$$
 and $\hat{S}_{ly} = (S\hat{P}_{ly})(\hat{S}_{y})$ (14-

512 15)

513

514 with variances
$$
Var(\hat{S}_y \hat{S}_{sy})
$$
 and $Var(\hat{S}_y \hat{S}_{ty})$,

515

516
$$
Var(b_yc_y) \approx c_y^2 Var(b_y) + b_y^2 Var(c_y) - Var(b_y)Var(c_y)
$$
 (16)

517

518 where $b = \hat{S}$, and $c = \hat{S}_s$ for small fish and \hat{S}_t for large fish. For the total variances *Var* (

519 \hat{S} \hat{S}_s and *Var* (\hat{S} \hat{S}_t), *b* is the mean \hat{S} , and *c* is the mean \hat{S}_s for small fish and mean

- 520 \hat{S}_l for large fish in equation (16).
- 521 Combining the tag and sonar data, we:

522 9. Estimated the in-river abundance for small \hat{A}_{s_y} and large \hat{A}_{t_y} Chinook salmon each year 523 by,

524
$$
\hat{L}_{xy} = \frac{\hat{S}_{sy}}{\hat{P}_{gy}}
$$
 and $\hat{L}_{ty} = \frac{\hat{S}_{ly}}{\hat{P}_{gy}}$. (17-18)

526 Variances $Var(\hat{A}_{xy})$ and $Var(\hat{A}_{ty})$ were calculated using equations (8) and (9) where *I*

$$
527 = L\hat{A}_s, \ a = \hat{P}_s, \text{ and } b = \hat{S}_s \text{ for small fish, and } I = L\hat{A}_l, \ a = \hat{P}_l \text{ and } b = \hat{S}_l \text{ for large fish.}
$$

528 For total variances $Var(L\hat{A}_s)$ and $Var(L\hat{A}_l)$, $Var(\hat{S})$ and mean values of *a* and *b* were 529 used in equations (8) and (9).

- 530 10. Summed \hat{A}_{sy} and \hat{A}_{rb} to obtain \hat{A}_{y} , the length-stratified estimates. Summed the 531 variances from the small *Var* (\hat{LA}_{sy}) and large *Var* (\hat{LA}_{ly}) fish estimates to obtain the 532 yearly *Var* ($L\hat{A}_y$) and total *Var* ($L\hat{A}$) variances for all sizes of fish.
- 533 11. Determined the length-stratified proportion $L\hat{P}_y$ for the combined small and large fish 534 by,

$$
L\hat{P}_y = \frac{\hat{S}_y}{L\hat{A}_y},\tag{19}
$$

536 and calculated yearly *Var* ($L\hat{P}_y$) and total variances ($L\hat{P}$) using equations (5) and (6)

537 where $a = L\hat{P}$ and $n = n$.

538 Ideally, the sonar estimates would be reapportioned daily; however, apportioning daily estimates 539 by length and species was not possible because zone information (i.e., right-bank nearshore, 540 right-bank offshore, left-bank nearshore, and left-bank offshore) is not part of the ASL database. 541 Instead, annual sonar estimates \hat{S}_y apportioned into length categories \hat{S}_y and \hat{S}_y served as a 542 reasonable proxy for the reapportioned daily estimates.

543

544 2.10. Bank ratios

546 We examined the bank orientation of Chinook salmon for both acoustic tagged fish and fish 547 captured at the sonar site. To compare these two datasets, we removed tagged fish that swam 548 outside of the beam footprints, comparing the remaining right *Rratioy* and left *Lratioy* bank ratios 549 from the tagged fish,

550
$$
Rratio_y = \frac{\hat{R}_y}{(\hat{R}_y + \hat{L}_y)} \text{ and } Lratio_y = \frac{\hat{L}_y}{(\hat{R}_y + \hat{L}_y)},
$$
 (20–21)

551 with bank ratios from the sonar project.

552

553 2.11. Climate and water data

554

555 Climate and water data were collected across all study years. In 2011, climate data included daily 556 precipitation and twice daily (0800 and 2000 hours) measurements of wind speed, direction, and 557 air temperature from Meteorological Terminal Aviation Routine Weather Report (METAR data) 558 stations located at airports in Dillingham (46 km northwest of the sonar site) and New Stuyahok 559 (60 km north of the sonar site). From 2012 to 2014, these same data were obtained using a Davis 560 Vantage Vue wireless weather station at the sonar site. Water temperature was recorded using a 561 HOBO Model UA-001-08 data logger attached to the right-bank DIDSON mount with settings 562 of 1-h (2011), 2-h (2013), and 5-min (2012 and 2014) increments. Light penetration of the water 563 column was measured using a HOBO Model UA-002-08 attached to the left-bank DIDSON 564 mount in 2011, 2013, and 2014. In 2012, the unit wasn't functioning so water clarity was 565 recorded based on a visual assessment of the water's color.

569 3.1. Tags inserted

568

590 3.3. Quality of fish tracks and position uncertainty

591

592 The quality of fish tracks varied widely. Many tracks formed a single smooth line through the 593 array (Figure 8A), some tracks included numerous detections, but it was unclear where the actual 594 track was headed (8B), others produced too few detections to make a track (Figure 8C), and 595 some resembled a random collection of points (Figure 8D). Of the tracks shown in Figure 8, all 596 were discarded except for 8A. Many tracks required a second level of filtering to produce a 597 smooth track (Figure 9). 598 599 We evaluated position uncertainty using data from the 2014 stationary beacons. Of the minimally 600 filtered position, estimates a mean of 70.6% of all beacons were within 5 m of their GPS 601 locations and 90.8% within 10 m; for the nonfiltered position estimates, 65.0% were within 5 m 602 and 84.1% within 10 m (Table 2). Beacon position estimates were centered on or near their GPS 603 locations, noted by small triangles in Figure 10. Position uncertainty was higher for the LB 604 beacons, with a mean of $67.9 \pm 2.2\%$ of position estimates within 5 m from the actual beacon 605 locations compared to RB beacons, with a mean of $81.3 \pm 1.8\%$. For position estimates within 10 606 m from the actual beacon location, the LB mean was $89.9 \pm 1.5\%$, the RB $96.4 \pm 0.9\%$. The most 607 tightly clustered position estimates for LB were around beacon 31, the middle beacon, with the 608 widest spread around beacon 36, the most downriver one (Figure 11). For the RB, the beacon 609 with the lowest uncertainty was beacon 34; the highest was beacon 35. 610 611 3.4. Classification of fish tracks

613 Most tagged fish went straight though the array one time (83.4%), and this percentage was 614 relatively constant between years (Table 3). These fish were classified as single-trip fish. Fish 615 that traveled through the array more than once were classified as multi-trip fish. We also 616 classified tagged fish based on whether they moved through the right or left bank beam footprint 617 (Figure 12A), through the middle of the river (Figure 12B), or produced short (Figure 12C), edge 618 (Figure 12D), 2-bank (Figure 13) or implied tracks. Short tracks averaged 6.0% across study 619 years with more observed along the left bank (4.7%) than the right bank (1.3%). Edge tracks 620 averaged 16.0%. Fish that traveled through both beam footprints made up 5.9–12.1% of the total. 621 The edge fish, fish that travel through both beams, and multi-trip fish are problematic for 622 DIDSON counting by creating more uncertainty in the counts. Plots of all fish tracks are 623 included in Maxwell et al. (2019).

624

625 The most common scenario for MT fish was either three trips/fish (upriver, downriver, and 626 upriver) or two trips/fish (upriver, downriver), but a few fish made as many as 5, 6, or 7 trips 627 through the array. The percentage of implied upriver trips was highest in 2011, likely due to the 628 frequent downloading of the DLs that interrupted the ALPS synchronization process and caused 629 more missed or incomplete tracks. The fewest implied upriver trips occurred in 2014, the year 630 the mid-river DL was deployed. Only 1 implied downriver trip was observed and that occurred in 631 2013.

632

633 Multi-trip fish often presented unexpected behavioral patterns as they traveled back and forth 634 through the array. Fish 332 (Figure 14), a 3-trip fish, made a large loop in trip 1 that spanned 635 most of the river, moved toward left bank, and exited the array upriver. The fish returned 4 d

636 later traveling down the middle of the river (trip 2), and again 2 d later moving upriver along the 637 left bank. The track stopped short of the left-bank probability region (trip 3). Assignments for the 638 3 trips were summed for a final assignment of [*0 LB, 0 RB, 1 Os*].

639

640 Fish 416 (Figure 15), a 2-trip fish, traveled upriver along the left bank (trip 1), then returned 7 d 641 later traveling beyond the left-bank beam footprint where it held for a period of time before 642 crossing the river and looping between right bank (an edge fish) and the river's center, spending 643 close to 6 h in the array (trip 2). The final assignment for this fish was [*1 LB, 0.5 RB, 0.5 Os*]. 644 This fish showed a typical example of holding behavior, first spending a considerable amount of 645 time offshore of the right bank and then offshore of the left bank before moving on (Figure 15B). 646

647 3.5. Fish depth

648

649 Tagged fish swimming upriver migrated near the river bottom most of the time, as shown by the 650 close alignment of fish depth with the bathymetry map (Figure 16). Fish depths were nearly 651 identical between years with only occasional tracks observed out of alignment with the river 652 bottom. In 2012, the example shown, one upriver track ran mid-river at a near-surface depth. 653 Downriver-moving fish, which were relatively rare, tended to swim near the surface except when 654 traveling through the deepest portion of the river where they traveled closer to the bottom, likely 655 due to the propensity of fish to hold in the deeper portions of the river (Figure 17). Note, tracks 656 from these downriver fish showed active movement with the fish often moving upriver, cross-657 river, or downriver, but the net result of the movement was downriver, indicating these fish were 658 alive.

660 3.6. Fish length distributions

662 Mean lengths for the Sonar ASL fish were 44.2, 57.8, and 80.1 cm for length categories 1 (<50 663 cm), $2 \le 50$ cm & ≤ 66 cm), and 3 (≥ 66 cm), respectively (Table 4). Length category 1 contained 664 52 Sonar ASL fish, and only 1 Tag inside and 1 Tag outside fish. The mean lengths in categories 665 2 and 3 were similar between the Sonar ASL and tagged fish. For length category 2, the mean 666 Tag inside, Tag outside, and all Tagged fish lengths were 58.7, 59.4, and 58.8 cm, respectively, 667 for category 3, mean lengths were 78.3, 79.7, and 79.2 cm, respectively. 668 669 Length frequency curves from acoustic tagged fish and sonar fish were mostly bimodal with 670 peaks at 60 and 80 cm that represent the peak efficiency of the two gillnets (Figures 18 and 19). 671 Comparing sonar fish with tagged fish that traveled inside the sonar footprint (Figure 18), the 60- 672 cm peaks from the tagged inside fish were smaller every year except 2012, with 60-cm and 80- 673 cm peaks from the sonar fish more similar to each other in 2011, 2013, and 2014. For tagged fish 674 that traveled outside the sonar footprint, the second peak was slightly larger each year (Figure 675 19). The most notable difference between the sonar fish and tagged fish was the lack of the 60- 676 cm peak in 2014. The K-S tests showed that length frequency distributions for the inside versus 677 outside tagged fish were significantly different in 2014 but not in the other years, while the sonar 678 versus inside fish lengths were significantly different in all years except 2013 (Table 5). The only 679 year where length frequency distributions were significantly different for both tests was 2014, 680 suggesting that length-stratified in-river abundance estimates would better represent the true 681 population for that year.

- 683 3.7. Tag proportions and in-river abundance estimates
- 684

685 On average, 32% of tagged Chinook salmon passed through the RB DIDSON beam footprint, 686 24% through the LB footprint, and 44% outside of either footprint during the four study years 687 (Table 6). Due to the length bias observed in 2014 and because we wanted consistency in the 688 data processing between years, we calculated length-stratified data for all years to compare with 689 non-stratified data (Table 6). Length-stratified abundance estimates (\hat{LA}_y) were lower than non-690 stratified estimates in 2011, 2013, and 2014, and higher in 2012. The across years' average 691 length-stratified abundance estimate of Chinook salmon (204,512) was lower than the non-692 stratified abundance estimate (209,264) by 4,752 fish, a percent difference of 0.57. The largest 693 difference between the two methods occurred in 2014 when the percent difference between them 694 was 2.47. Annual proportions from the length-stratified method (i.e., dividing the apportioned 695 sonar estimate by the length-stratified estimate) ranged from 0.47 to 0.65, averaging 0.57, an 696 across years' average that is similar to the non-stratified proportion of 0.56 (Table 6). 697 698 Bank ratios of the percentage of tagged fish that traveled through the DIDSON beam (inside 699 fish) were mostly the reverse of sonar bank ratios (Table 7). On average, more than half (57%) of 700 the inside tagged fish passed through the RB footprint while the sonar RB ratio averaged 35%. 701 The tagged fish RB ratios of 54–59% were more consistent between years, while the sonar ratios 702 were more dynamic (24–41%). 703

704 3.8 Climate and water data

728 Much of the river is divided into multiple channels. Apart from a small, shallow slough that runs 729 behind the sonar site, the site is considered a single channel relative to salmon passage. Chinook 730 salmon assessment was an add-on to an existing sockeye salmon project and subsequently, 731 Chinook salmon estimates became part of fishery management plans. Although the site is not

732 ideal, the Chinook salmon passage estimates obtained were better than no information.

733

734 Miller (2000) estimated that only 18% of Chinook salmon traveled within the sampling range of 735 the sonar used at the time, a Bendix echo-counting sonar (Gaudet, 1990). Although the DIDSON 736 covers more of the river's width, two-thirds of the ~300-m width is not sampled. In a comparison 737 study of the two sonars, Maxwell et al. (2011) found that fish counts in the nearshore strata were 738 similar between the Bendix counter and DIDSON, but offshore counts, where most Chinook 739 salmon are captured, were highly variable, suggesting Chinook salmon shift their migration 740 toward and away from shore—moving in and out of what was the sampling range of the Bendix 741 counter. The acoustic telemetry showed that, on average, 57% of tagged salmon migrated 742 through regions sampled by DIDSON during the study years, a much higher percentage than 743 Miller's 18%. This suggests that much of the shifting movement occurred within and not beyond 744 the larger sampling range of the DIDSON. From 2011 to 2014, the acoustic telemetry study 745 found that percentages of fish moving through ensonified regions were 65, 54, 64, and 47, 746 respectively. These percentages show that a relatively stable proportion of Chinook salmon 747 passage is ensonified and apportioned each year, making the indices of abundance used by 748 fishery managers reasonable, unlike the indices produced by the Bendix counter prior to the 749 transition to DIDSON.

751 4.2 Effects of Chinook salmon behavior on sonar estimates

752

753 This study provided a wealth of information on Chinook salmon behavior within the acoustic 754 array that highlighted some of the limitations of the sonar/test fishing system. The most obvious 755 shortfall of the sonar is the inability to ensonify the entire river. Like Miller's (2000) gillnet 756 study, the acoustic telemetry showed that tagged Chinook salmon used the entire river width as 757 they traveled through the array, whereas sonar and test fishing covered approximately a third of 758 the river. Expanding the indices of abundance to full-river estimates required knowing the 759 proportion of Chinook salmon that traveled through each sonar beam. This knowledge was 760 confounded by uncertainty in the position estimates for the 16% of fish tracks classified as 761 'edge' fish (Table 3). Although we attempted to identify tagged fish in DIDSON images based 762 on time, the average combined return of largely overlapping Chinook, chum, coho, and sockeye 763 runs to this river total 1.3 million fish within a few months. This results in multiple fish passing 764 through the beam at one time, which made it impossible to determine if a given DIDSON image 765 was from a tagged fish. Measuring fish image lengths might appear to be a good method to 766 narrow the search since the actual lengths of the tagged fish were known. Burwen et al. (2010) 767 measured DIDSON image lengths of tethered Chinook and sockeye salmon and showed that they 768 were similar to live fish measurements. Based on this research, they were able to enumerate and 769 separate large Chinook from sockeye salmon using DIDSON fish lengths (Burwen et al., 2011). 770 Measuring image lengths was not an option for us due to equipment limitations. For the offshore 771 strata, the DIDSON's low frequency setting is needed to achieve the desired sampling range. At 772 low frequency, the DIDSON transmits half the number of beams (48). The number of pixels 773 from these beams does not provide enough data to obtain a reasonable and repeatable length

774 measure unless a high-resolution lens is used. This lens reduces the composite beam to one-half 775 the field of view compressing the size of the 48 beams to more closely match the individual 776 beam widths of the 96 beams. At close range, where many sockeye salmon migrate, this 777 narrowed beam is smaller than the length of a sockeye salmon which makes measurement 778 impossible and makes it more difficult to count fish in large schools. Burwen et al. (2011) were 779 able to use a high-resolution lens because they weren't interested in estimating sockeye salmon. 780 At the Nushagak River, sockeye salmon are the primary species of interest for managing the 781 commercial fishery so adding a high-resolution lens would only be an option if we had been able 782 to deploy side-by-side DIDSONs along each bank, one with a high resolution lens for the 783 offshore strata and one with a standard lens for the nearshore strata. Instead, to account for the 784 uncertainty in 'edge' fish, we set up probability regions around the beam edge (Figure 7) and 785 assigned a probability of detection by DIDSON to each fish (Maxwell et al., 2019).

786

787 One assumption of salmon migration behavior is that fish conserve energy by traveling where the 788 flow is less, staying within shallower regions and remaining close to the river bottom (Hinch and 789 Rand, 2000). Chinook salmon do not fit this assumption. Many tracks did not make sense 790 energetically as fish traveled upriver through the deepest, higher flow regions. Most fish did, 791 however, take advantage of the resistance between flow and the river bottom. Comparing fish 792 depth with bathymetry (Figure 16) showed that most upriver-moving fish swam near the river 793 bottom. Chinook salmon are the largest salmon species that migrate the Nushagak River. Their 794 size and musculature allow them to swim against stronger current. Hughes (2004) explored a 795 hypothesis that larger salmon may experience wave drag from the river's surface when traveling 796 in shallower water and may benefit energetically from traveling farther offshore. Wave drag may
797 explain why Chinook salmon don't migrate close to shore, but it doesn't explain why they move 798 so far offshore. A potential reason for moving farther offshore into higher flow regions may be 799 congestion. As smaller salmon species (sockeye and chum) arrive in large numbers, the high 800 density of fish may push Chinook salmon farther offshore. This topic needs further exploration 801 and could be resolved from existing sonar and acoustic tag datasets.

802

803 When it was decided to use the sonar estimates of Chinook salmon to manage fisheries, little was 804 known about their behavior at this site other than the limited range coverage. Additional 805 problems for the sonar and test fish sampling include cross-over behaviors, multiple trips through 806 the array, and milling. Fish traveling back and forth across the river have the potential of being 807 double counted by the sonar and captured in test nets on both sides of the river. Tagged fish were 808 not bank oriented as they traveled from the tag insertion site to the detection array 13 km upriver. 809 The probability of a fish captured in Zone 1 and then entering the array in same zone purely by 810 chance would be 33.3%. Our results showed a percentage only slightly higher than chance 811 (33.7%), which indicates migrating Chinook salmon cross the river at least once or potentially 812 multiple times as they travel between the two sites. The 68 fish that stayed true to a zone may 813 have traveled within the same zone or may have made multiple crossings to end up in the same 814 zone. We frequently observed tagged fish crossing the river within the array, some traveling 815 through both sonar beams. For example, Fish 433 traveled through the right-bank beam footprint, 816 crossed the river and then traveled through the edge region of the left-bank beam footprint 817 (Figure 13). Although Fish 332 trip 1 (Figure 14A) and Fish 416 trip 2 (Figure 15B) also traveled 818 cross-river, both passed through only 1 beam footprint. Fish that traveled through both beam 819 footprints were assigned a probability of capture greater than one to account for potential double

820 counting. The percentage of fish that passed through both beam footprints while within the array 821 (5.9–12.1%) suggest that cross-over behavior is common in this species and may involve

822 multiple cross-over points, behaviors that are not possible to assess with DIDSON.

823

824 Although sonar operators count upriver fish and subtract downriver fish, if some of the trips 825 occur outside the ensonified area, this biases the sonar estimate. On average, 16.6% of tagged 826 fish were classified as multi-trip fish (Table 3). The proportions from this study account for these 827 multiple trips and the expansion to a full-river estimate reduces this bias. For some of the multi-828 trip fish, their last observed trip was downriver. This occurred in an average of 6.5% of fish 829 (Table 3). Anecdotal evidence suggests that a small portion of Chinook salmon may spawn 830 downriver of the sonar site. The test-fish crew at the sonar site has reported catching blushed and 831 spawned out Chinook salmon in early August during years the project was extended to assess 832 coho salmon, but this occurred after the time frame covered in this study. Another possibility is 833 that spawned-out salmon may have drifted downriver. We know of no reports of sport fishermen 834 catching spawning Chinook salmon below the sonar site during June and July. These fish may 835 have gone back downriver to spawn, returned upriver to spawn after the project ended, not been 836 detected on their final upriver trip, or died without spawning. We have no evidence to suggest 837 which possibility is the most probable.

838

839 We observed several instances of milling fish similar to Fish 416 trip 2 (Figure 15B) (Maxwell et 840 al., 2019). Whether fish are crossing back and forth, milling, or making multiple trips through the 841 array, the more time they spend within the sonar sampling region, the more likely they are to be 842 counted more than once. This information is not available from sonar images, nor would it be

843 available from a simple mark-recapture project which would only tell us if the sonar estimates 844 were biased high or low, not why. An acoustic tag study similar to this one is beneficial when 845 setting up a new site or adding a new species with potentially different behaviors to an existing 846 project. The acoustic tag information provided a means to better understand the bias in our 847 estimates due to these behaviors and correct for it.

848

849 4.3 Bias in bank ratios

850

851 A bias was found in the bank ratios between the sonar and acoustic tag projects. The sonar 852 project estimated more Chinook salmon along LB, 59–76%, with more variability between years, 853 while the acoustic tag study estimated fewer Chinook salmon along LB, 41–46%, with less 854 variability (Table 7). We explored three potential explanations for this reversal in the bank ratios 855 between the projects. First, netting zones that are not well matched to the sonars' nearshore and 856 offshore sampling regions would bias the species apportionment. Large numbers of sockeye and 857 chum salmon migrate close to shore resulting in nearshore fish counts that comprise 87–96% of 858 the RB count and 77–92% of the LB count (Maxwell et al., 2011). Because of the higher number 859 of nearshore fish counted, a Chinook salmon inappropriately classified as a nearshore fish would 860 substantially increase the estimate of that species for that day. Buoys mark the dividing line 861 between strata and the end point of the counting range on both banks. They are placed at the start 862 points of the drifts but do not define the entire drift corridor. Although it is possible that current 863 could push the boat and nets closer to one bank than the other, we have no reason to believe that 864 this is more likely on one bank than the other.

888 Comparing Chinook salmon lengths from the sonar test fishing program (sonar fish) and tagged 889 fish that traveled through a sonar beam footprint (inside fish) showed that sonar fish lengths 890 contained a larger proportion of small Chinook salmon, suggesting that netting at the tagging site 891 was biased toward larger fish (Figure 18). This bias is not important if the distribution of fish 892 across the river at the sonar site is not segregated by size (Figure 19). We found that hypothesis 893 to be true in every year except 2014 (Table 5). Because of the bias in the 2014 data, we stratified 894 estimates for each year by length. If larger fish are more likely to migrate outside of the sonar 895 beam footprints and if the ratio of tagged small fish is less than the true proportion, the mid-river 896 population would be over-represented. Correcting for the length bias resulted in small 897 differences between the original proportion (0.42) and length-stratified proportion (0.47) for 898 2014 (Table 6). As expected from the K-S tests, differences between original and length-899 stratified proportions for 2011—2013, the years that there was no significant bias, were even 900 smaller (-0.01 to 0.02).

901

902 We have no evidence to explain why the 2014 tagged fish lengths were more biased against 903 small fish, but in that year, changes were made to the tag-insertion portion of the project. A 904 mark-recapture study utilizing pit tags was initiated and the same fishing crew inserted both 905 acoustic tags and a much larger number of pit tags. We originally thought the crew might have 906 inadvertently saved the expensive acoustic tags for the more fit fish; i.e., larger fish, but length 907 frequency curves for all fish captured at the lower site showed a lack of small fish with curves 908 very similar to the acoustic tag fish curves (Figure 21; from Maxwell et al. *In press*). 909 Environmental conditions were different in 2014. The river's mean temperature rose steadily 910 from 2011 to 2014 (Table 8). Water level appeared to be unusually low in 2014 based on

911 comments from fishermen and barge operators on the river. Low water may affect the

912 catchability of fish in drift nets, but whether these changes affected our ability to catch smaller

913 Chinook salmon at the lower site is unknown.

914

915 4.5. Accuracy of position estimates

916

917 Ehrenberg and Steig (2003) demonstrated that an acoustic array with receivers placed 15 m apart 918 in X, Y, and Z directions produced more accurate position estimates for a stationary tag than an 919 array with X and Y spaced 15 m with the Z direction only 3–4 m apart. Error estimates from both 920 placements were small, 0.2–1.8 m. In our study, the acoustic tags had pressure sensors to 921 measure depth, leaving estimation error in the X and Y dimensions. Our DLs were spaced 922 approximately 50 m apart in the upriver-downriver dimension and close to 300 m apart in the 923 cross-river dimension, which should have yielded more accurate position estimates, yet 924 uncertainty in the stationary tag position estimates (Figure 10) and the poor quality of some of 925 the fish tracks (Figure 8C and 8D) raised questions regarding accuracy. Our study was performed 926 over a much longer time period allowing for interference from many factors not observed in the 927 19-min sample from Ehrenberg and Steig (2003). Interference and multi-pathing from passing 928 boats, stationary objects like weirs, sonar mounts, and parked boats likely contributed to position 929 error in the fish tracks and stationary tags. From the stationary beacon analyses, we found that 930 70.6% of the minimally filtered position estimates from all beacons were within a 5-m radius of 931 the beacons' actual locations and 90.8% were within 10 m (Table 2, Figure 10). Watching the 932 data plot point by point made it apparent which position estimates were the result of 933 multipathing. The plotting resembled a bull's eye pattern as points plotted around a central point

934 with very narrow scatter but occasionally 'jumped' to a distant region. Distal points accumulated 935 as the number of detections grew creating the smear of points observed around the beacons' GPS 936 location. Position estimates for the fish tracks were similar with some containing jumped points, 937 but their time in the array was short compared to the stationary beacons, so there were far fewer 938 distal points. Distal points were readily apparent in the fish tracks and were removed during the 939 filtering process.

940

941 Additional evidence for the accuracy of the position estimates came from the depth of tagged 942 fish heading upriver (Figure 16). The depths of most tagged fish traveling upriver through the 943 array aligned well with the river bottom bathymetry, indicating fish were traveling near the river 944 bottom and their positions estimates were aligned correctly with their depths. The alignment of 945 these very different methods provided additional evidence for the reliability and accuracy of the 946 fish tracks.

947

948 Environmental interference caused errors in position estimates that resulted in occasional random 949 patterns in fish tracks, outliers, missing segments (shorts) or entire trips through the array 950 (implied trips). Kessel et al. (2013) describes the influence of numerous environmental 951 parameters on detection rates including physical and chemical properties, surface conditions such 952 as wind and wave action, water depth and tides, bathymetry and obstructions. Humston et al. 953 (2005) selected deployment sites protected from wind to improve their detection range, which 954 varied from 230 m to 750 m. These detection ranges were similar to detection ranges in our 955 initial testing (300 m to 600 m; Maxwell et al., 2019). Gjelland and Hedger (2013) found that 956 detection rate was strongly dependent on wind speed, dropping off dramatically as wind speed

957 increased. Wind mixes air bubbles into the water, which dissipates transmitted sound and may 958 prevent detection. Gjelland and Hedger (2013) found that wind can drive air bubbles deeper into 959 the water than rain, noting that during an hour of wind, sound reflections from entrained air 960 reached 4–5 m deep. Their models confirmed that rain had less effect on detection rate than 961 wind. At the Nushagak site, rain was less of a factor than wind, with mean rainfalls of less than 962 0.3 cm, while wind speeds reached maximum speeds great than 30 km/h; i.e., 8.3 m/s, (Table 8). 963 These wind speeds are higher than Gjelland and Hedger's maximum observed speed of 6 m/s 964 which reduced their detection rate to 0. The strong winds at the Nushagak River undoubtedly had 965 a large effect on detection rates.

966

967 Transmitted signals propagating into shallow water bounce off river bottom or surface and have 968 the potential to misdirect signals and cause multipathing. Obstructions in the river, such as the 969 weirs used to route migrating fish offshore of the sonars or parked boats, may also misdirect 970 transmitted signals from the tagged fish and reduce detection rates. Claisse et al. (2011) found 971 that detection range was greater in deeper, less structurally complex habitats. Their shallow 972 receivers (water depths 5 to 10 m) had a maximum detection distance of 30 m compared to 973 deeper receivers (15 to 20 m) whose maximum detection was 50 m. Ideally, receivers are 974 deployed in deeper water than our deployments. The DLs in our acoustic array were deployed at 975 similar depths in approximately chest-deep water. Low tides reduced the depth even further. To 976 ensure that fish were detected in the nearshore regions, it was necessary to position the DLs as 977 close to shore as possible because triangulating positions outside of the array is less effective. 978 Gjelland and Hedger (2013) found that depth range of transmitters (defined as the difference 979 between the deepest and shallowest transmitters) was second to wind in importance as a

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1026 Research Initiative [appropriation number 43106-18; Project 11329046, Nushagak Adult In-river

- 1027 Abundance].
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- **References**
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- Belcher, E.O., Hanot, W., Burch, J., 2002. Dual-Frequency Identification Sonar. Pages 187–192 [*In*] Proceedings of the 2002 International Symposium on Underwater Technology, April 16–19. Tokyo, Japan.
- Buck, G.B., 2013. Sonar Enumeration of Pacific Salmon Escapement into The Nushagak River, 2009. Alaska
- Department of Fish and Game, Fishery Data Series No. 13–34, Anchorage.
- Buck, G.B., Brazil, C.B., West, F., Fair, L., Zhang, X., Maxwell, S.L., 2012. Stock assessment of Chinook, sockeye, and chum salmon in the Nushagak River. Alaska Department of Fish and Game, Fishery Manuscript Series No. 12–05, Anchorage.
- Childs, A.R., Cowley, P.D., Naesje, T.F., Booth, A.J., Potts, W.M., Thorstad, E.B., Økland, F., 2008. Estuarine use

by spotted grunter *Pomadasys commersonnii* in a South African estuary, as determined by acoustic telemetry.

- African Journal of Marine Science 30(1), 123–132. https://doi.org/10.2989/AJMS.2008.30.1.12.462.
- Chittenden, C.M., Beamish, R.J., Neville, C.M., Sweeting, R.M., McKinley, R.S., 2009. The Use of Acoustic Tags
- to Determine the Timing and Location of the Juvenile Coho Salmon Migration out of the Strait of Georgia,
- Canada. Transactions of the American Fisheries Society 138(6), 1220–1225. https://doi.org/10.1577/T09-037.1.
- Claisse, J.T., Clark, T.B., Schumacher, B.D., McTee, S.A., Bushnell, M.E., Callan, C.K., Laidley, C.W., Parrish,
- J.D., 2011. Conventional tagging and acoustic telemetry of a small surgeonfish*, Zebrasoma flavescens*, in a
- structurally complex coral reef environment. Environ Biol Fish 91, 185-201.
- Crossman, J.A., Martel, G., Johnson, P.N., Bray, K. 2011. The use of Dual-frequency IDentification SONar
- (DIDSON) to document white sturgeon activity in the Columbia River, Canada. J. Appl. Ichthyol. 27(2), 53-57.
- https://doi.org/10.1111/j.1439-0426.2011.01832.x.
- Egg, L., Pander, J., Mueller, M., Geist, J., 2018. Comparison of sonar-, camera- and net-based methods in detecting riverine fish-movement patterns. Marine and Freshwater Research 69(12), 1905-1912.
- https://doi.org/10.1071/MF18068.
- Ehrenberg, J.E., Steig, T.W., 2003. Improved techniques for studying the temporal and spatial behavior of fish in a
- fixed location. ICES Journal of Marine Science 60, 700–706. https://doi.org/10.1016/S1054-3139(03)00087-0.
- El Mejjati, S., Bell, J., Botz, J., Faulkner, A., Maxwell, S., 2010. Using hydroacoustic methods to enumerate
- migrating salmon in the Copper River, Miles Lake sonar project, 2007. Alaska Department of Fish and Game, Fishery Data Series No. 10–98, Anchorage.
- English, K.K., English, B.L., Roias, S.M., 2017. Waanukv River Multi-Species Escapement Monitoring System
- Using Dual-Frequency Identification Sonar (DIDSON), ARIS Sonar, and Test Fishing. Rivers Inlet Salmon
- Initiative Steering Committee. LGL Limited, 9768 Second St., Sidney, BC, V8L 3Y8.
- Faulkner, A. V., Maxwell, S.L., 2015. The feasibility of using sonar to estimate adult sockeye salmon passage in the lower Kvichak River. Alaska Department of Fish and Game, Fishery Manuscript Series No. 15-05, Anchorage.
- Gaudet, D.M., 1990. Enumeration of migrating salmon populations using fixed-location sonar counters. Rapports
- et Proces-Verbaux des Reunions, Conseil International pour l'Exploration de la Mer 189, 197–209.
- Gjelland, K.Ø., Hedger, R.D., 2013. Environmental influence on transmitter detection probability in biotelemetry:
- developing a general model of acoustic transmission. Methods in Ecology and Evolution 4, 665-674.
- https://doi.org/10.1111/2041-210X.12057.
- Grote, A.B., Baily, M.M., Zydlewski, J.D., Hightower, J.E., 2014. Multibeam sonar (DIDSON) assessment of
- American shad (Alosa sapidissima) approaching a hydroelectric dam. Can. J. Fish. Aquat. Sci. 71, 545–558.
- https://doi.org/10.1139/cjfas-2013-0308.
- Hayden, T.A., Holbrook, C.M., Fielder, D.G., Vandergoot, C.S., Bergstedt, R.A., Dettmers, J.M., 2014. Acoustic
- Telemetry Reveals Large Scale Migration Patterns of Walleye in Lake Huron. PLoS ONE 9(12): e114833.
- https://doi.org/10.1371/journal.pone.0114833.
- Heublein, J.C., Kelly, J.T., Crocker, C.E., Klimley, A.P., Lindley, S.T., 2009. Migration of green sturgeon,
- *Acipenser medirostris*, in the Sacramento River. Environ Biol Fish 84, 245–258. https://doi.org/10.1007/s10641- 008-9432-9.
- Holmes, J.A., Cronkite, G.M.W., Enzenhofer, H.J., Mulligan, T.J., 2006. Accuracy and precision of fish-count data
- from a "dual-frequency identification sonar" (DIDSON) imaging system. ICES Journal of Marine Science 63,

543–555. https://doi.org/10.1016/j.icesjms.2005.08.015.

- Hinch, S.G., Rand, P.S., 2000. Optimal swimming speeds and forward-assisted propulsion: energy-conserving
- behaviours of upriver-migrating adult salmon. Can. J. Fish. Aquat. Sci. 57, 2470–2478.
- https://doi.org/10.1139/f00-238.
- Hughes, N.F., 2004. The wave-drag hypothesis: an explanation for size-based lateral segregation during the upstream migration of salmonids. Can. J. Fish. Aquat. Sci. 61, 103–109. https://doi.org/10.1139/f03-144.
- Humston, R., Ault, J.S., Larkin, M.F., Luo, J., 2005. Movements and site fidelity of the bonefish Albula
- vulpes in the northern Florida Keys determined by acoustic telemetry. Marine Ecology Progress Series 291, 237–
- 248. https:// doi:10.3354/meps291237.
- Kessel, S.T., Cooke, S.J., Heupel, M.R., Hussey, N.E., Simpfendorfer, C.A., Vagle, S., Fisk, A.T., 2014. A review
- of detection range testing in aquatic passive acoustic telemetry studies. Rev Fish Biol Fisheries 24, 199–218. https://doi.org/10.1007/s11160-013-9328-4.
- Mathes, M.T., Hinch, S.G., Cooke, S.J., Crossing, G.T., Patterson, D.A., Lotto, A.G., Farrell, A.P., 2010. Effect of
- water temperature, timing, physiological condition, and lake thermal refugia on migrating adult Weaver Creek
- sockeye salmon (*Oncorhynchus nerka*). Can. J. Fish. Aquat. Sci. 67, 70–84. https://doi.org/10.1139/F09-158.
- Maxwell, S.L., Gove, N.E., 2007. Assessing a dual-frequency identification sonar's fish counting accuracy,
- precision, and turbid river range capability. Journal of the Acoustical Society of America 122(6), 3364–3377. https://doi.org/10.1121/1.2799500.
- Maxwell, S.L., Faulkner, A.V., Fair, L., Zhang, X., 2011. A comparison of estimates from 2 hydroacoustic systems used to assess sockeye salmon escapement in 5 Alaska Rivers. Alaska Department of Fish and Game Fishery
- Manuscript Series No. 11–02, Anchorage.
- Maxwell, S. L., Faulkner, A.V., Hacklin, T.D., 2013. Evaluating error in sockeye salmon abundance estimates from
- salmon traveling outside the sonar beam at the Yentna, Copper, and Kenai rivers. Alaska Department of Fish and Game, Fishery Manuscript Series No. 13-07, Anchorage.
- Maxwell, S.L., Buck, G.B., Faulkner, A.V., 2019. Using Acoustic Telemetry to Expand Sonar Escapement Indices
- of Chinook Salmon to In-river Abundance Estimates, Mendeley Data, V1, https://doi: 10.17632/c8jr8x9r4y.1.
- Maxwell, S.L., Buck, G.B., Faulkner, A.V., *In press*. Expanding Nushagak River Chinook salmon escapement
- indices to inriver abundance estimates using Acoustic Tags. Alaska Department of Fish and Game, Fishery 1107 Manuscript Series No. , Anchorage.
- McDougall, M. J., Lozori, J.D., 2018. Sonar estimation of Chinook and fall chum salmon passage in the Yukon
- River near Eagle, Alaska, 2017. Alaska Department of Fish and Game, Fishery Data Series No. 18-20,
- Anchorage.
- McMichael, G.A., Eppard, M.B., Carlson, T.J., Carter, J.A., Ebberts, B.D., Brown, R.S., Weiland, M., Ploskey,
- G.R., Harnish, R.A., Deng, Z.D., 2010. The juvenile salmon acoustic telemetry system: A new tool. Fisheries 35(1), 9–22. https://doi.org/10.1577/1548-8446-35.1.9.
- Miller, J.D., 2000. Sonar enumeration of Pacific salmon escapement into Nushagak River, 1999. Alaska Department
- of Fish and Game Regional Information Report No. 2A00–19, Anchorage.
- Miller, J.D., Burwen, D.L., Fleischman, S.J., 2013. Estimates of Chinook salmon passage in the Kenai River using
- split-beam and dual-frequency identification sonars, 2010. Alaska Department of Fish and Game, Fishery Data Series No. 13–58, Anchorage.
- Mora, E.A., Lindley, S.T., Erickson, D.L., Klimley, A.P., 2015. Estimating the Riverine Abundance of Green
- Sturgeon Using a Dual-Frequency Identification Sonar. N. Am. J. Fish. Manage. 35(3), 557–566.
- https://doi.org/10.1080/02755947.2015.1017119.
- Pipal, K.A., Notch, J.J., Hayes, SA, Adams, P.B., 2012. Estimating escapement for a low-abundance steelhead
- population using Dual-Frequency Identification Sonar (DIDSON). N. Am. J. Fish. Manage. 32(5), 880–893.
- https://doi.org/10.1080/02755947.2012.697096.
- Rakowitz, G., Kubečka, J., Fesl, C., Keckeis, H., 2009. Intercalibration of hydroacoustic and mark–recapture
- methods for assessing the spawning population size of a threatened fish species. Fish Biology 75(6), 1356–1370. https://doi.org/10.1111/j.1095-8649.2009.02368.x.
- Rawding, D., Liermann, M., 2011. Comparison of DIDSON sonar based estimates of Chinook salmon escapement
- with other methods in the Coweeman River, Washington. The Journal of the Acoustical Society of America
- 129(4), 2692. https://doi.org/10.1121/1.3589030.
- Reimer, A. M., Fleischman, S.J., 2016. Stock-specific abundance and run timing of Chinook salmon in the Kenai
- River, 2007–2014. Alaska Department of Fish and Game, Fishery Manuscript Series No. 16-06, Anchorage.
- Reynolds, J.H., Woody, C.A., Gove, N.E., Fair, L.F., 2007. Efficiently estimating salmon escapement uncertainty using systematically sampled data. American Fisheries Society Symposium 54, 121–129.
- Schumann, K. J., McIntosh, B.C., 2017. Sonar estimation of salmon passage in the Yukon River near Pilot Station,
- 2014. Alaska Department of Fish and Game, Fishery Data Series No. 17-31, Anchorage.
- 1137 Seber, G.A.F., 1982. On the estimation of animal abundance and related parameters. 2nd-edition. Charles Griffin and
- Sons, Ltd., London, pp. 7–9, 64–69, 654.
- Seibel, M.C., 1967. The use of expanded ten-minute counts as estimates of hourly salmon migration past counting
- towers on Alaskan rivers. Informational Leaflet 101, Alaska Department of Fish and Game, Division of Commercial Fisheries, Juneau.
- Simmonds, E.J., MacLennan, D.N., 2005. Fisheries Acoustics: Theory and Practice. Blackwell Science Ltd.,
- Oxford, UK, pp.39 and 69.
- Starr, R.M., O'Connell, V., Ralston, S., Breaker, L., 2005. Use of Acoustic Tags to Estimate Natural Mortality,
- Spillover, and Movements of Lingcod (*Ophiodon elongatus*) in a Marine Reserve. Marine Technology Society
- Journal, 39(1), Spring, 19–30(12). https://doi.org/10.4031/002533205787521677.
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- **Tables**
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1158 a Detected by one or more data loggers. 1158
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1159 **b** Position estimates output by ALPS or U-Map (Lotek, Inc.'s Asynchronous Logging Positioning Software).

Table 2. Deviations of beacon position estimates from their actual locations.

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1165 Table 3. Fish tracks categorized by their movement through the acoustic array (%).

1166 ^a An implied trip is from a tagged fish that traveled through the array but wasn't detected.

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- 1170
- 1171 Table 4. Mean lengths (mid eye to tail fork) of Chinook salmon from the sonar project's Age-Sex-

^a Inside fish are tagged Chinook salmon that passed through a sonar beam footprint.

^b Outside fish passed offshore of the beam footprints.

 c The 66 cm cutoff is used to separate small and large Chinook salmon at the Nushagak River (Chuck Brazil, ADF&G, Anchorage, personal communication).

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1178 Table 5. Kolmogorov-Smirnov Goodness-of-Fit results (KS) comparing length distributions of
- 1179 Chinook salmon.

^a Inside fish are acoustically tagged Chinook salmon that passed through a sonar beam footprint. Outside fish passed 1181 offshore of sonar beam footprints. 1181 offshore of sonar beam footprints.
1182 bSonar fish refers to Chinook salm

^b Sonar fish refers to Chinook salmon lengths from gillnet captures used to apportion sonar estimates.

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Table 6. Proportions and in-river abundance estimates for Chinook salmon.

^a SE values in this column are not averages, they are the SE for all years.

^b The number of tagged fish includes the filtered fish tracks; incoherent tracks were removed from the dataset. (Note: this is the number of fish, not fish trips.)

^c Greg Buck, ADF&G, Anchorage, Alaska personal communication.

^d These totals do not match n above because 2 fish from 2011 did not get length measurements, and 4 tags from 2011 and 1 from 2011 of detected fish did not match any tag insertion numbers.

Table 7. Bank ratios of Chinook salmon from DIDSON and tagged fish estimates.

^a Greg Buck, ADF&G, Anchorage, Alaska, personal communication.

^b Tagged fish percentages only include tagged fish that traveled through a DIDSON footprint.

Table 8. Climatological and water data, Nushagak River sonar site, 2011–2014.

Wind speed (km/h) 2000

Air temperature (°C) 2000 h

^a In 2011, daily precipitation, and twice daily wind speed and air temp were averaged from METAR (Meterological Terminal Aviation Routine Weather Report) stations at Dillingham and New Stuyahok. From 2012-2014, measurements were from the sonar site using a Davis Vantage Vie wireless weather station.

^b Water temperature was recorded with a UA-001-08 HOBO data logger attached to the right-bank DIDSON mount

in 1-h (2011), 2-h (2013) and 5-min (2012 and 2014) increments.

Figure Captions for: Using Acoustic Telemetry to Expand Sonar Escapement Indices of Chinook Salmon to In-river Abundance Estimates

- 1. Acoustic tag insertion and detection sites in the lower Nushagak River.
- 2. Bathymetry of the acoustic tag insertion site within the Nushagak River's main channel (top-left) and side channel (bottom-right), 6/8/2011.
- 3. The tube used to insert acoustic tags into salmon (left), and a tag recovered by the testfish crew at the sonar site (right).
- 4. Bathymetry map of the Nushagak River at the sonar site (top) and a river bottom profile (bottom) extracted from the bathymetry data to show the DIDSON deployment sites along either side of the river, 6/7/2011.
- 5. Bathymetry map of the region encompassing the acoustic array, 7/2/2012.
- 6. Change in the river bottom from 2011 to 2012 showing the deposition and erosion that occurred.
- 7. A) The acoustic array showing the position of the DIDSON beams (large triangles), shorelines, data loggers (solid, black diamonds), and rectangular regions of uncertainty around each beam. B) The probability regions used to classify tagged fish as either inside or outside of the DIDSON beam, with a probability of 1.0 assigned to fish passing through the inner rectangle and probabilities of 0.5 and 0.25 assigned to fish passing through the middle and outer rectangles, respectively.
- 8. The quality of the fish tracks varied widely from high-quality tracks that required minimal filtering (A), to tracks that were discarded because the actual route of the fish could not be determined (B), too few points were obtained (C), or the points were randomly scattered with no coherent track (D). Open black circles represent tagged fish position estimates, closed circles represent the shoreline, H's represent beacons, and triangles the sonar positions.
- 9. Example of a fish track showing the first level of filtering (top) and secondary level (bottom). The left side of the track was discarded during the second stage because the segment was outside the array boundaries.
- 10. Layout of the beacons (B) and data loggers (DL) along with a 20,000-point random sample of minimally filtered position estimates from the stationary beacons used to synchronize the data loggers, 2014.
- 11. A 2,000-point random sample of position estimates showing the beacons from LB and RB with the tightest spread of points (beacons 31 and 34) and with the widest spread (beacons 36 and 35), with 5 and 10 m distances marked around the GPS-measured beacon locations (rectangles), 2014.
- 12. Examples of tagged fish clearly traveling through a DIDSON beam footprint (A), traveling outside of the footprint (B), creating a short track that ended before reaching the footprint (C), and traveling through the edge of the footprint (D).
- 13. Example of a tagged fish that moved through both DIDSON beam footprints.
- 14. Fish 332, a 3-trip fish that created a large looping track in its first trip (A), returned 4 d later heading downriver (B), and then 2 d later traveled upriver along left bank where detection was lost prior to reaching the DIDSON beam footprint (C).
- 15. Fish 416, a 2-trip fish that traveled upriver along the left bank, returned 7 d later along left bank where it held for a period of time before crossing the river and looping between

right bank and the river's center. No additional upriver trip was observed, so it was assumed the fish moved downriver and did not return.

- 16. Depth of tagged Chinook salmon moving upriver through the study area, 2012.
- 17. Depth of tagged Chinook salmon moving downriver through the study area, 2013.
- 18. A comparison of Chinook salmon length distributions from fish captured at the sonar site for apportioning the sonar estimates to species (Sonar) and acoustic-tagged fish that migrated through a sonar beam footprint (Tag-inside).
- 19. A comparison of Chinook salmon length distributions from acoustic-tagged fish that migrated through a sonar beam footprint (Tag-Inside) and fish that migrated offshore of the sonar beam footprint (Tag-Outside).
- 20. A comparison of Chinook salmon length distributions from fish captured for the acoustic tag and mark-recapture (MR) studies (Maxwell et al. *In press a*).
- 21. Fish holding regions within the acoustic array based on the number of detections per tagged fish where areas with more than 50 detections per fish per 100 m^2 indicate holding fish, 2011 (Maxwell et al. *In press a*).

Distance from left bank (m)

Beacon 31

Beacon 36

Note: All fish tracks are moving upriver (dark to light progression from left to right). Diamonds show the positions of the hydroacoustic receivers.

(07/10/2011 00:25:02 - 07/10/2011 04:53:46)

300 400 500 600 700 800 900 Easting-570000 (m)
(06/29/2012 22:46:36-06/29/2012 22:57:11)

