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ARTICLE

Understanding Smolt Survival Trends in Sockeye Salmon

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Abstract

Many populations of Sockeye Salmon *Oncorhynchus nerka* in the eastern North Pacific Ocean experienced significant productivity declines that began about 1990, but there is no consensus on the mechanisms responsible. To better understand Sockeye Salmon survival trends, we examined the 50-year time series for two age-classes of Sockeye Salmon smolts from Chilko Lake in central British Columbia. Arranging survival time series for both age-classes by ocean entry year and combining them, weighted by a proxy model of sampling variance, reduced the sampling variance in the original age-1 smolt survivals sufficiently to indicate a linear trend of increasing survival from 1960 to 1990 that suddenly changed at or near 1991 to a lower and declining trend from 1992 to 2008. Neither density nor mean length influenced smolt survival. Returns in a given year were not good predictors of siblings returning in subsequent years. Time spent at sea increased linearly beginning around 1970. Although smolt survivals differed between ecosystem regimes, there was only the one clear pattern break about 1991. To improve our understanding of mechanisms, survival trends were compared with environmental indices that included catches and hatchery releases of potentially competing salmon from around the North Pacific Ocean. Smolt survivals were more similar to abundance indices of Sockeye Salmon, Chum Salmon *O. keta*, and Pink Salmon *O. gorbuscha* than to indices of global, regional, or local ocean climate. Our results are consistent with the hypothesis that salmon productivity in the North Pacific declined soon after 1990. We present a simple model to illustrate how increased competition at sea, related to the release of large numbers of hatchery salmon, in conjunction with changes in ocean productivity, may have played a significant role in improving Sockeye Salmon survivals while reducing their growth before 1991. After 1991, these factors may have acted to reduce survivals while the growth of survivors showed no effect.

Anadromous Pacific salmon *Oncorhynchus* spp. comprise a multispecies complex of varying productivities. Their recent abundance in the Pacific Ocean, as reflected by commercial catch, is as high as it has ever been (Irvine and Fukuwaka 2011). Global abundances are driven primarily by Pink Salmon *O. gorbuscha* and Chum Salmon *O. keta*, as well as, particularly in the eastern North Pacific Ocean, by Sockeye Salmon *O. nerka* (Eggers 2009; Ruggerone et al. 2010; Irvine and Fukuwaka 2011). The status of Sockeye Salmon populations varies among regions however, and in British Columbia's Fraser River, low

numbers of returning salmon in recent years are a major concern (Grant et al. 2011; Rand et al. 2012).

The Fraser River watershed is one of the world's greatest salmon producers (Northcote and Larkin 1989), although numbers returning annually are highly variable. Sockeye Salmon are the most economically valuable salmon species in the watershed, and have provided a commercial harvest since the early 1870s (Meggs 1991) and a First Nations (native North American) harvest for millennia. Returning Sockeye Salmon are divided into three major groups, based on run timing, that comprise at least

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22 Conservation Units, biological groups of salmon that are genetically or ecologically distinct from each other (Holtby and Ciruna 2007). Declining Sockeye Salmon returns from 1992 to 2009, and an exceptionally low 2009 return (smallest since 1947) resulted in the Canadian government establishing a judicial inquiry on Fraser River Sockeye Salmon (Cohen 2012a, 2012b, 2012c). Ironically, the inquiry was barely underway when 2010 saw the largest Sockeye Salmon return in the previous 100 years. The inquiry represented 2.5 years of work that included 128 d of evidentiary hearings, 2,145 exhibits, and testimony from 179 expert witnesses (Cohen 2012c:86). Fifteen detailed technical reports plus various primary publications (e.g., Beamish et al. 2012; Connors et al. 2012; Peterman and Dorner 2012; Preikshot et al. 2012; Thomson et al. 2012) were produced. Beamish et al. (2012) attributed the low returns for Fraser River Sockeye Salmon in 2009 to local effects within the Strait of Georgia, while Thomson et al. (2012) documented the importance of various factors at different life history stages in determining survival. The inquiry concluded that marine factors in particular were implicated in the broad-based regional decline of Sockeye Salmon stocks, but was unable to determine the relative importance of specific factors (Cohen 2012c:88).

Temporal patterns in salmon survival have been influenced by many factors including major ecosystem regime shifts in 1977 and 1989 (e.g., Beamish and Bouillon 1993; Hare and Mantua 2000; Irvine and Fukuwaka 2011). Some researchers also identified 1999 as a shift (e.g., Peterson and Schwing 2003). With respect to Fraser River Sockeye Salmon, Beamish et al. (2004b) found survivals were high following the 1977 shift and declined after 1989. In contrast, McKinnell and Reichardt (2012) reported no increase in Fraser River Sockeye Salmon survival after 1977, although they found first-year marine growth abruptly declined about 1977. Ruggerone et al. (2003) attributed reduced survival of Sockeye Salmon in Alaska to increased competition with Pink Salmon. In the case of Fraser River Sockeye Salmon, Connors et al. (2012) concluded the effect of competition with Pink Salmon could be exacerbated if Sockeye Salmon were exposed to farmed salmon during their out-migration, particularly during warm ocean conditions. Ruggerone et al. (2010) and others have raised concerns that density-dependent interactions in the ocean resulting from hatchery releases may reduce Pacific salmon growth and survival.

Many analyses of Pacific salmon survivals rely on stock-recruit data. Results from these studies, often conducted across many stocks, have greatly improved our understanding of the underlying patterns of Sockeye Salmon growth (Peterman 1984) and survival (Peterman et al. 1998; Pyper et al. 2005; Peterman and Dorner 2012). For example, Peterman and Dorner (2012) demonstrated productivity declines since the 1990s or earlier for many Sockeye Salmon stocks from Washington to southeast Alaska and suggested that climate-driven increases in mortality induced by pathogens, as well as increased predation or reduced food due to oceanographic changes, may be the explanation. Yet stock-recruit data typically do not allow one to separate effects

occurring at different life history stages. Of the numerous Fraser River Sockeye Salmon populations, there are only two, Cultus Lake in the lower Fraser River watershed and Chilko Lake in the central, where survival is estimated before and after smolts leave their rearing lake. The Cultus Lake time series is relatively short, but at Chilko Lake, 775 km upstream from the ocean, a reasonably consistent and nearly unbroken time series of abundance and length by age for spawners and smolts has been collected since the early 1950s. Most Sockeye Salmon smolts exit Chilko Lake during spring at age 1, but some spend a second year in freshwater. Age-1 and age-2 smolts leaving the lake during the same year have different parents and a different freshwater life history but are exposed to similar marine conditions. Previous researchers examining the Chilko Sockeye Salmon time series generally focused on the predominant life history, age-1 smolts, most of which return as adults after 2.5 years at sea (e.g., Henderson and Cass 1991; McKinnell 2008; Grant et al. 2011). Although survival estimates for age-2 smolts have a larger sampling error than estimates for the more abundant age-1 smolts (Bradford et al. 2000), we recognized that information gleaned from age-2 smolt survivals could help to clarify survival patterns and understand the processes shaping survival and population dynamics for Sockeye Salmon.

To better understand the role of marine factors affecting Sockeye Salmon survival, we focused on the postsmolt life history of Chilko Lake Sockeye Salmon. We compared age-1 smolt survival estimates with those of age-2 smolts leaving the lake in the same year, investigated uncertainty associated with these estimates, and developed a new survival time series. We evaluated the influence on survival of smolt abundance (density dependence), smolt quality (age, length, and condition), and environmental trends (various North Pacific and regional ocean climate indices). Finally, we examined the potential effects on Sockeye Salmon survival and growth patterns of density-dependent processes at sea as represented by indices of multispecies salmon abundance.

METHODS

Sockeye Salmon life history and data sources.—A brief synopsis of the life history of Chilko Lake Sockeye Salmon and how data were gathered follows. We began our analyses with the fish that were spawned in 1958, the beginning of a continuous series. Smolt numbers were not estimated in 1991. Smolt and spawner sampling methodologies were reviewed by Henderson and Cass (1991), Roos (1991), and Grant et al. (2011).

Chilko Lake Sockeye Salmon start life as fertilized eggs in the fall of the brood year, emerge as fry from spawning sites the next spring, and spend 1 or 2 years in Chilko Lake before exiting as smolts during spring. When smolts exit the lake after usually 1 year, they pass through an enumeration fence at the lake outlet where their numbers are estimated photographically. A random sample of smolts is measured daily for length, and smolts larger than a specified length are collected for age determination.

That specified length, typically but not always 85 mm (K. Brenner, Fisheries and Oceans Canada [DFO], Kamloops, personal communication) is smaller than the minimum length of age-2 smolts. The first smolt records we analyzed were collected in 1960, recording age-1 smolts (S_1) from brood (spawning) year 1958 and age-2 smolts (S_2) from brood year 1957. In addition, samples of age-1 and age-2 smolts (determined by size) were taken at the counting facility, from near the beginning, the peak, and the end of most smolt runs, often several hundred smolts per year, and preserved in 10% formalin. Subsequently, ages were confirmed from scales, and fish were weighed and measured for length to evaluate changes in fish size and condition (J. Hume, DFO Cultus Lake, personal communication).

Most S_1 return to freshwater as maturing adults in late summer after two winters in the ocean and were categorized as $R_{1,2}$, meaning they lived a total of three winters, one in freshwater (in addition to one winter as a developing egg), and two in the ocean (Figure 1). A small proportion of S_1 spend one or three winters at sea, returning as $R_{1,1}$ and $R_{1,3}$. Age-2 smolts (S_2) stay for a second growing season and a second winter in Chilko Lake and also typically spend two winters in the ocean ($R_{2,2}$), but again some return after one or three winters at sea, as $R_{2,1}$ and $R_{2,3}$. Some life history stages such as S_3 and $R_{2,3}$ were rare and we have no records of $R_{3,2}$.

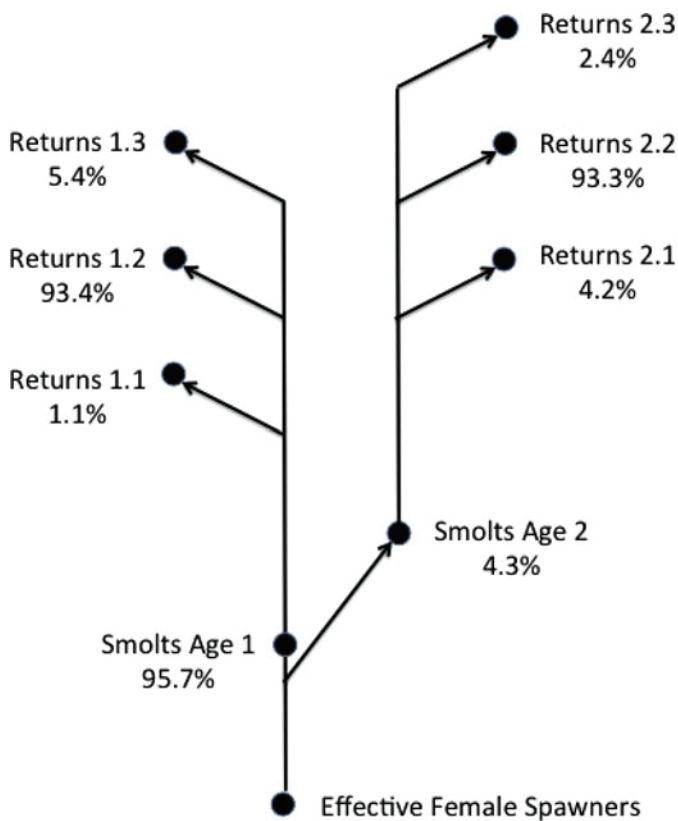


FIGURE 1. Chilko Sockeye Salmon life history paths. There are nine observations over 6 years for each brood. Percentages given are the mean values over the period 1960–2008.

The Pacific salmon fisheries were sampled for stock and age composition to estimate the numbers of Chilko Lake Sockeye Salmon mixed with other salmon stocks in catches, on their return migration. Allocation of Sockeye Salmon to discrete stocks was based on distinctive freshwater growth patterns seen on the scales (Gable and Cox-Rogers 1993), supplemented by DNA markers beginning in 2000 (Grant et al. 2011).

The number of fish escaping fisheries and returning to Chilko Lake to spawn were estimated prior to 2009 by mark–recapture (e.g., Schubert and Fanos 1997), except for 1967 when they were estimated by expanding visual counts at Henry’s Bridge, 12 km below the lake (Grant et al. 2011). Sockeye Salmon carcasses were sampled throughout the Chilko Lake spawning areas to obtain scales and measure length. Scales and otoliths were collected so that estimates of spawner abundance could be partitioned into age-classes. Since 2006, imaging sonar (DIDSON) has been used to estimate spawner abundance (Cronkite et al. 2006; Holmes et al. 2006). Numbers estimated by DIDSON and by mark–recapture were similar and, since 2009, DIDSON estimates have replaced mark–recapture (Benner, and T. Cone, DFO Annacis Island, personal communication[s]). Total adult returns, as catch in the fisheries plus escapement to the spawning grounds, were estimated each year and then, based on the ages of spawners collected at Chilko Lake, divided into six life history categories (Figure 1).

Terms and abbreviations for this analysis are described in Table 1. Variables are in italic uppercase: e.g., $S_{1,t}$, $R_{1,2,t+2}$, SAR_1 . Fitted regression parameters are in italic lowercase: e.g., a , b_1 . Means were reported with the standard error in parentheses, i.e., (SE), unless explicitly stated otherwise. Correlation probabilities were corrected for shared autocorrelation by estimating the effective degrees of freedom (Pyper and Peterman 1998).

Smolt survival estimates.—Smolt survival was referenced to the ocean entry year t when smolts left Chilko Lake and were first observed. This survival calculation included the period when smolts migrated downstream from Chilko Lake to the ocean, as well as their time at sea. The terms $S_{1,t}$ and $S_{2,t+1}$ thus refer to age-1 and age-2 smolts from the same brood year but having different ocean entry years. The total returns from age-1 smolts,

$$R_{1,t} = R_{1,1,t+1} + R_{1,2,t+2} + R_{1,3,t+3}, \quad (1)$$

were the survivors of $S_{1,t}$ from brood year $t - 2$ that entered the ocean in year t and returned 1, 2, or 3 years later. Similarly, $R_{2,t}$ were the survivors of $S_{2,t+1}$ from brood year $t - 3$ that also entered the ocean in year t . The estimate of postlake survival for S_1 is the ratio of smolts to adult returns, i.e., $SAR_1 = R_1/S_1$; SAR_1 was a ratio of estimated variables, each with considerable sampling variance, and had a skewed distribution.

Losses between returns and spawning.—A Chilko Lake Sockeye Salmon that recruited to the fishery in a given year could be from one of six age-groups and four brood years (Figure 1). Recruitment for each year was rebuilt from data

TABLE 1. Chilko Sockeye Salmon life history stages, definitions of abbreviations, and statistics for brood years 1956–2006. Spawners, smolts, recruits, and returns are expressed as millions of fish.

Variable	Definition	Mean	SD	Range
<i>EFS</i>	Effective female spawners	0.204	0.152	0.0150–0.598
Ψ	Loss of recruits before spawning (%)	69.4	19.9	5.9–94.9
	Ψ 1960–1994	79.1	8.3	51.6–94.9
	Ψ 1995–2009	48.6	21.7	5.9–82.0
L_1	Mean fresh FL of S_1 (mm)	83.24	6.04	73.06–100.00
L_2	Mean fresh FL of S_2 (mm)	117.27	11.51	98.42–160.31
R_1	Returns from S_1 , all ages	1.469	1.211	0.0646–4.741
R_2	Returns from S_2 , all ages	0.0524	0.0577	0.000714–0.321
<i>Recruits</i>	Adults entering the fisheries in a year	1.52	1.16	0.152–4.63
S_1	Age-1 smolts	19.92	15.54	0.159–77.128
S_2	Age-2 smolts	0.630	0.623	0.0469–2.479
SAR_1	Survival of age-1 smolts, R_1/S_1 (%)	8.5	5.2	0.34–25
	harmonic mean of SAR_1 (%)	6.9	3.8, 7.6 ^a	1.6–31 ^a
SAR_2	Survival of age 2 smolts (%)	9.9	7.3	1.3–29
	weighted mean of SAR_2 (%)	10.4	4.7	1.9–30 ^b
	weighted harmonic mean SAR_2 (%)	8.22	3.1, 13.2 ^a	1.9–34.6 ^b
SAR	Weighted mean of SAR_1 and SAR_2 (%)	8.57	5.01	0.38–21.4
	harmonic mean of SAR (%)	6.94	2.7, 11.9 ^a	1.6–30.8 ^b
W_1	Preserved weight of S_1 (g)	4.75	1.31	3.11–10.47
W_2	Preserved weight of S_2 (g)	14.07	4.95	7.780–32.35

^aOne SD below and above mean, transformed from log scale to arithmetic scale.

^b95% confidence limits: $\exp\{\text{mean}[\log(x)] \pm 2 \text{SD}[\log(x)]\}$.

on abundance by age-class by brood, such that the youngest recruits in a given year are $R_{1,1}$ spawned 3 years previously, and the oldest are $R_{2,3}$ spawned 6 years previously.

The time series of effective female spawners, *EFS*, was used to estimate smolt survival after fish could recruit to fisheries. Our estimate of losses between recruitment and spawning, primarily due to fisheries exploitation but including en route mortality and pre-spawning mortality, was

$$\Psi = 1 - 2 \text{EFS} / \text{Recruits} \quad (2)$$

and assumed that total spawners were approximately twice the number of female spawners.

Sampling variance.—About 2,400 smolts and 660 spawners were sampled for age determination each year, and S_2 and R_2 comprised about 4% of these samples, so the age distribution of smolts and returns had a large binomial sampling variance. There were additional sources of error in the estimates of total abundance of smolts, returns, and spawners before applying age distributions. To understand the precision of S_2 and R_2 , we examined the sample sizes for age data, although these were only available for recent years: 2005–2010 for smolts and 2000–2009 for returns. Binomial confidence limits (BCL) (Zar 1999:528) were calculated for smolts and returns using the R function `qbinom` (Pr, n, p), where n was the total number of smolts of age 1 and 2, $p = S_2/n$, and Pr was 0.975 or 0.025 for the upper or lower

95% BCL. This function calculated the smallest value of x such that $f(x) \geq p$, where f was the cumulative binomial frequency distribution (R Development Core Team 2012). Because fish below a specific size were assumed to be S_1 and not sampled for scales, the proportion of S_2 in the age samples were higher than in the smolt population, but we expected that BCLs for S_2 and R_2 in the age samples would be proportional to those of the population and would estimate the relative precision between years for S_2 and R_2 abundance.

Comparing and combining the SAR_1 and SAR_2 time series.—We compared the time series for SAR_1 and SAR_2 with plots, correlations, and harmonic means (log transformations produced nearly normal distributions for survival). Principal components analysis was used to determine a common factor (PC_1) from the two time series. We used generalized additive modeling (GAM, R package `mgcv`) on log-transformed survival as an objective comparison of the trends in SAR_1 and SAR_2 (Wood 2006; Zuur et al. 2009; R Development Core Team 2012). We tested whether a GAM smoother for each age was better than a single smoother for combined ages using the Akaike information criterion (AIC) and ANOVA.

We wished to improve the time series SAR_1 with information provided by SAR_2 but this required appropriate weights (w) for SAR_2 because the estimates for S_2 and R_2 abundances had wide BCLs in some years, which resulted in some unreliable estimates for SAR_2 . Despite having details for age samples for only

a few years, we developed a model (see Appendix) to parameterize the sampling variance of SAR_2 and provide appropriate weights, based on consistency between the age distribution of emigrating smolts in a year and the age distribution of returning survivors from those smolts 2 years later. These weights were used whenever the variables S_2 and R_2 appeared in regressions. We combined the survival estimates for each year as a weighted geometric mean that would have the same mean as SAR_1 ,

$$\log_e(SAR) = \{\log_e(SAR_1) + w[\log_e(SAR_2) - D]\}/(1 + w), \quad (3)$$

where w was a weight from 0 to 1 and D was $\text{mean}[\log_e(SAR_2)] - \text{mean}[\log_e(SAR_1)]$.

Effect of smolt density on survival.—We used the Ricker model, $R = \alpha S e^{\beta S}$, to examine decreasing smolt survival with increasing smolt abundance. After log transformation,

$$\log_e(R/S) = b_0 + b_1 S; \quad (4)$$

α was estimated as $\exp(b_0)$. Tanaka (1962) and Larkin (1973) pointed out that independent random numbers for R and S in this regression will produce a significant value of b_1 . McKinnell (2008) compared fits for Fraser River Sockeye Salmon stocks to equation (4) and showed that as certainty in the stock–recruit relationship decreased, estimates for α and β increased. Incorrectly high values for Ricker model parameters suggest higher exploitation rates and lower target stock sizes, inappropriate for management under uncertainty, so a better understanding of equation (4) was important. We added the line $f(S) = \log_e[\text{mean}(R)/S]$ to plots from equation (4) to indicate the expected result when there was no relationship between R and S , but note that R approaches 0 as S approaches 0. Using nonlinear regression to fit the Ricker model indicated a bias from equation (4), but when the data were log transformed to account for increasing residuals with increasing smolt abundance, then nonlinear regression produced the same parameter estimates as did equation (4).

The second measure of smolt survival, SAR_2 , allowed us to develop a novel test for density dependence that avoided some of the problems of equation (4), i.e.,

$$\log_e(R_2/S_2) = b_0 + b_1 S_1, \quad (5)$$

where the three variables were registered to the same ocean entry year. This approach assumed that S_2 and S_1 coexist in the ocean such that density was effectively described by S_1 (which constituted, on average, 96% of the returns of Chilko Lake Sockeye Salmon).

Effects from ecosystem regime shifts and smolt quality.—Our starting point for examining temporal trends in smolt survival was

$$\log_e(R) = b_0 + b_1 \log_e(S) + N(0, w\sigma), \quad (6)$$

where S was the number of smolts of age 1 and age 2 (separately) in a specific ocean entry year, R was the number of returns over all ages from those smolts see equation (1), and the error term, $N(0, w\sigma)$, described normal residuals weighted by w to correct for heterogeneous variance in residuals due to heterogeneous sampling variance in S_2 and R_2 . The x -intercept, b_0 , was the mean log survival. Because the regression slope, b_1 , was related to the correlation coefficient, $b_1 = r(\sigma_y/\sigma_x)$, sampling variance would have resulted in $b_1 < 1$ when the true value was 1. If b_1 was significantly less than 1, indicating diminishing returns in R with increasing S , we interpreted that as density-dependent survival. We preferred equation (6) to equation (4) because it avoided the ratio estimators SAR_1 and SAR_2 , identified the effect of smolt abundance explicitly, and the increased range of smolt abundance reduced the effect of error in the independent variable. The residuals from this regression arose from sampling variance, environmental effects, and smolt quality, all of which were important to know.

To evaluate ecosystem regime shifts, we assumed that shifts began in specific years 1977, 1989, and 1999 (as referenced in Introduction). We examined the effect on smolt survival of four regimes between 1960 and 2008 using the regression

$$\log_e(R) = b_0 + b_1 \log_e(S) + b_{2,i} P_i + N(0, w\sigma), \quad (7)$$

where P_i was a factor identifying $i = 1-4$ regimes. We compared the fit for equation (7) to the fit for equation (6) using ANOVA to determine overall significance of the regimes, and then compared regimes using Tukey's honestly significant difference test (Zar 1999) to adjust probabilities for multiple comparisons. As a contrast to discrete regimes, we also fit a smoothly varying environmental trend to the survival from both smolt age-groups, as follows:

$$\log_e(R) = b_0 + b_1 \log_e(S) + s(t) + N(0, w\sigma), \quad (8)$$

where $s(t)$ was a GAM smoother for the ocean entry years. We simplified the environmental trends from equation (8) using segmented linear regression,

$$\log_e(R) = b_{0,i} + b_1 \log_e(S) + b_{2,i} T_i(t) + N(0, w\sigma), \quad (9)$$

where $T_i(t)$ was a factor identifying the years in different survival regimes. The result was a series of lines with intercepts $b_{0,i}$ and slopes $b_{2,i}$ that described temporal trends in smolt survival.

We added the mean annual values for smolt length, L , condition, K , and a factor for age, A_j to equation (9) to establish the following:

$$\log_e(R) = b_{0,i} + b_1 \log_e(S) + b_{2,i} T_i(t) + b_3 L + b_4 K + b_{5,j} A_j + N(0, w\sigma), \quad (10)$$

where K was the deviation (percent) of observed from predicted weight based on a log-log regression of preserved weights on

preserved lengths (as opposed to assuming isometric growth). This was processed as forward and backward stepwise regression to assess the influence of these variables on smolt survivals.

Variation in age composition and marine residence time.—We examined interannual variation in the proportions of S_1 and S_2 that spent 1 or 3 years in the ocean, rather than the usual 2 years (Figure 1) by plotting time series of age composition of returns. We also examined correlations between returns within a brood but after differing lengths of time at sea. We removed the smolt effect from these correlations, hoping to clarify an effect due to the ocean entry year, as follows. Regression through the origin, $R_{1,3} = bR_{1,2}$, established the base case. We removed the effect of smolt numbers from $R_{1,3}$ by fitting $R_{1,3} = bS_1$ and extracting residuals which we notated as $R^*_{1,3}$. We calculated $R^*_{1,2}$ similarly. The regression $R^*_{1,3} = a + bR^*_{1,2}$ then tested for an effect from their common ocean entry year and from two sea winters in common.

Marine indicators of high seas survival processes.—We created time series plots for 24 environmental variables arranged by salmon ocean entry year and calculated the trends in these before and after 1991 for comparison to trends in SAR. We also examined scatter plots of SAR or $\log_e(\text{SAR})$ against each variable and used linear regression to measure how well each variable predicted SAR before and after 1991. In almost all cases, data sets covered the full duration of the SAR time series. We computed results with and without the 2007 SAR₁ ocean entry year, an outlier associated with the 2009 fisheries failure, but differences were minor and we only provided results for the complete data set.

To represent density-dependent processes at sea, we used commercial catch data and hatchery release estimates from Irvine et al. (2012). Commercial catch data adequately track total salmon abundance at the species level and at the scale of the northeast and northwest Pacific Ocean (Irvine and Fukuwaka 2011). Although others (Eggers 2009; Ruggerone et al. 2010) have combined catch and spawner escapement data to develop run size biomass estimates for salmon, statistical comparisons between the time series generated by these approaches have not been made. We worked with data for Sockeye Salmon as well as for Pink and Chum Salmon since these were the most abundant salmon species and density-dependent interactions among them may be occurring (Ruggerone et al. 2010). Data were aggregated for the northeast Pacific (North America) and northwest Pacific (Asia). Catches were shifted backward by 1 year for Pink Salmon and 2 years for Chum and Sockeye Salmon to correspond to the same ocean entry year as the SAR index. Because interstock salmon abundance was hypothesized to be a density-dependent effect, corresponding to equation (4), we measured the ability of international salmon abundance indices to predict $\log_e(\text{SAR})$ before and after 1991.

To evaluate large-scale ocean climate effects we used the Multivariate El Niño Southern Oscillation (ENSO) Index in May–September, the Pacific Decadal Oscillation (PDO) in November–March and May–September (<http://jisao.washington.edu/pdo/PDO.latest>;

Mantua and Hare 2002), the Aleutian Low Pressure Index (ALPI; Trenberth and Hurrell 1995), and the Northern Oscillation Index (NOI; Schwing et al. 2002).

To evaluate more local effects we examined the sea surface temperature (SST) data from Chrome Island in the Strait of Georgia (www-sci.pac.dfo-mpo.gc.ca/osap/data/SearchTools/Searchlighthouse_e.htm) and the upwelling index at 48°N (<ftp://orpheus.pfeg.noaa.gov/outgoing/upwell/monthly/upindex.mon>). We were limited by the availability of local biological indices with time series that included the 1980s and relied upon estimates of Pacific Herring *Clupea pallasii* abundance in the Strait of Georgia (J. Schweigert, DFO Nanaimo, personal communication). Beamish et al. (2012) demonstrated that recent herring recruitments covaried with Sockeye Salmon survivals.

Autocorrelation and heterogeneous variance.—After examining partial autocorrelation plots for all the time series, regressions from equations (7–9) were recomputed using generalized least squares, specifically the CorAR1 function within the regression routine gls provided by the package nlme in the R statistical language (Zuur et al. 2009; Pinheiro et al. 2013), to estimate first-order autocorrelation and to correct regression standard errors and probabilities. Heterogeneity of variances within R_2 , S_2 , and SAR₂ was addressed using the weights described in the Appendix. These weights were also used as a variance covariate via the VarConstPower function with gls. In theory, a GAM smoother, autocorrelation coefficients, and weights for heterogeneous variance can all be calculated simultaneously (e.g., GAMM, R package mgcv, R Development Core Team 2012). Zuur et al. (2009) recommended fitting a variance model first, then a predictors model. As a methodological note, we found that fitting a GAM smoother in the predictors model while simultaneously correcting for autocorrelation in the variance model resulted in these components of the regression competing for temporal patterns in variance. It was not clear how that trade-off should be resolved.

RESULTS

Summary Statistics and Time Series

Time series of smolts (S), spawners (EFS), and returns (R) varied by a factor of about 50, with numbers of S_1 always exceeding S_2 (Figure 2A, B) and numbers of R_1 almost always exceeding R_2 (Figure 2C, D). On average, 4.3% (SD = 4.7) of each brood of Chilko Sockeye Salmon left after a second year in freshwater (Figure 1), but occasionally the estimates for that fraction were much higher: 25.8% for brood 1965, and 16.7% for brood 1970. Age-3 smolts were only recorded recently, two each in 2009 and 2010, and were not considered in this analysis. Fish of both smolt ages usually spent 2 years at sea, but a few returned after 1 or 3 years. On average, 94.6% of the Chilko Lake Sockeye Salmon returning as adults entered the sea as S_1 . Within that group, 1.1% were $R_{1,1}$ (jacks) that returned to freshwater

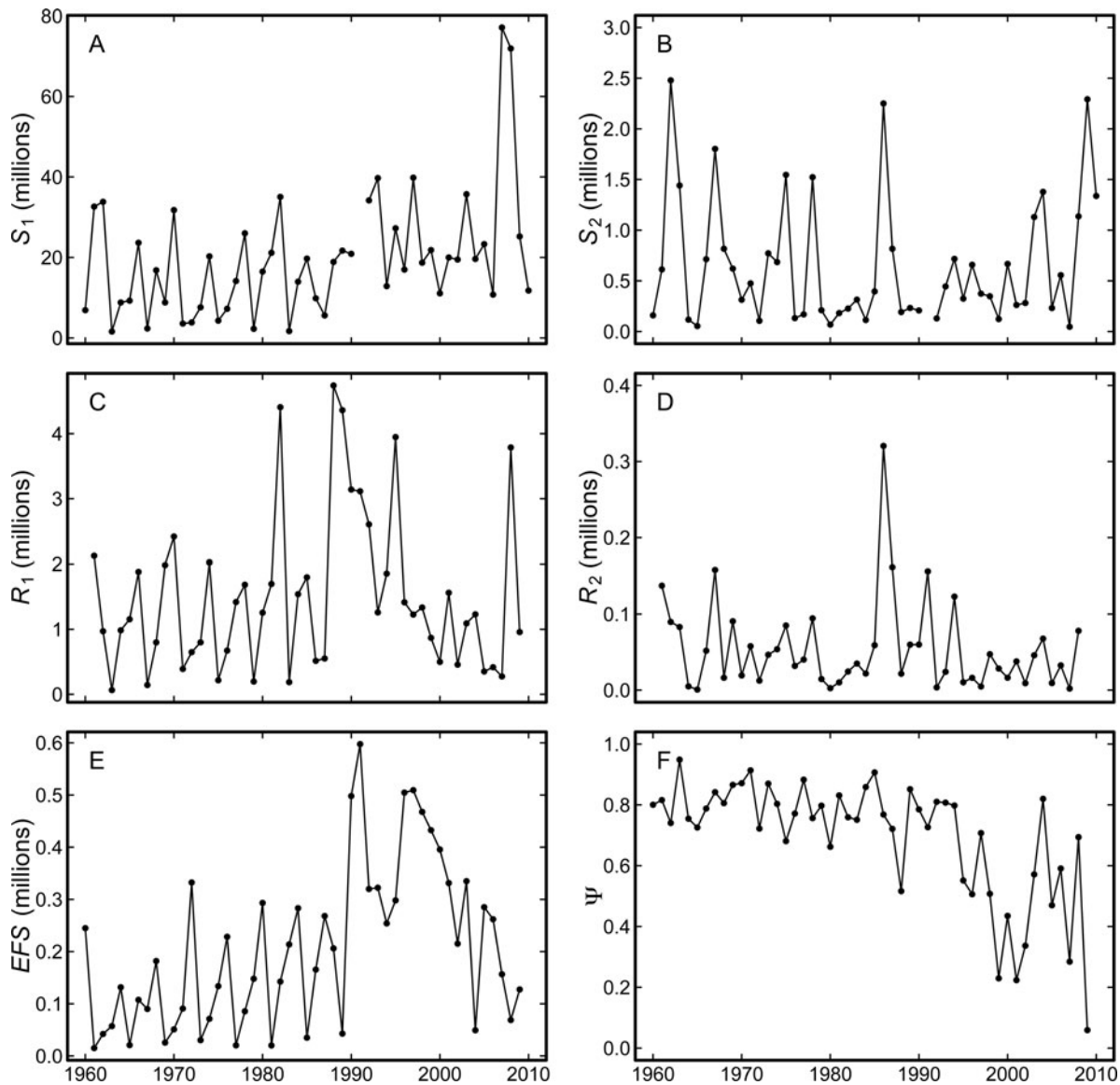


FIGURE 2. Time series of (A) age-1 Sockeye Salmon smolts, S_1 , by ocean entry year, (B) age-2 smolts, S_2 , by ocean entry year, (C) adult returns of all ages from S_1 , R_1 , by ocean entry year, (D) returns of all ages from S_2 , R_2 , by ocean entry year, (E) effective female spawners, EFS , by brood year, and (F) ψ , by return year is the fraction of returns of all ages killed by fisheries (primarily), upstream migration, and prespawning mortality.

after 1 year in the ocean, 93.4% were $R_{1,2}$ that returned after 2 years, and 5.4% were $R_{1,3}$ (Figure 1). Similarly for S_2 , the proportions were 4.2% $R_{2,1}$, 93.3% $R_{2,2}$, and 2.4% $R_{2,3}$. Smolt mean lengths varied by years (Table 1) but S_2 were, on average, 40% longer than S_1 and 300% heavier. Smolt condition statistics were almost identical for both age-groups (we did not assume isometric growth).

Precision of Age-2 Smolt Estimates

The proportion of S_2 in six recent age samples ranged from 0.6% to 48% and the corresponding BCLs for S_2 abundance within age samples reflected that variability. Because of length-stratified sampling for ages, the proportion of S_2 in age samples

was 2–11 times greater than the proportion in the population of smolts (Table 2). For instance, the Pacific Salmon Commission (PSC) estimated that 46,940 S_2 from brood 2004 left Chilko Lake in 2007, mixed in with 77,130,000 S_1 from brood 2005. In 2007, a sample of 2,067 smolts, of which 12 were S_2 (Table 2), was aged. The 95% BCLs for those 12 S_2 were 6 and 19 or about $\pm 50\%$ of the count in the age sample. Based on the ratio of 12 S_2 observed to 46,940 S_2 estimated, the lower and upper 95% BCLs for S_2 leaving Chilko Lake in 2007 were 23,471 and 74,325, respectively. This sampling variance for S_2 amplified the sampling variance for $SAR_2 = R_2/S_2$ because S_2 was the denominator. In 2010, as a comparison, 1,072 S_2 were observed in a sample of 3,827, or about 28%, and the 95% BCLs are only

TABLE 2. Age-2 Sockeye Salmon smolt composition of age sample and of all smolts, 2005–2010. The 95% binomial confidence limits (BCLs) for S_2 in age sample are based on proportions within each age sample. The difference in the proportion of S_2 in age samples compared with the proportion of S_2 in estimated smolts reflects stratified sampling by length because S_2 are, on average, longer than S_1 .

Ocean entry year	Estimated smolts ($\times 10^6$)	Sample size	S_2 counted in age sample	BCLs for S_2 counted	S_2 in age sample (%)	S_2 in estimated smolts (%)
2005	23.54	2,174	81	64–99	3.7	1.0
2006	11.32	4,535	416	378–454	9.2	4.9
2007	77.18	2,067	12	6–19	0.6	0.06
2008	73.05	946	157	135–180	16.6	1.5
2009	27.51	1,871	894	852–936	47.8	8.3
2010	13.11	3,827	1,072	1,018–1,127	28.0	10.2
Mean	37.61	2,570	439		17.65	4.34
SD	29.71	1,340	447		17.76	4.19

± 55 , or about $\pm 5\%$. The estimate for S_2 abundance for 2010 was much more precise than the one for 2007. The PSC estimates of abundance at age involved the application of length-at-age matrices (an age–length key) to length frequency distributions, so BCLs as calculated for Table 3 only indicated the relative precision between years of abundance-at-age estimates.

Precision of Adult Returns from Age-2 Smolts

For 10 recent years of available data, between 0.18% and 0.48% of the Chilkco Lake Sockeye Salmon returning each year were sampled for age determination. The median sample size was about 700 (Table 3), one-third of the median sample size for smolt ages. If 20 R_2 are observed in a sample of 700, then the 95% BCLs for the number of R_2 that might be observed in repeated similar samples are 14 and 30. The upper confidence limit for an estimate of SAR_2 would therefore be 50% higher than the estimate before considering uncertainty from S_2 and any other sources of uncertainty. If as few as seven R_2 were observed

with 95% BCLs of 4 and 15, the upper confidence limit for SAR_2 would be over 200% greater than the estimate. Four out of the 10 age samples had counts for $R_{2,2}$ of less than 20 (Table 3), which suggests that roughly half of the $R_{2,2}$ estimates have high sampling variance (and there are additional sources of sampling variance). The sampling variance was also frequently large for $R_{1,3}$ and $R_{2,2}$ in these age samples, and the counts for $R_{2,3}$ were low enough (0–3) that the abundance estimates for that age-class were overwhelmed by sampling variance. The binomial sampling variance for $R_{1,2}$ was much lower than for $R_{2,2}$. The $R_{1,2}$ counted in 2009 had 95% BCLs of less than $\pm 3\%$, and the worst case in the 10 years we analyzed was 2002 at $\pm 5.6\%$.

Of 393 returns aged in 2009, three were $R_{2,2}$ from brood year 2004 and ocean entry year 2007, with 95% BCLs of 0–7. That suggested a range of $\pm 100\%$ in the estimate for $R_{2,2}/S_2$ for ocean entry year 2007, even before considering the sampling variance for S_2 (which was $\pm 50\%$). We concluded that the 2007 data for age-2 smolts and their returns provided a poor estimate

TABLE 3. Age-2 Sockeye Salmon smolt composition in returns, 2000–2009. The 95% binomial confidence limits (BCLs) are based on age counts considered separately within each age sample, as opposed to a multinomial confidence limit. R_2/R_1 is the ratio of returns from age-2 smolts ($R_{2,1} + R_{2,2} + R_{2,3}$) to returns from age-1 smolts ($R_{1,1} + R_{1,2} + R_{1,3}$).

Year	Estimated returns ($\times 10^6$)	Sample size	R_2/R_1 (%)	$R_{2,2}$ count in age sample	BCLs for $R_{2,2}$ count	$R_{2,3}$ count in age sample	BCLs for $R_{2,3}$ count
2000	1.40	699	10.8	64	49–79	0	0–3
2001	0.85	706	5.7	36	25–48	2	0–5
2002	0.65	586	6.7	36	25–48	1	0–3
2003	1.56	747	2.2	16	9–24	0	0–3
2004	0.55	567	3.1	15	8–23	2	0–7
2005	1.08	711	6.9	46	34–59	0	0–3
2006	1.28	711	8.3	54	41–68	1	0–3
2007	0.44	719	2.6	18	10–27	0	0–3
2008	0.45	811	7.4	55	41–69	1	0–3
2009	0.27	393	1.6	3	0–7	3	0–7
Mean	0.85	665	5.45	34		1.00	
SD	0.45	119	2.92	25		1.05	

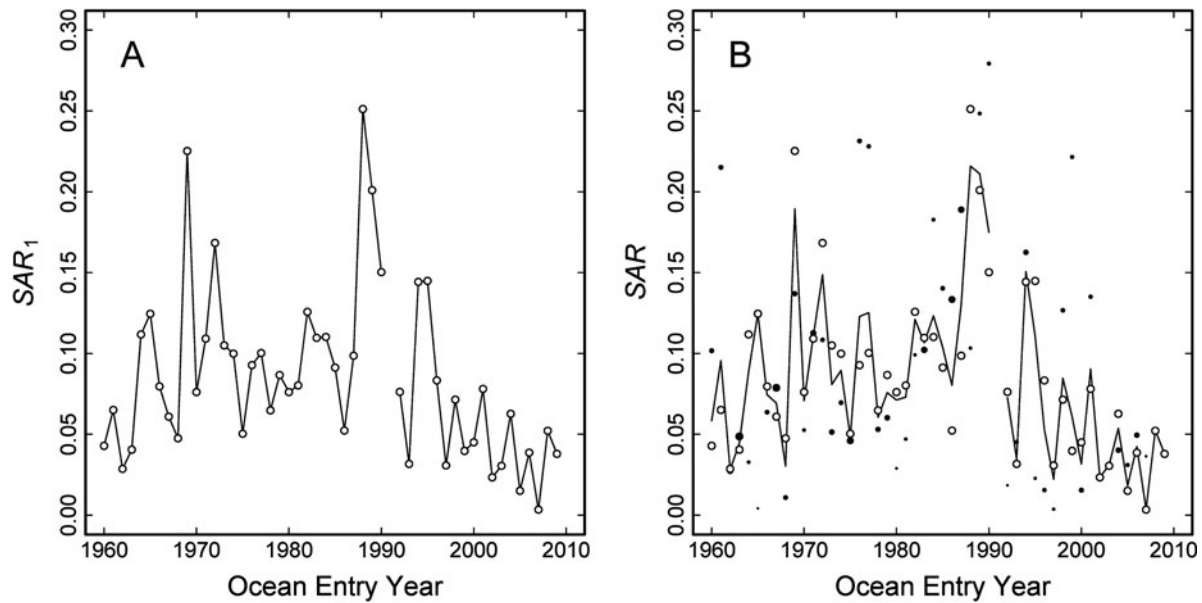


FIGURE 3. (A) Sockeye Salmon smolt survival time series 1960–2009 as the ratio of survivors ($R_{1,1} + R_{1,2} + R_{1,3}$) to the corresponding age-1 smolts, by ocean entry year, (B) combined SAR (line) from adding SAR_2 estimates (solid dots) with varying weights to SAR_1 estimates (open circles). The size of dots is proportional to the weights applied to SAR_2 .

of smolt survival that could not be compared with the unusually low survival of age-1 smolts that also entered the ocean in 2007.

Comparison of SAR_1 and SAR_2

The accepted measure of Chilkco Lake Sockeye Salmon smolt survival, SAR_1 , shows substantial year-to-year variability as well as decadal trends (Figure 3A). The mean of age-2 smolts to adult returns (SAR_2) was 16% higher than for SAR_1 (Table 1), but this was not statistically different (paired t -test: $P = 0.20$; when log transformed, $P = 0.57$). Neither SAR_1 nor SAR_2 was correlated by brood year ($n = 45$, $df = 35$, $r^2 = 4\%$, $P = 0.24$), but both were weakly correlated by ocean entry year ($n = 45$, $df = 39$, $r^2 = 18\%$, $P = 0.09$). Deleting the 2007 outlier in SAR_1 had little effect. Both smolt survival time series (Figure 4A) showed similar trends: survival generally rose from 1960 until 1990 and decreased thereafter. Based on the determination that one GAM smoother was as good as two (ANOVA: $P [>F] = 0.29$), the low-frequency (decadal) signals in SAR_1 and SAR_2 were indistinguishable. The temporal trends for SAR were calculated from log-transformed data (Figure 4A, B) because those GAM models had normally distributed residuals. Although SAR_1 and SAR_2 were weakly correlated, the first principal component explained 71% of the variance of both despite some of the SAR_2 estimates being poor (but note that when $r^2 = 0$, PC_1 will capture 50% of the variance). These results encouraged us to find an estimate for SAR that combined the estimates of SAR_1 and SAR_2 , as described in the Appendix.

Weighting SAR_2 and combining SAR_1 and SAR_2

The age distribution of smolts by ocean entry year (OEY) and the age distribution of their survivors were compared as logits,

plotting $\log(R_{2,2}/R_{1,2})$ against $\log(S_2/S_1)$. The result (Figure 5A) confirmed our hypothesis that the age distribution of smolts was consistent with the age distribution of returns for those years when S_2 and their survivors, $R_{2,2}$, had relatively low sampling variance. When S_2 was below 3% of S_1 (below -1.5 on the x -axis of Figure 5A), the scatter began to increase, indicating that the relationship was being lost due to high sampling variance. This result was in agreement with the binomial sampling variance for the cases where the sample sizes were known (Tables 2, 3), e.g., below 20/700 for $R_{2,2}$. The median weight (Table 4; Appendix) was 0.42, so many cases were strongly downweighted (Figure 5B, D). Low weights for ocean entry years 1965 (related to visual counts in 1967), 1980, 1992, and 2007 essentially removed their influence. The strongest downweighting (six lowest weights) were for SAR_2 values below 0.05, so one of the effects of weighting was to remove spuriously low values of SAR_2 that may be due to underestimates for $R_{2,2}$. High values of SAR_1 in 1969 and 1972 were outliers compared with adjacent years and these were reduced by including SAR_2 (Figure 3B). Not all the extreme values of SAR_2 were strongly downweighted; the six highest values of SAR_2 had weights near the median weight. High values of SAR_2 with high weights in 1986 and 1987 were particularly influential and reinforced the pattern of high SAR_1 in the late 1980s. Correlation between SAR_2 and SAR_1 improved after applying these weights (Figure 6) from $r^2 = 17\%$ to $r^2 = 22\%$ and log transformations increased the weighted correlation to $r^2 = 33\%$. Deleting all cases with below-median weight had a similar effect on the correlation (from $r^2 = 18\%$ to $r^2 = 22\%$). Deleting the cases with the lowest 25% of weights increased the proportion of SAR_1 and SAR_2 variance explained by the first principal component from 71% to 83%.

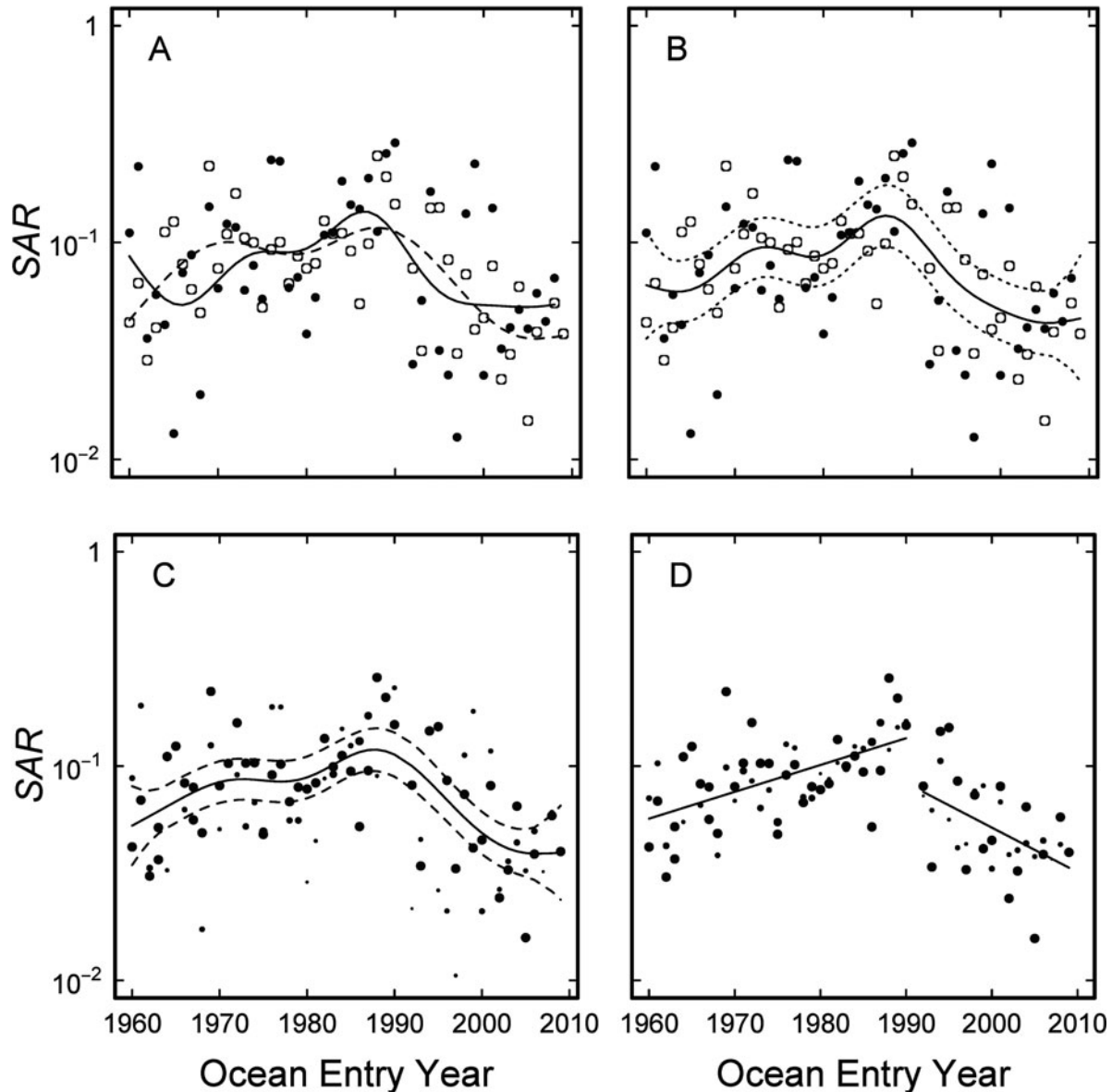


FIGURE 4. (A) GAM smoothers showed similar trends for survival of Sockeye Salmon S_1 (open circles and dashed line) and S_2 (solid dots and solid line). Survivals were \log_{10} transformed to provide normal residuals. (B) A single smoother for both ages is statistically indistinguishable from the two smoothers in plot A. Dashed lines are ± 2 SE. (C) A smoother for the effect of ocean entry year in equation (8) was a highly significant component of that model. (D) The fit using two trend lines is statistically indistinguishable from the model in plot C ($P = 0.79$) and suggests a sharp break in trends near 1991. Residuals were shifted in proportion to their influence. This discontinuous model was used to examine environmental covariates. The size of dots is related to the weights applied to SAR_2 .

We examined the effect on SAR of downweighting SAR_1 by 50% for years with low values of $R_{1,2}$ as suggested by Bradford et al. (2000). There was almost no effect because SAR_1 and SAR_2 estimates were similar in most of those ocean entry years (1963, 1967, 1975, 1979, 1983, 1987; see Table 4) although the value for 1987 increased from 13.0% to 14.6%. We did not weight SAR_1 or SAR on this basis, and we had no other indicators for the reliability of $R_{1,2}$ estimates.

Density Dependence

The conventional fit of R_1 and S_1 to a Ricker curve (equation 4; Figure 7A) indicated that density dependence was a signifi-

cant effect in S_1 survival ($r^2 = 25\%$, $P = 0.0002$ when the 2007 outlier is included). The resulting estimates for Ricker parameters α and β (0.11 and -0.024 , respectively) were sensitive to the 2007 outlier, such that maximum returns were 1.7×10^6 at 41×10^6 S_1 before deleting the 2007 case and 2.7×10^6 at 77×10^6 S_1 after. The regression was less convincing after correcting for autocorrelation ($r^2 = 7.8\%$, $P = 0.070$). As expected because of low densities, R_2 and S_2 considered by themselves (Figure 7B) did not show density-dependent survival ($P = 0.13$). Our novel consideration of S_2 survival in relation to S_1 density (Figure 7C) also did not show density-dependent survival ($P = 0.17$). We also determined that the slope from a regression of \log

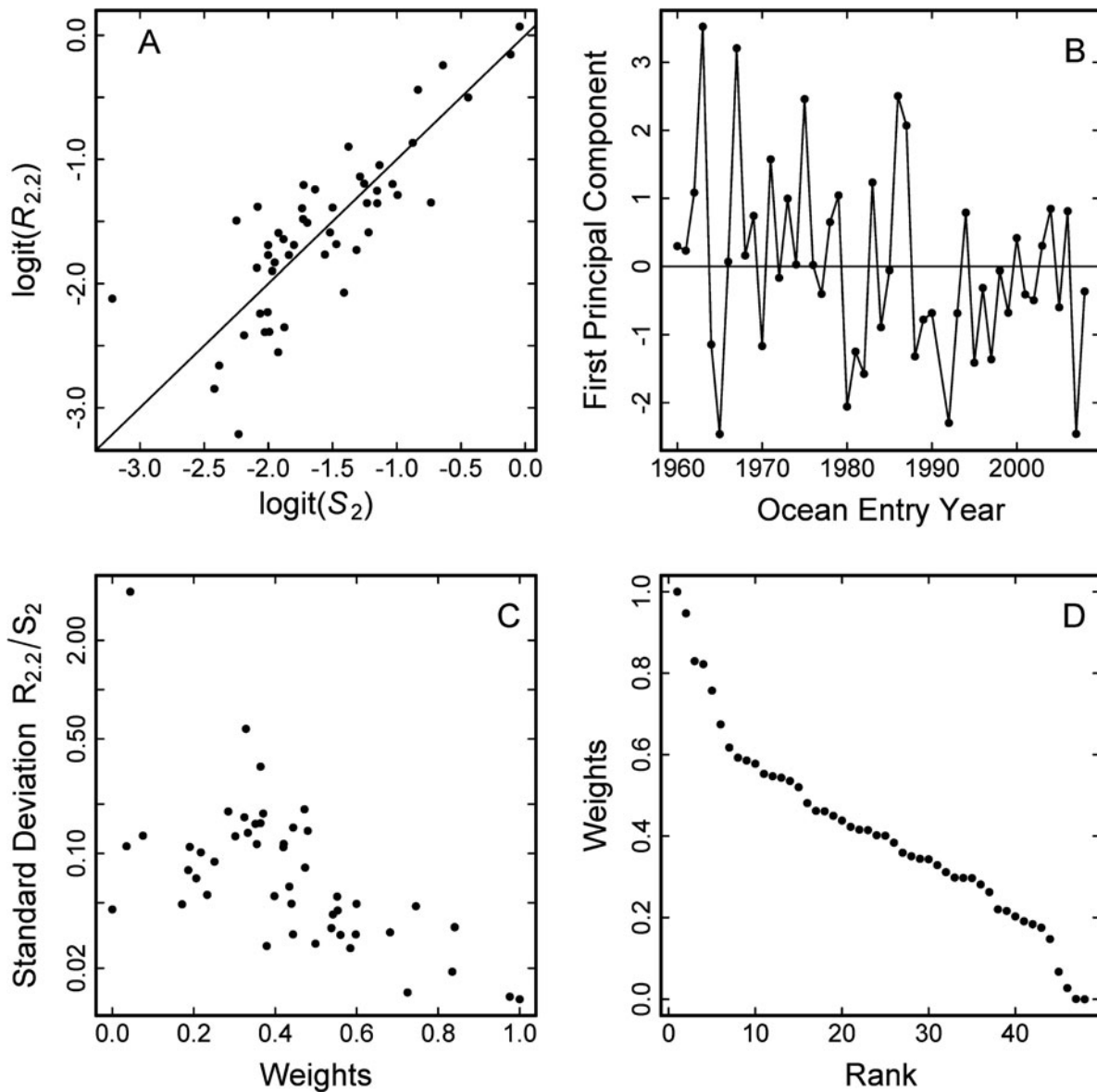


FIGURE 5. (A) The ratio of ages in Sockeye Salmon returns was consistent with the ratio of ages in smolts but the variance increased as proportions of S_2 and R_2 decreased. The term $\text{logit}(S_2)$ means $\log_{10}(S_2/S_1)$. The line is 1:1, not fitted. (B) The first principal component of the relationship in plot A was used as the weights for survival of age-2 smolts, SAR_2 . (C) These weights are similar to a CV for SAR_2 based on equation (A.5). This is a log-log plot. (D) The resulting weights in rank order. Six cases from before 1991 were the most reliable, in contrast to four cases (1965, 1980, 1992, and 2007) with weights low enough to essentially remove their effect in regressions (see also Table 4).

R on $\log S$ equation (9) was statistically indistinguishable from 1.00, indicating no density dependence (this result is described below).

Regime Effects

Survivals based on the SAR estimates (SAR_1 and SAR_2 combined) differed among four ecosystem regimes (ANOVA based on equation 7: $P < 0.001$). Mean SAR (as percent) increased from 8.6% during 1960–1976 to 10.7% during 1977–1988 and 10.2% during 1989–1998, and then decreased to 4.1% during

1999–2008 (Table 5). After correction for multiple comparisons, the only significant pairwise differences indicated that 1999–2011 survival was lower than for 1977–1988 ($P = 0.008$) and 1989–1998 ($P = 0.028$). Adjusting the regimes to change at 1991 instead of 1989 showed that SAR (in percent) dropped by 4.3% (2.2) between 1977–1990 and 1992–1998, but this was not a significant difference ($P = 0.17$).

Fitting a weighted smoother to the smolt and adult returns time series (equation 8; Figure 4C) indicated a trend of increasing survivals from 1960 to 1990 without a clear change at 1977,

TABLE 4. Percent survival from smolt to adult returns by ocean entry year (OEY) for age-1 and age-2 Sockeye Salmon smolts (SAR_1 and SAR_2), the estimated reliability of SAR_2 expressed as a weight ranging from zero to one, and SAR calculated as the weighted geometric mean of SAR_1 and SAR_2 .

OEY	SAR_1	SAR_2	Weight	SAR	OEY	SAR_1	SAR_2	Weight	SAR
1960	4.29	11.1	0.48	5.84	1985	9.13	14.9	0.42	10.37
1961	6.51	22.4	0.47	9.55	1986	5.23	14.2	0.84	8.03
1962	2.87	3.6	0.60	2.82	1987	9.86	19.8	0.75	13.03
1963	4.05	5.8	1.00	4.45	1988	25.11	11.2	0.21	21.59
1964	11.18	4.2	0.23	8.88	1989	20.10	25.7	0.30	21.11
1965	12.46	1.3	0.00	12.46	1990	15.02	28.8	0.32	17.49
1966	7.96	7.3	0.43	7.45	1991				
1967	6.09	8.8	0.97	6.93	1992	7.62	2.7	0.03	7.27
1968	4.75	2.0	0.44	3.05	1993	3.17	5.4	0.33	3.47
1969	22.53	14.6	0.53	18.95	1994	14.42	17.2	0.55	15.06
1970	7.61	6.2	0.25	7.07	1995	14.48	3.2	0.17	11.07
1971	10.92	12.2	0.68	11.07	1996	8.33	2.5	0.38	5.28
1972	16.83	11.7	0.39	14.86	1997	3.08	1.3	0.19	2.23
1973	10.50	6.0	0.58	8.08	1998	7.14	13.6	0.42	8.47
1974	9.99	7.8	0.44	8.95	1999	3.98	23.0	0.33	6.08
1975	5.03	5.5	0.83	4.84	2000	4.50	2.4	0.50	3.17
1976	9.28	24.0	0.44	12.29	2001	7.81	14.4	0.36	9.04
1977	10.03	23.7	0.37	12.53	2002	2.34	3.2	0.35	2.35
1978	6.48	6.2	0.54	6.05	2003	3.06	4.1	0.47	3.10
1979	8.67	6.9	0.59	7.58	2004	6.27	4.9	0.55	5.36
1980	7.61	3.8	0.07	7.12	2005	1.51	4.0	0.35	1.83
1981	8.02	5.6	0.21	7.30	2006	3.87	5.8	0.56	4.24
1982	12.58	10.8	0.19	12.11	2007	0.34	4.5	0.04	0.38
1983	10.97	11.1	0.72	10.66	2008	5.21	6.2	0.36	5.23
1984	11.02	19.2	0.29	12.33	2009	3.79			3.79

then a trend of lower and decreasing survivals from 1992 to 2008 (there is no estimate for 1991). Survival effects from ecosystem regime changes in 1977 and 1999 appeared, in this case, to be artifacts from describing continuous trends as discrete means.

Fisheries Management

As well as the change in SAR after OEY 1991, the effects of fisheries management created large changes to postlake survival of Sockeye Salmon beginning in the 1990s (Table 6). Until 1990 there were consistently high losses between recruits and spawn-

TABLE 5. Mean and SE for SAR (%) of Sockeye Salmon for ocean climate regimes beginning in 1977, 1989, and 1999. The last two rows are similar to rows 2 and 3, but assume that an ocean survival regime started in 1991 instead of 1989, as indicated by the trends in SAR.

Regime	N (years)	Mean	SE
1960–1976	17	8.6	1.1
1977–1988	12	10.7	1.2
1989–1998	9	10.2	2.2
1999–2008	10	4.1	0.8
1977–1990	14	11.9	1.3
1992–1998	7	7.6	1.7

ers (Ψ) and a consistent 4-year cycle of low returns and low escapements (1953, 1957, . . . 1985, 1989) (Figure 2E, F). Starting in 1990, escapement suddenly increased and subsequently remained high for every year with the exception of 2004. The mean escapement from 1990 to 2008 (0.33×10^6) was 250% of that from 1960 to 1989 (0.13×10^6). There were no low escapements comparable with the 1953 cycle line for 14 years after 1989. Starting in 1995, Ψ decreased from about 80% to 50% and remained low until 2009, our last estimate. In 2009, approximately 94% of adult returns from the ocean reached Chilko Lake because the fishery was immediately closed when the record-low smolt survival from ocean entry year 2007 was recognized (Figure 2F).

Smolt Length, Condition, and Age

Mean lengths and weights of age-1 smolts (S_1) were not significantly different among time periods (Table 7). On average, S_2 were longer and correspondingly heavier after 1991, largely due to S_2 in 2007, when the mean length of S_2 was 160 mm. During the 1977–1991 period of high survivals, S_1 were in significantly better condition than during the 1992–2009 period of low survivals (t -test: $P = 0.003$). Condition of S_2 was also better before 1991 than after ($P = 0.06$).

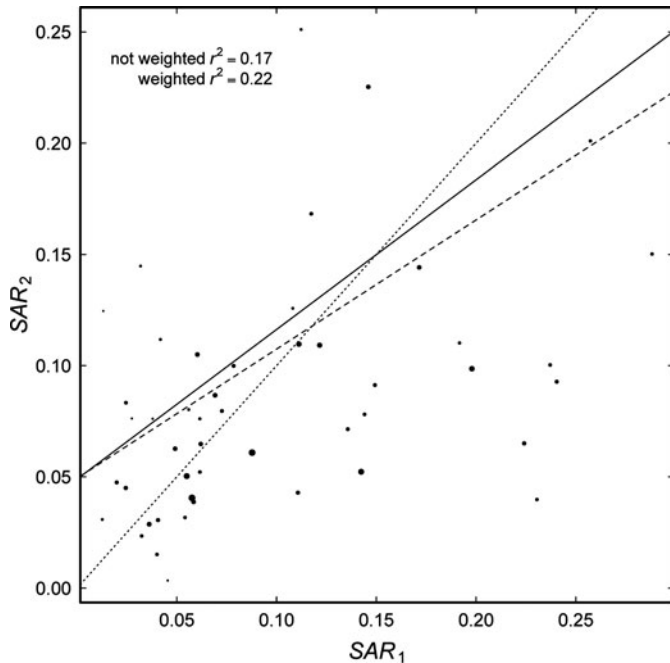


FIGURE 6. Correlation of the survival of age 2 Sockeye Salmon smolts, SAR_2 , with the survival of age-1 smolts, SAR_1 , that had the same ocean entry year. The correlation was 17% (solid line) but increased to 22% (dashed line) after applying weights. The size of dots is proportional to the weight assigned to SAR_2 . The dotted line is 1:1, not fitted.

The “base case” for analyzing the effect on smolt survival of smolt quality (age, length, and condition) and time trends was the plot of $\log_e(R)$ versus $\log_e(S)$ corresponding to equation (10), Figure 8, and model A in Table 8. The 2007 outlier was deleted

TABLE 6. Prespawning loss (primarily exploitation), Ψ , and female escape-ment statistics for Sockeye Salmon before and after 1994.

Period	Loss of recruits before spawning (Ψ) (%)		Effective female spawners (EFS) ($\times 10^6$)	
	Mean (SD)	Range	Mean (SD)	Range
1952–1994	78 (8)	52–95	156 (14)	0.007–0.60
1995–2009	49 (22)	6–82	293 (15)	0.05–0.51

for this analysis. Plots of partial autocorrelation for the residuals of the full model (row D in Table 8) showed that autocorrelation was not significant, and fitting a first-order autocorrelation model resulted in a small value ($\phi = 0.024$) for off-diagonal elements of the residuals covariance matrix. We proceeded without including a correction for autocorrelation in these regressions. Heterogeneous sampling variance in R_2 and S_2 was explored further by using our weights (Figure 5D) as a variance covariate (a predictor of the variance of residuals but not a predictor of R_2 ; Zuur et al. 2009) to fit new estimates of sampling variance in a generalized least-squares regression. The result was similar to our weights, but suggested a smaller variance (higher weights) for the most reliable cases of R_2 and S_2 . This supported our parameterization of sampling variance, and we proceeded with our weights as the more conservative approach. Fitting autocorrelation and heterogeneous variance models simultaneously showed little interaction. Plots of residuals (adjusted for weights) against fitted values for the full model showed no trends, and the residuals had a near-normal distribution.

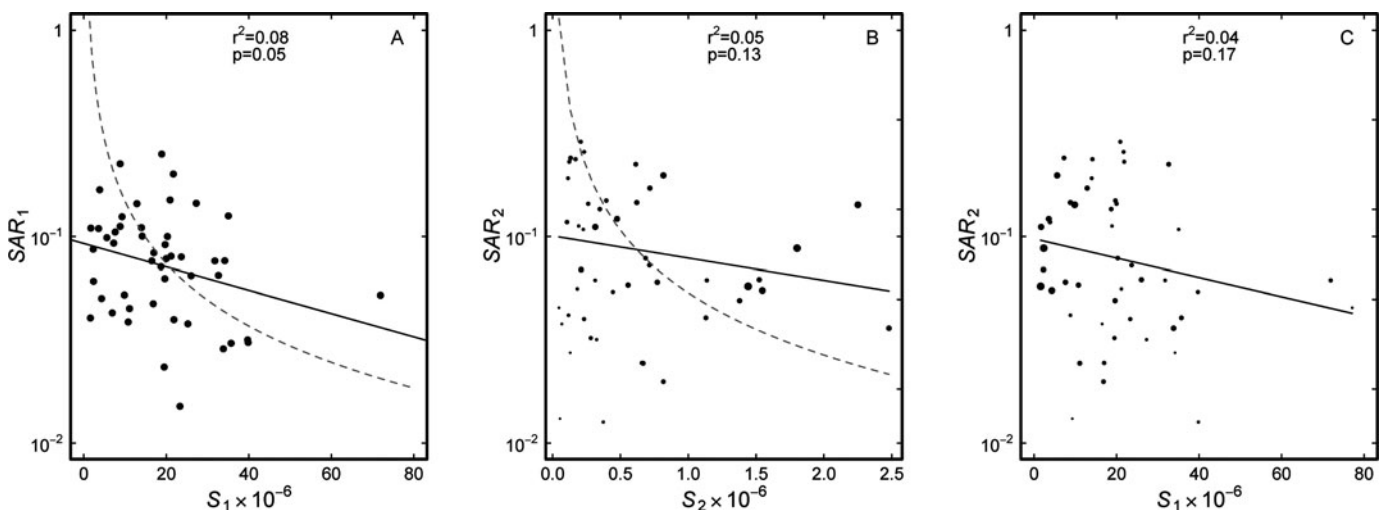


FIGURE 7. Examination of density dependence in the survival of Sockeye Salmon smolts to adult returns, expressed as $\log_e(SAR)$ based on the Ricker model equation (4). Solid lines are from linear regression, and the curved dashed lines correspond to no relationship between smolts and corresponding adult returns. (A) The survival of age-1 smolts, SAR_1 , as a function of their abundance upon leaving Chilko Lake is marginally significant before considering autocorrelation. The outlier for SAR_1 in ocean entry year 2007 was not included. (B) Age-2 smolt survival, SAR_2 , was not significantly related to the low densities of age-2 smolts. The size of the dots is proportional to the weights used in the regression. (C) Age-2 smolt survival was examined as a function of age-1 smolt density in the same ocean entry year, but that weighted regression was not significant.

TABLE 7. Length (mm), weight (g), and condition (%) of Sockeye Salmon smolts by age and by selected ranges of ocean entry years. Weights are from means of preserved samples. Condition is deviation (%) of observed from predicted mean weight based on means of preserved lengths.

Smolt age (years)	Smolt quality metric	Year range	N		
			(years)	Mean	SE
1	Length	1950–1976	18	83.1	1.61
		1977–1991	14	82.96	1.46
		1992–2009	19	84.0	1.39
1	Weight	1950–1976	25	4.53	0.25
		1977–1991	15	4.72	0.27
		1992–2009	18	4.86	0.28
1	Condition	1950–1976	25	96	1.1
		1977–1991	15	105	1.5
		1992–2009	18	99	0.7
2	Length	1950–1976	14	114.24	2.19
		1977–1991	14	114.97	2.06
		1992–2009	18	121.44	3.52
2	Weight	1950–1976	23	11.68	0.42
		1977–1991	12	13.87	1.28
		1992–2009	17	16.32	1.54
2	Condition	1950–1976	23	97	1.4
		1977–1991	12	106	2.1
		1992–2009	17	100	0.9

Covariates such as age and length were interpreted as a multiplicative effect on survival rates in these models. We expected that accounting for the effects of climate trends and smolt abundance would reveal patterns in the residual variance that were due to smolt quality; however, we could not discern a significant effect from smolt age, length, or condition in the regressions with various combinations of covariates (Table 8). Linear temporal trends broken at 1991, without effects from smolt quality (trial B in Table 8), provided the best model as identified by the AIC and explained 91% the returns across both age-classes

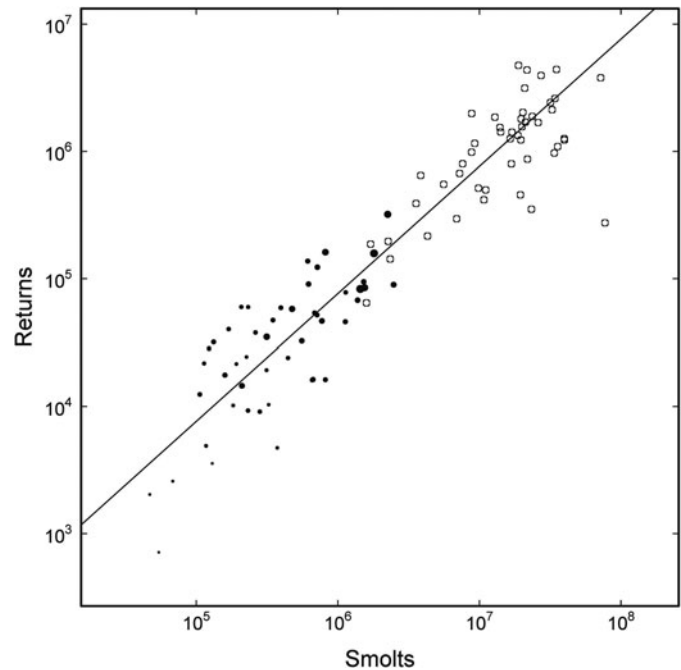


FIGURE 8. A single line on a \log_{10} - \log_{10} plot relates Sockeye Salmon smolts to returns over three orders of magnitude in abundance. The line shown has a slope of 1.00 and the fitted intercept corresponds to 7.55% mean survival (95% CL, 6.63% to 8.61%). Both smolt ages are included; S_1 are open circles and S_2 are solid dots. The size of the dots is proportional to the weights assigned to SAR_2 .

of smolts (Figure 4D). For this model, the fitted parameter for $\log_e(S)$ was 0.955 (0.033), within two SEs of 1.00, implying that there was no curvature in the smolts-to-adult returns relationship. When the parameter for $\log_e(S)$ was forced to 1.00 (trial G), the change in r^2 and AIC was small and the difference between 0.95 and 1.00 for this parameter could have arisen due to sampling variance (ANOVA between trials B and G: $P = 0.17$).

TABLE 8. Regression results from examining smolt age, smolt length, and time trends as predictors of Sockeye Salmon smolt survival. In the model notation, S is $\log_e(\text{smolts})$ of both ages, R is $\log_e(\text{returns})$ from those smolts, L is the mean length of smolts (by age), A is a factor for smolt age (1 or 2), T is a factor indicating whether ocean entry year (OEY) is before or after 1991 for fitting two trends with time, and $\exp(S)$ is the Ricker term for density dependence. The number of cases is 93 for all models, weights for R_2 and S_2 were applied, and the 2007 outlier for R_1 was excluded. Trial D is the full model that was used to examine autocorrelation and other patterns in residuals. Trials with smolt condition (K) were deleted from this table because it was never a significant effect when added to these models.

Trial	Model	r^2 (%)	AIC	Test	$P(>F)$
A: base	$R \sim S$	86.4	193.6		
B: add time trends	$R \sim S + OEY \times T$	91.0	161.6	B:A	<0.001
C: add smolt age	$R \sim S + A + OEY \times T$	91.1	162.8	C:B	0.40
D: add length	$R \sim S + L + A + OEY \times T$	91.1	164.6	D:C	0.69
E: drop smolt age	$R \sim S + L + OEY \times T$	91.0	163.5	E:D	0.36
F: drop time trends	$R \sim S + L + A$	86.8	194.7	F:D	<0.001
G: slope of $S = 1.00$	$R \sim S + OEY \times T$	91.5	161.5	G:B	0.17
H: slope of $S = 1.00$	$R \sim S + \exp(S) + OEY \times T$	91.2	162.2	G:H	0.21

TABLE 9. Abundance of Sockeye Salmon returns by age-groups (mean, SD, CV [SD/mean]) and correlations between returns within a brood but after differing lengths of time at sea. Degrees of freedom (df) for correlations reflect correction of probabilities for autocorrelation.

Age-group	Abundance ($\times 10^6$)			Correlation				
	Mean	SD	CV	Comparison	<i>N</i>	df	r^2 (%)	<i>P</i> (> <i>t</i>)
1.1	0.0134	0.0157	1.17	1.1–1.2	51	23	10	0.12
1.2	1.372	1.133	0.825	1.2–1.3	51	33	36	0.0001
1.3	0.0822	0.115	1.40	1.1–1.3	51	23	<1	0.98
2.1	0.00175	0.00349	1.99	2.1–2.2	50	27	3	0.36
2.2	0.0495	0.0564	1.138	2.2–2.3	50	35	9	0.05
2.3	0.00107	0.00184	1.71	2.1–2.3	50	27	<1	0.91

The model in trial G predicted $\log_e(SAR)$ without density dependence and the parameter estimates (with SEs) were $\log_e(R/S) = -58.81(14.69) + 0.02857(0.00744) \times OEY$ before 1991, and $\log_e(R/S) = 92.59(43.21) - 0.04777(0.0217) \times OEY$ after 1991. When we added $-\beta S$, the Ricker term for density dependence (trial H), the change in r^2 and AIC was small and the difference was not significant (ANOVA between trials G and H: $P = 0.21$). After removing temporal trends, intrastock density-dependent survival of Chilko Lake Sockeye Salmon smolts was undetectable.

Marine Survival after the Ocean Entry Year

Precocious males (i.e., jacks, $R_{1,1}$ or $R_{2,1}$) were poor predictors of the subsequent abundance of their siblings, $R_{1,2}$ and $R_{1,3}$ or $R_{2,2}$ and $R_{2,3}$ (Table 9). There was a significant positive correlation between $R_{1,2}$ and subsequent $R_{1,3}$, both of which were abundant and estimated relatively precisely. Similarly, $R_{2,2}$ and $R_{2,3}$, although less abundant, were also correlated. We pursued the correlation between $R_{1,2}$ and $R_{1,3}$, noting that it was affected by three factors: (1) shared dependence on S_1 , (2) year-to-year variation in the proportions that return after 1, 2, and 3 years at sea (a separate effect from survival), and (3) shared mortality in the first 2 years after leaving Chilko Lake. After removing the contribution of S_1 to both $R_{1,2}$ and $R_{1,3}$, both time series shifted to lower survivals after 1991, but numerous independent outliers in both time series resulted in a nonsignificant correlation ($P = 0.29$). There was no evidence of an effect from a common ocean entry year between successive returns within broods.

Changes in Sea Residence

We examined how the age distributions of returns varied with time (Figure 9) because year-to-year variability in duration of sea life contributed to the residual variance of the regressions described above. Duration of marine life for age-1 smolts increased from 2.015 (0.005) years during 1948–1967 to 2.092 (0.007) years during 1989–2008, a 3.8% (0.88) change in 40.5 years. This effect was almost entirely because the proportion of $R_{1,3}$ increased 350% from 2.7% (0.84) in 1948–1967 to 9.5% (1.16) in 1989–2008. There was a linear trend of 30.5% (10.0) per century in the proportion of $R_{1,3}$ from 1974 to 2009 (Figure 9B).

In the same period, the proportion of $R_{1,1}$ (jacks) decreased by 75% from 1.2% (0.21) to 0.3% (0.05) (Figure 9A). The equivalent pattern of older returns from S_2 is less clear (Figure 9D); however, the proportion of returns from S_2 that were $R_{2,1}$ (jacks) dropped to nearly zero after 1994, whereas proportions greater than 10% had been frequent previously (Figure 9C).

The Marine Environmental Signal

Corresponding to the temporal pattern of SAR for Chilko Lake Sockeye Salmon (Figure 4D), there was a break in trends near 1991 for many indices of the density of salmon in the North Pacific Ocean that might have been competing with Chilko Sockeye Salmon (columns 1 [North America] and 3 [Asia] of Figures 10, 11). For North America, this included catches of Sockeye and Chum Salmon, and hatchery releases of Sockeye, Pink, and Chum Salmon, but Pink Salmon catches continued to increase after 1991. For Asia, this included catches of all three species, and the trends broke near 1987 for Sockeye Salmon and 1994 for Chum Salmon. Asian hatchery releases of Sockeye Salmon and particularly the massive releases of Chum Salmon changed near 1991, but Pink Salmon releases were variable and did not show a clear break. Reliable estimates of hatchery releases of Sockeye Salmon by Russia (and therefore Asia) before 1991 were not available, but the numbers released were probably low. Climate indices (columns 1 and 3 in Figure 12), in contrast to salmon density indices, did not show a pattern break near 1991 with the exception of PDO (winter and summer) and ENSO. Coastal indices of upwelling, SST, and Strait of Georgia Pacific Herring did not have trends with a break near 1991.

To compare trends in SAR , with its substantial year-to-year variance, with trends in climate and salmon density, we examined the correlation of SAR with each of the 24 indices, before and after 1991 (columns 2 and 4 in Figures 10–12). For North America, there were weak but significant correlations (at $\alpha = 0.05$) with catches (column 2 in Figure 10) and hatchery releases (column 2 in Figure 11) of competing salmon species before 1991 (except for catch of Chum Salmon and hatchery release of Sockeye Salmon) but only Chum Salmon hatchery releases after 1991 were close to significance ($P = 0.07$). For Asia there was a significant correlation of $\log_e(SAR)$ with catches of Sockeye

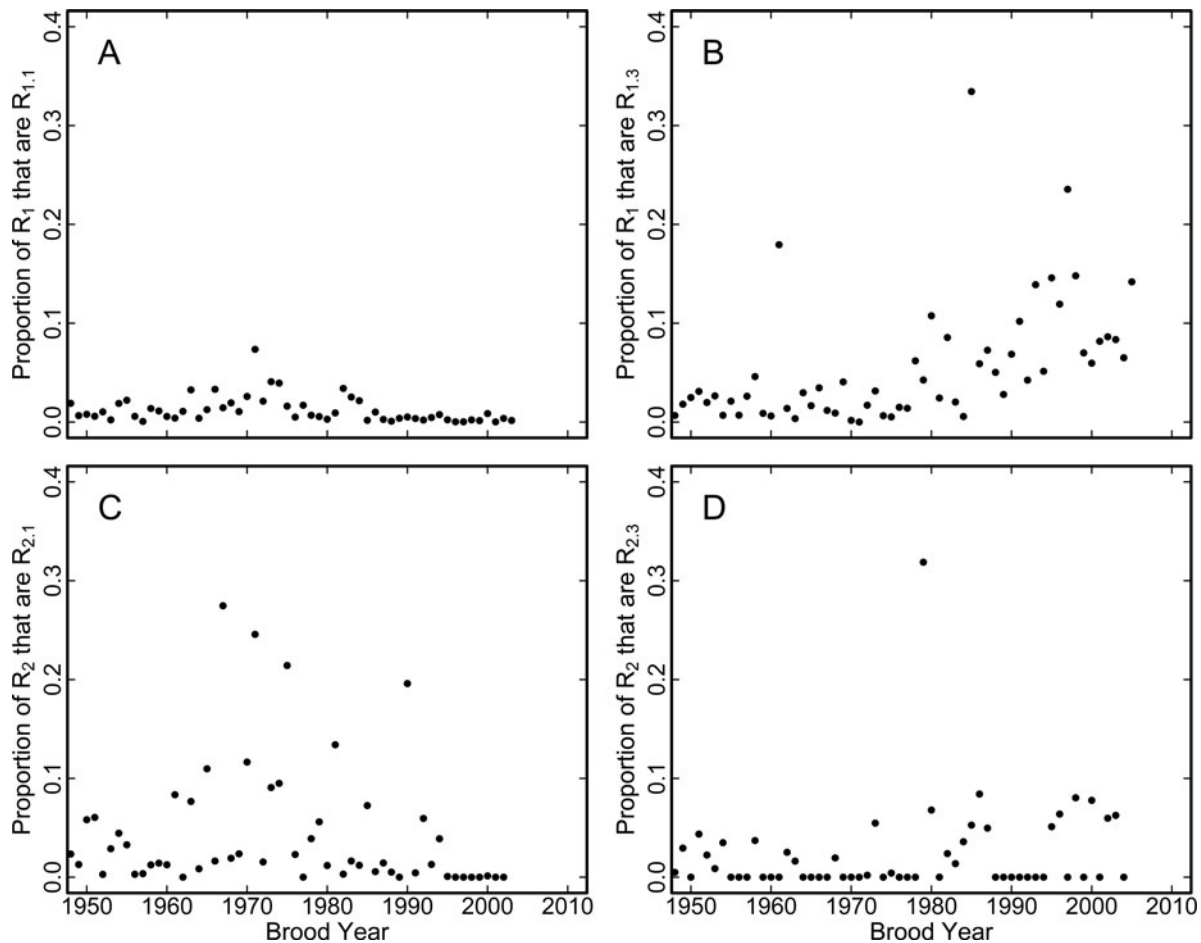


FIGURE 9. Proportion of Sockeye Salmon smolts that returned after 1 or 3 years at sea instead of the usual 2 years. (A) Age-1 smolts that returned as $R_{1,1}$ after 1 year at sea, (B) Age-1 smolts returning as $R_{1,3}$ after 3 years at sea had a significant linear trend after 1970 of increasing proportions of $R_{1,3}$ [$-5.99(1.97) + 0.00305(0.0010) \times \text{Year}$, $r^2 = 0.22$, $P = 0.0041$], (C) Age-2 smolts that returned as $R_{2,1}$ after 1 year at sea have rarely been observed after 1995, whereas previously they were frequently a large proportion of returns from S_2 , and (D) Age-2 smolts that returned as $R_{2,3}$ after 3 years at sea were sufficiently rare that a trend corresponding to that of $R_{1,3}$ was not apparent.

Salmon before 1991 (column 4 in Figure 10) that suggested interstock density dependence, but evidence of this after 1991 was weak. Asian Chum Salmon catches were also correlated with $\log_e(SAR)$ before 1991 but the slope was positive, indicating an effect on both from increasing ocean productivity, but that relationship disappeared after 1991. Sockeye Salmon catches in Asia before the early 1980s were primarily of Russian and North American fish by Japanese fishers, but afterwards, Russia caught increasing proportions of the catch, and now virtually all the Sockeye Salmon caught in Asia are by Russia (Irvine et al. 2012). The only evidence of SAR being depressed after 1991 by hatchery releases in Asia was a marginal correlation with Pink Salmon (column 4, row 2 in Figure 11; $P = 0.053$). The only significant correlation of SAR with climate and coastal indices that we examined was from coastal upwelling during summer at 48°N after 1991, despite there being no trend in that index and no correlation before 1991 (column 2, row 4 in Figure 12).

DISCUSSION

Results from our examination of 50 years of information on two age-classes of Sockeye Salmon smolts emigrating from and returning to Chilko Lake helped us understand reasons underlying variability in abundance and survival for Sockeye Salmon in general. When arranged by ocean entry year, survivals of age-1 and age-2 smolts were weakly positively correlated, suggesting that each was a noisy estimate of survival. To understand the error associated with these survival estimates, we used a novel principal components approach to estimate the relative sampling variance for uncommon age-2 smolts and their survivors, allowing a weighted combination of survival estimates from the two smolt ages. The combined time series reduced the sampling variance in the original age-1 smolt survivals sufficiently to indicate a linear trend of increasing smolt survival during 1960–1990 that suddenly changed at or near 1991 to a lower and declining trend in survival during 1992–2008.

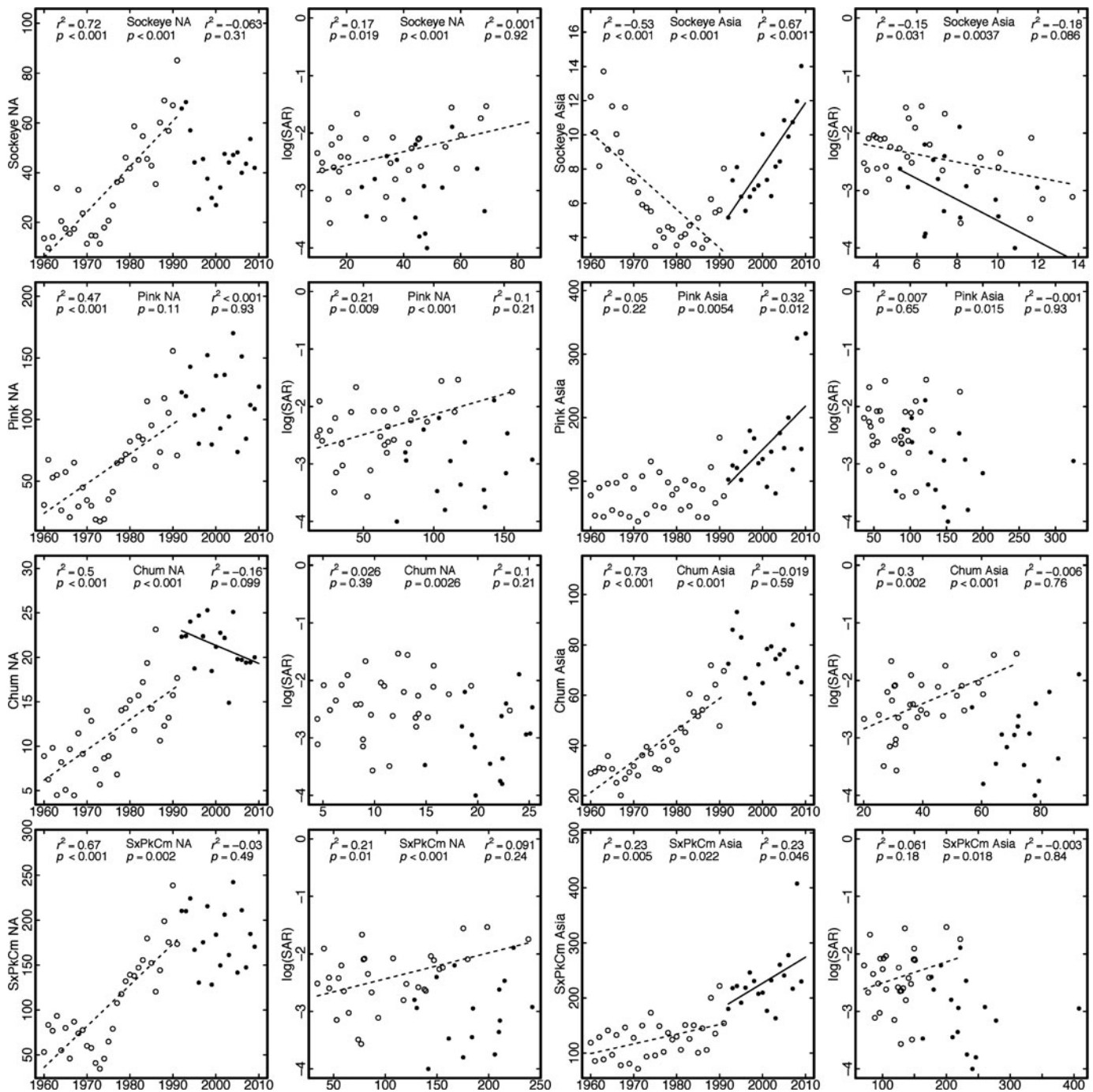


FIGURE 10. Landings (millions of fish) of Sockeye (Sx), Pink (Pk), and Chum (Ch) Salmon from North America (NA) and Asia compared with the temporal pattern of SAR and as predictors of $\log_e(SAR)$. SxPkCm is the total for the three species. The first and third columns of plots are the time series with fitted linear trends for different variables; r^2 and probability for a linear trend before 1991 (open circles, dashed lines) are in the top left in each plot, and the statistics for trends after 1991 (solid dots, solid lines) are top right. Lines are drawn when $P < 0.20$. The probability from ANOVA that two trend lines are statistically equivalent to a single trend is the P -value, top center. The second and fourth columns are scatter plots and regression lines that predict $\log_e(SAR)$ based on the same variables (see equation 4), with statistics for fits before 1991 in the top left and statistics for fits after 1991 in the top right. Negative r^2 values indicate negative slopes.

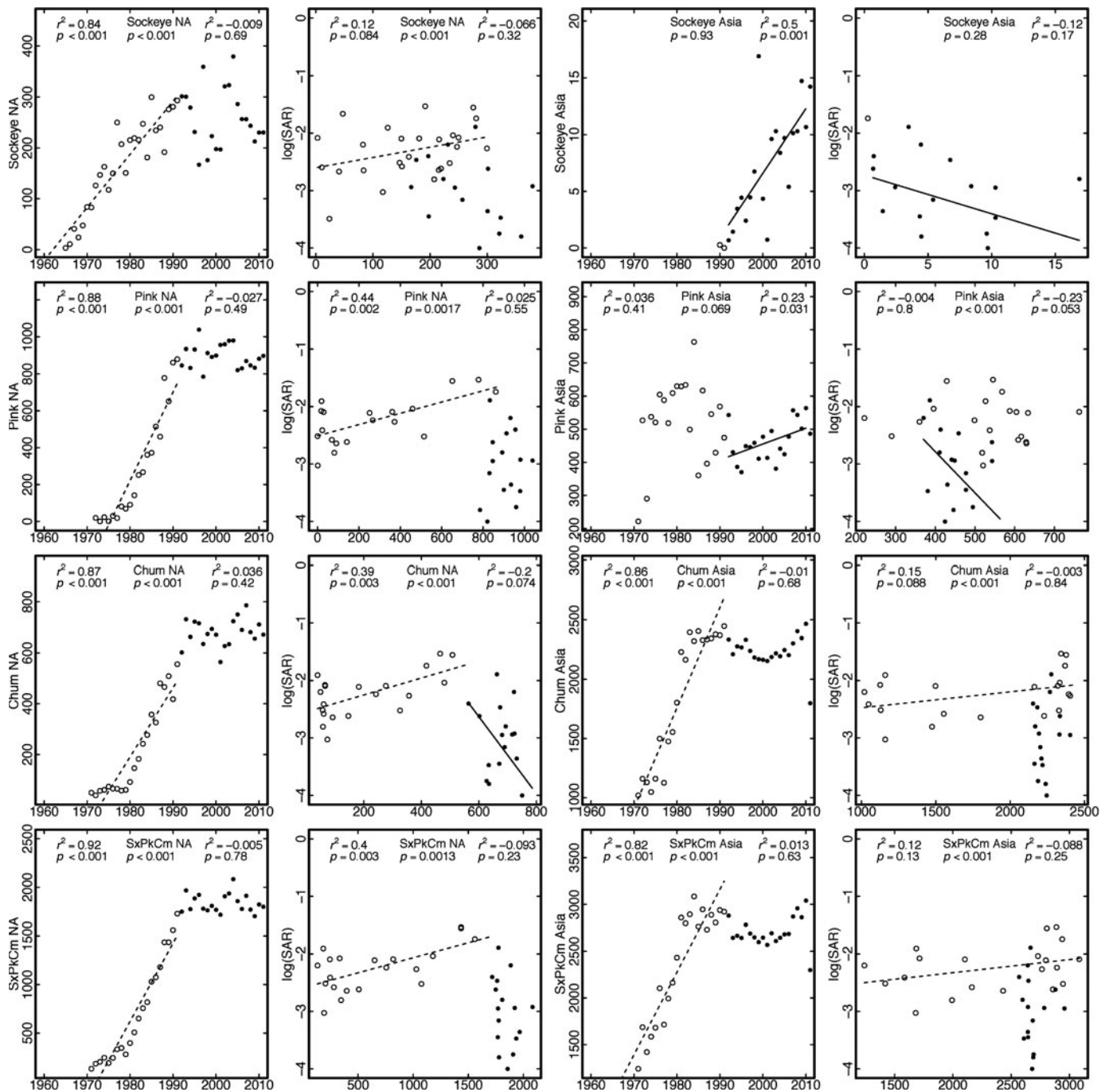


FIGURE 11. Hatchery releases (millions of fish) of Sockeye (Sx), Pink (Pk), and Chum (Ch) Salmon from North America (NA) and Asia compared with the temporal pattern of SAR and as predictors of $\log_e(SAR)$ as in Figure 10. SxPkCm is the total for the three species. The first and third columns of plots are the time series with fitted linear trends for different variables; r^2 and probability for a linear trend before 1991 (open circles, dashed lines) are in the top left in each plot, and the statistics for trends after 1991 (solid dots, solid lines) are top right. Lines are drawn when $P < 0.20$. The probability from ANOVA that two trend lines are statistically equivalent to a single trend is the P -value, top center. The second and fourth columns are scatter plots and regression lines that predict $\log_e(SAR)$ based on the same variables (see equation 4), with statistics for fits before 1991 in the top left and statistics for fits after 1991 in the top right. Negative r^2 values indicate negative slopes.

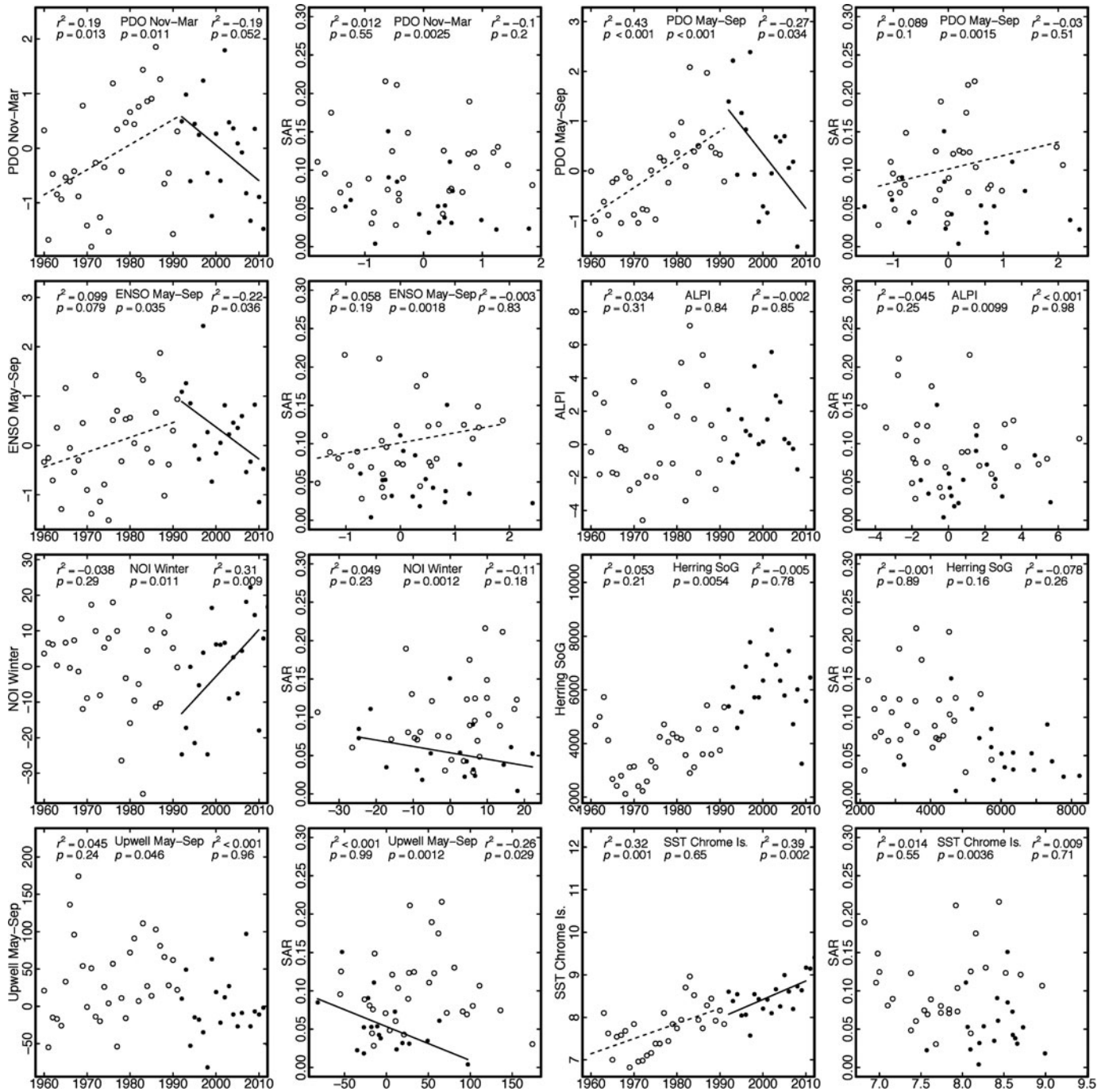


FIGURE 12. Environmental variables (see text for explanations of abbreviations) compared with the temporal pattern of Sockeye Salmon SAR and as predictors of SAR. The first and third columns of plots are time series and fitted temporal trends for different variables; r^2 and probability of zero correlation for the fitted trend before 1991 (open circles, dashed lines) are in the top left in each plot, and the statistics for trends after 1991 (solid dots, solid lines) are top right. Lines are drawn when $P < 0.20$. The probability from ANOVA that two trend lines are statistically equivalent to a single trend is the P -value, top center. The second and fourth columns are scatter plots and regression lines that predict SAR based on the same variables, with statistics for fits before 1991 in the top left and statistics for fits after 1991 top right. Negative r^2 values indicate negative slopes.

The judicial inquiry on Fraser River Sockeye Salmon was precipitated in part by exceptionally low Sockeye Salmon returns in 2009 (Cohen 2012a, 2012b, 2012c). The survival of Chilko Lake Sockeye Salmon that went to sea as a record high abundance of age-1 smolts in 2007 and returned as $R_{1,2}$ in 2009 was anomalously low in our analyses (i.e., an outlier), consistent with the extremely low returns that year for Fraser River Sockeye Salmon in general. We were unable to compare this low survival of age-1 smolts with the survival of age-2 Chilko Lake smolts that also entered the ocean in 2007 and returned in 2009 because of the sampling variance associated with low counts for age-2 smolts within the age samples. In contrast to Fraser River Sockeye Salmon, survivals for many non-Fraser River Sockeye Salmon (e.g., Columbia River, west coast of Vancouver Island, central and northern British Columbia) that returned in 2009 were not abnormally low (K. D. Hyatt, DFO, Pacific Biological Station, personal communication). It appears that the low survivals of Fraser River Sockeye Salmon returning in 2009 were primarily a consequence of 2007 marine conditions in the Strait of Georgia that also affected other species including Pacific Herring (Beamish et al. 2012), as well as conditions to the north of the Strait of Georgia, especially within the Queen Charlotte Sound–Hecate Strait region (Thomson et al. 2012). The high returns to Chilko Lake in 2010, coincident with the highest returns to the Fraser River watershed in a century, were the consequence of high egg-to-smolt survival in freshwater yielding large numbers of smolts going to sea in 2008 followed by average (compared with long term) smolt-to-adult survivals.

Insignificant differences between survivals of age-1 smolts and the larger age-2 smolts were unexpected. Koenings et al. (1993) found that 30% of the variation in Sockeye Salmon smolt survival was explained by smolt size, much more than smolt age, as well as a south-to-north cline of increasing survivals. When Henderson and Cass (1991) compared scale-based growth rates for 3 years of age-1 Chilko Lake Sockeye Salmon smolts to those of returning adults, there was evidence that larger age-1 smolts survived better than smaller smolts in the same cohort. However, as in our study, Henderson and Cass (1991) found no significant effect of mean size on the survival of age-1 smolts, based on 34 years of data. In a more recent examination of growth patterns for Fraser River Sockeye Salmon, McKinnell and Reichardt (2012) found that early marine growth, measured as size at the end of the first marine growing season of returning adults, had a minor effect on interannual survival differences. It appears the complex relationship between smolt size and survival is affected by many factors including smolt age and timing and stream latitude.

Our finding of increased sea residence for Chilko Lake Sockeye Salmon agreed with the findings of others who have documented increasing numbers of older Sockeye Salmon recruiting in recent years (e.g., Holt and Peterman 2004). Increased sea residence was consistent with a reduction in first-year marine growth for Fraser River Sockeye Salmon commencing in 1977,

which occurred in spite of concurrent increases in biological productivity in the coastal marine zone (McKinnell and Reichardt 2012).

A long-standing puzzle in Fraser River Sockeye Salmon biology is the explanation for cyclic dominance, where some Fraser River Sockeye Salmon populations exhibit a pattern of strong returns every 4 years, often with a subdominant year and two weak years in between, while other populations do not (e.g., Ricker 1950; Welch and Noakes 1990). In the case of Bowron Lake (mid-Fraser River watershed), Sockeye Salmon demonstrated a pattern of cyclic dominance from 1959 to 1982, but not before or after that time; reasons for this shift are not clear (Walters et al. 2004; Grant et al. 2011). We found that before 1990, Chilko Lake Sockeye Salmon exhibited clear cycles marked by a weak return every 4 years (i.e., 1961, 1965, ... 1989). This pattern changed suddenly after 1989 when spawner escapement doubled as a result of unusually high recruitment during 1990–1994 (predominately R_1 , Figure 2C) and then reduced fishing during 1995–2009 (Ψ , Figure 2E). The pattern of cyclic dominance before 1990 influenced our weights for survival estimates for age-2 smolts. During the period of cyclic dominance, large escapements one year followed by small escapements the next year resulted in a large proportion of age-2 smolts comigrating with relatively few numbers of age-1 smolts. This age distribution was repeated in the adult returns 2 years later. We used that consistency in age distributions to parameterize sampling variance for SAR_2 . In ocean entry years with a high ratio of age-2 to age-1 smolts, survival estimates for age-2 smolts were the most reliable and had a strong influence (heaviest weights) on the combined survival estimates (SAR). These weights may be revised if or when additional sample size information becomes available.

Declining survivals for Chilko Lake Sockeye Salmon smolts after 1991 corresponded to decreasing productivity (returns per spawner) noted about that time for many North American Sockeye Salmon stocks outside of central and western Alaska (Peterman and Dorner 2012). We have no real explanation for differences among populations in the timing of declines although stock-specific migration patterns (Tucker et al. 2009) or differences in relative strengths of particular brood years may be important. Our results suggest that this general pattern of decline for Sockeye Salmon resulted from factors downstream from rearing lakes, which is in agreement with findings from a 2010 workshop on Fraser River Sockeye Salmon (Peterman et al. 2010). Although the influence of factors between the lake and the ocean on survival estimates cannot be excluded, the wide geographic distribution of Sockeye Salmon populations with similar patterns in productivity (Peterman and Dorner 2012) suggests that freshwater factors are unlikely to have played a major role in the widespread decline.

The onset of declining survival about 1991 also corresponded approximately to a shift in climate regimes from a period of above-average productivity for many populations of Pacific salmon to one that was less productive (e.g., Hare and Mantua

2000; Irvine and Fukuwaka 2011). Our analysis did not suggest breaks in the survival time series corresponding to ecosystem regime shifts in 1977 and 1999.

After removing temporal trends, we found no evidence of an intrastock density-dependent effect on smolt-to-adult survival. Similarly, McKinnell and Reichardt (2012) found no relationship between the numbers of smolts leaving Chilko Lake and their subsequent first year of marine growth. In the ocean, individuals from any one salmon population will potentially compete with fish from many other populations, so one might expect density-dependent effects in the ocean to be due to interstock competition. A logical extension of this concept would be to replace estimates of Chilko Sockeye Salmon smolt numbers with estimates of Sockeye Salmon smolts from the Fraser River and perhaps other rivers draining into the northern California Current. In this paper we extended the concept even further to include interspecies competition and investigated correlations between Sockeye Salmon survival and abundance estimates of Sockeye, Pink, and Chum Salmon in the eastern and western North Pacific Ocean.

It is interesting that trends in Asian Sockeye Salmon catches reversed in the 1980s (row 1, column 3 in Figure 10) and that these trends were correlated with smolt survivals in Chilko Sockeye Salmon (row 1, column 4 in Figure 10). Declining catches before 1990 were chiefly the result of high catches of Russian and North American Sockeye Salmon in the high seas by Japan early in the time series. In contrast, increasing catches after 1990 were primarily the result of increasing catches of Russian Sockeye Salmon by Russia. Are Russian fisheries the real reason for the apparent declines in survivals of North American Sockeye Salmon after 1990? Using data from Irvine et al. (2012, their Tables 1 and 7), we determined that since 1990, Sockeye Salmon catches by Russia have been increasing and in recent years have constituted 25–30% of the total North Pacific Sockeye Salmon catch. These estimates may be low; Dronova and Spiridonov (2008) reported that Sockeye Salmon catches in Kamchatka were 1.5–3 times higher than reported. However, Russian salmon fisheries primarily occur close to Russia, and although some Canadian Sockeye Salmon are found in the Bering

Sea (Habicht et al. 2010; Beacham et al. 2011), their numbers in the western North Pacific Ocean are low. We therefore think it is unlikely that declining North American Sockeye Salmon survivals after 1991 are a consequence of these salmon being caught in Russian fisheries and not included in marine harvest estimates. As an alternate explanation, the correspondence of Chilko Lake Sockeye Salmon survival to the patterns of North American and Asian catches and hatchery releases suggests that density-dependent processes in the North Pacific may be affecting Sockeye Salmon survival.

Peterman and Dorner (2012) suggested that mechanisms controlling Sockeye Salmon marine survival patterns would likely (1) operate at large spatial scales or where many Sockeye Salmon populations overlap; (2) affect populations from Puget Sound to southeastern Alaska similarly, while having an inverse effect on stocks from central and western Alaskan; or (3) have been present historically, but intensified recently. With this in mind, and using results from our examination of correlations between various data sets and our new time series supplemented by results from the literature, we developed a conceptual model to explain how changes in ocean climate and salmon species abundance might control the survival and growth of Sockeye Salmon (Table 10).

In our model, we suggest that increasing biological productivity before 1991 resulted in greater carrying capacity for salmon. Also during this period, expanding hatchery programs yielded greater Pink, Chum, and Sockeye Salmon biomass. Responses to more salmon competing for more resources included generally improved survivals for Sockeye Salmon but negative responses in terms of growth (Table 10). Survival was essentially independent of growth during this period. After 1991, reduced ocean productivity resulted in declining carrying capacity. Continuing increases in Pacific salmon numbers and biomass, largely from hatcheries, further intensified inter- and intraspecific competition for food. In response, Sockeye Salmon growth rates continued to decline, and as a result of many fish not achieving some critical size, survivals began to decline.

Evidence to support this model comes from various sources. Kaeriyama et al. (2009) concluded that Pacific salmon

TABLE 10. Conceptual model illustrating possible impacts of changing marine productivity, hatchery releases, and hatchery effectiveness on survival and growth responses of Sockeye Salmon from British Columbia and southern Alaska. Up arrows “↑” and down arrows “↓” indicate increases and decreases while diagonal arrows “↗” and right arrows “→” indicate minor increases and little change, respectively.

Period	Stressor	Pathway	Effect	Response
Before 1991	↑ Productivity	↑ Salmon carrying capacity	↗ Competition for food	↑ Survival ↓ Growth
	↑ Hatchery releases	↑ Salmon biomass		
After 1991	↓ Productivity	↓ Salmon carrying capacity	↑ Competition for food	↓ Survival ↓ Growth
	→ Hatchery releases	↑ Salmon biomass		
	↑ Hatchery effectiveness			

carrying capacities in the North Pacific Ocean peaked between 1985 and 1994 for Sockeye, Pink, and Chum Salmon, presumably alleviating competition effects until, in the case of Chilkco Lake Sockeye Salmon, approximately 1991. We showed earlier that release numbers from salmon hatcheries generally increased until about 1991 (columns 1 and 3 in Figure 11), and the literature provides widespread evidence of increasing Pacific salmon abundance and survivals between 1977 and 1991 (e.g., Beamish and Bouillon 1993; Mantua et al. 1997). With respect to the three primary salmon species, Chum Salmon consistently constitute the largest biomass in the North Pacific (smaller Pink Salmon are the most numerous), followed by Pink Salmon and then Sockeye Salmon (Eggers 2009). Japanese Chum Salmon migrate seasonally between the Bering Sea (summer–fall) and the Gulf of Alaska (winter–spring) (Urawa et al. 2009). Sockeye Salmon from Puget Sound to southeastern Alaska spend most of their time within the Gulf of Alaska where they overlap with, and presumably compete with many species, including Asian Chum Salmon during the winter. Recent genetic evidence also demonstrates North American Sockeye Salmon extend farther northward and westward into the Bering Sea than thought previously where they overlap with Asian Sockeye and Pink Salmon (Habicht et al. 2010). Body size for many Sockeye Salmon populations increased from the 1960s until the late 1970s, followed by decreases (Eggers and Irvine 2007). Similarly, McKinnell and Reichardt (2012) reported a decline in early marine growth from 1970 to 1990 of Chilkco Lake and Birkenhead River Sockeye Salmon, based on scale growth of survivors. We suggest that before about 1991 expanding numbers of Pacific salmon were increasingly competing for a healthy but still finite food resource, and this view is consistent with the overall finding of reduced growth rates but not reduced survival for Sockeye Salmon (Table 10).

Kaeriyama et al. (2009) concluded that carrying capacities for Pacific salmon declined after the mid-1990s. Yet continued increases in hatchery salmon releases in Asia after 1991, as well as improved survival for many hatchery fish resulting from the release of larger or better-adapted young salmon (Dushkina 1994; Morita et al. 2006; Kaev and Ignatiev 2007), resulted in increasing salmon biomass. Proportions of hatchery-origin fish in the North Pacific have been increasing since 1990 and constitute 50–62% of the total Chum Salmon, 10–13% of the Pink Salmon, and 4–10% of the Sockeye Salmon in the North Pacific Ocean (Ruggerone et al. 2010; Kaeriyama et al. 2012). Eggers (2009) estimated that at least 39% of the salmon biomass in a recent 10-year period was made up of hatchery-origin Pink and Chum Salmon. We suggest that after 1991 more salmon were competing for a reduced food supply, resulting in greater competition than before 1991 (Table 10).

Chum and Pink Salmon, which constitute the bulk of the salmon in the North Pacific Ocean, appear to have ecological advantages over Sockeye Salmon during periods of increased competition. For instance, Kaeriyama et al. (2012) found Chum Salmon were better able to alter their diet during periods of

changing prey communities than were other salmon species. Based on a detailed review of the literature, Ruggerone and Nielsen (2004) suggested that Pink Salmon are the dominant competitor among the salmon species, affecting other species by reducing the availability of prey. Ruggerone et al. (2003) found reduced Sockeye Salmon growth during their second and third year at sea in years when Asian Pink Salmon were most abundant.

After 1991 there was no evidence of further declines in Fraser River Sockeye Salmon growth rates, although the early marine growth of fish surviving to return was less than average due to earlier declines (McKinnell and Reichardt 2012). Perhaps growth rates continued to decline after 1991, but slow-growing Sockeye Salmon died at higher rates than did faster-growing Sockeye Salmon (Ruggerone et al. 2005; Farley et al. 2011) such that measured growth rates in the survivors did not decline. We suggest that Sockeye Salmon survivals began to decline after 1991 for many stocks because an increasing portion of the juveniles were unable to grow to the size (or possess sufficient energy reserves) required to survive some life history stanza, similar to the critical size hypothesis (e.g., Beamish et al. 2004a). Further evidence of increasing competition is the increased proportion of Chilkco Lake Sockeye Salmon returning late after three winters at sea instead of two. This linear trend started in the 1970s when 2% returned late, and by 2009, 12.6% returned late (Figure 9B), presumably as a result of slower growth.

Our results do little to clarify which life history stage is being affected by density-dependent effects in the ocean. The first year of marine growth for Chilkco Sockeye Salmon is greater during even years, when young Pink Salmon are abundant, than in odd years when they are not (McKinnell and Reichardt 2012), but it is not known whether this pattern continues during the Sockeye Salmon's life. Correlations between survival and abundance (catch) of Pink Salmon, lagged to align with ocean entry year, might be interpreted to indicate that density-dependent effects occurred in the first 18 months at sea for Chilkco Sockeye Salmon, the marine residence period for Pink Salmon. However, older Sockeye Salmon could also compete with Pink Salmon. There were also correlations between Sockeye Salmon survival and abundances of Sockeye and Chum Salmon as well as hatchery releases, and these fish would have shared the ecosystem with Chilkco Sockeye Salmon throughout their lives. Examining correlations at lags of 1, 2, and 3 years may help understand this issue. Ruggerone et al. (2005) hypothesized that large numbers of Pink Salmon depressed zooplankton and other prey during odd-numbered years, thereby reducing the growth rates of Bristol Bay Sockeye Salmon, primarily in their second ocean year.

Our conceptual model (Table 10) satisfies the first (large spatial scales) and third (present historically but intensified recently) requirements of Peterman and Dorner (2012). Although the linkages proposed here among ocean productivity, hatchery releases, and Sockeye Salmon survival and growth are simplistic, this model offers promise for a more complete explanation of how salmon population dynamics are linked to North

Pacific ecosystem dynamics. Further research, including the spatial and dietary overlap of marine salmon and consideration of nonsalmon competitors, is required to test the hypotheses of this model.

In conclusion, while both local- or regional-scale and ocean-basin-scale processes are important in controlling survival patterns for Pacific salmon, our detailed examination of data from one population suggests that density-dependent processes operating far from coastal North America may be important in determining long-term survival patterns for Sockeye Salmon. The importance of regional effects was evidenced by the anomalously low survival of smolts from ocean entry year 2007. We support the approach advocated by Peterman and Dorner (2012) of simultaneously evaluating multiple salmon populations, but also stress the importance of detailed examinations of populations such as Chilko Sockeye Salmon where mortality can be partitioned between events occurring primarily in the ocean versus events that occur in freshwater. We encourage other researchers to closely examine their data sets for additional time series of information that may be available, as exemplified by the relatively uncommon older smolts that we were able to include in this analysis.

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Appendix: A Proxy Model for Variance of Estimates of SAR_2

If two time series have some desired signal in common, but a low correlation because sampling error or other noise has been added to both of them, then they can be usefully combined because the noise will cancel out, but only if the noise is independent. Both SAR_1 and SAR_2 are unlikely to have completely independent noise because they are from the same field sampling operations; however, they are from different brood years (different parents), have a different freshwater history, enter the ocean at different sizes, and are tracked separately in fisheries statistics. Both time series showed substantial year-to-year variance and little covariance, so an attempt at noise reduction seemed worth pursuing.

We wished to weight estimates of SAR_2 in inverse proportion to their variance for regressions and statistical tests and wanted to identify cases where either or both of S_2 and $R_{2,2}$ had high sampling variance such that the ratio $R_{2,2}/S_2$ was an unreliable estimate of SAR_2 . The age distribution of smolts in an ocean entry year, t , should be consistent with the age distribution of returns from those smolts 2 years later, i.e.,

$$R_{2,2,t+2}/R_{1,2,t+2} = S_{2,t}/S_{1,t}, \quad (\text{A.1})$$

from which

$$R_{2,2,t+2}/S_{2,t} = R_{1,2,t+2}/S_{1,t}, \quad (\text{A.2})$$

and $SAR_2 = SAR_1$.

Consistency of age distributions is therefore a prerequisite for survival of age-2 smolts to be comparable with survival of age-1 smolts. When the binomial sampling variance of $S_{2,t}$ is large because $S_{2,t}$ are a small fraction of the smolts sampled for aging, the associated $R_{2,2,t+2}$ will also be rare and observed with large sampling variance. Consequently the observed age distribution will be highly variable and the ratio $R_{2,2,t+2}/S_{2,t} = SAR_2$ will be unreliable.

Our procedure was to plot $\log_{10}(R_{2,2,t+2}/R_{1,2,t+2})$ versus $\log_{10}(S_{2,t}/S_{1,t})$ to see whether there was consistency in the age distributions (Figure 5A). This was not a regression situation where we were attempting to predict $\log_{10}(R_2/R_1)$ or $\log_{10}(S_2/S_1)$; rather we were trying to observe the sampling variance in both. We used principal components to describe this relationship. The first principal component, PC_1 , will be a line describing the consistent relationship in age composition between smolts and returns, if it exists, and will have slope = 1 on this plot. This corresponds to equation (A.2) being true. The second component, PC_2 , will then capture the scatter of points orthogonal to this line, identifying cases that differ from the hypothesized consistent relationship. Inconsistent cases will have high loadings on PC_2 and low loadings on PC_1 . Note that PC_1 is not simply $\log_{10}(R_2/R_1)$ or $\log_{10}(S_2/S_1)$. Also, PC_1 can be used directly as regression weights, w , which was how we used it (Figure 5B), and can also be used as a “variance covariate” as a predictor for heterogeneity in the variance of residuals in generalized linear models (e.g., the function `gls` in the R package `nlme` as described by Zuur et al. 2009:chapter 4; see also Gelman et al. 2003:chapter 14.6). We explored the possibility of improving our weights by using PC_1 loadings as a variance covariate for the `varConstPower` function in GLS models and determined that only a small adjustment resulted.

That this consistency exists is a hypothesis to be tested. Subsequently, examining that consistency on a case-by-case basis will indicate the relative reliability (weight) for each estimate of SAR_2 . Changes in the duration of sea life will interfere in this scheme. We examined the variation in years at sea to see whether this is a problem (Figure 9), but note that sampling variance for age also applies to duration of sea life.

We also examined weights for SAR_2 as a theoretical variance for the ratio of binomial variables. Assuming that binomial sampling error for the age distribution is the predominant error for S_2 and $R_{2,2}$, we applied an approximation for the variance of a ratio, derived from the second-order Taylor series approximation (Stuart and Ord 1987). The resulting general formula is

$$\text{var}\left(\frac{R}{S}\right) \approx \frac{E^2(R)}{E^2(S)} \left[\frac{\text{var}(R)}{E^2(R)} + \frac{\text{var}(S)}{E^2(S)} - \frac{2\text{cov}(R, S)}{E(R)E(S)} \right], \quad (\text{A.3})$$

where $E(S)$ is the expectation of the random variable S . In the case of binomial distributions for estimates of age frequency each year, with parameters p , n , and $q = 1 - p$, the variance of

the observed number of S_2 in the age sample of n smolts is

$$\text{var}(S) = p_S q_S n_S \quad (\text{A.4})$$

in a specific year (not the variance of the proportion, pq/n) and $E(S) = np$ is the number of S_2 observed in the age sample of n smolts. Applying these binomial parameters in equation (A.3), then

$$\text{var}\left(\frac{R}{S}\right) \approx \frac{p_R q_R n_R}{p_S^2 n_S^2} + \frac{q_S p_R^2 n_R^2}{p_S^3 n_S^3}. \quad (\text{A.5})$$

Equation (A.3) can be treated more simply than this because $\text{var}(x) = E(x)$ for the binomial distribution and pqn approaches pn as p approaches 0. So when S_2 are a small fraction in the age samples,

$$\text{var}(SAR) = \text{var}\left(\frac{R_{2.2}}{S_2}\right) \approx \frac{R_{2.2}}{S_2^2} + \frac{R_{2.2}^2}{S_2^3}. \quad (\text{A.6})$$

The observed counts of S_2 appear as squared and cubed terms in denominators in equations (A.5) and (A.6), which is a warning that estimates of SAR_2 are sensitive to the sampling variance of

S_2 . We dropped the covariance term in equation (A.3) because it is not measured and thought to be small, despite the above argument that the proportions of $R_{2.2}$ and S_2 in age samples do covary, so equations (A.5) and (A.6) will overestimate $\text{var}(SAR_2)$. As well, we are applying an incomplete model when we estimate $\text{var}(SAR_2)$ based on simple binomial sampling; other sources of sampling variance include allocation of catches from mixed stocks of Fraser River Sockeye Salmon to the Chilko Lake stock and application of length-at-age matrices to smolt length distributions. Nevertheless, equation (A.5) was useful as a check for the weights derived from comparing the age distributions of smolts and returns (Figure 5C, D). To apply equation (A.5) when we did not know the size of the age samples, we used the proportions of S_2 and $R_{2.2}$ in the total smolts and total returns (not the age samples) for p , and we used the average of recent sample sizes for n .

APPENDIX REFERENCES

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